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1 Dynamic responses of the adrenal steroidogenic

2 regulatory network

- 3
- 4 Francesca Spiga^{1,†,*}, Eder Zavala^{2,3,4,†}, Jamie J Walker^{1,3,4,5,†}, Zidong Zhao¹, John R
- 5 Terry^{2,3,4}, Stafford L Lightman^{1,4}
- 6
- 7¹ Henry Wellcome Laboratories for Integrative Neuroscience and Endocrinology,
- 8 University of Bristol, Whitson St. BS1 3NY, Bristol, United Kingdom
- ² Living Systems Institute, University of Exeter, Stocker Rd. EX4 4QD, Exeter, United
- 10 Kingdom
- ³ Wellcome Trust Centre for Biomedical Modelling and Analysis, University of Exeter,
- 12 Stocker Rd. EX4 4QD, Exeter, United Kingdom
- 13 ⁴ EPSRC Centre for Predictive Modelling in Healthcare, University of Exeter, Stocker
- 14 Rd. EX4 4QD, Exeter, United Kingdom
- ⁵ College of Engineering, Mathematics and Physical Sciences, University of Exeter,
- 16 North Park Rd. EX4 4QF, Exeter, United Kingdom
- 17
- 18 † Denotes equal contribution
- 19 * Correspondence and requests for materials
- 20
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- 26 steroidogenesis.
- 27

- 28 Abstract

30	The hypothalamic-pituitary-adrenal axis is a dynamic system regulating
31	glucocorticoid hormone synthesis in the adrenal glands. Many key factors within the
32	adrenal steroidogenic pathway have been identified and studied, but little is known
33	about how these factors function collectively as a dynamic network of interacting
34	components. To investigate this, we developed a mathematical model of the adrenal
35	steroidogenic regulatory network that accounts for key regulatory processes
36	occurring at different time scales. We used our model to predict the time evolution of
37	steroidogenesis in response to physiological ACTH perturbations, ranging from basal
38	pulses to larger stress-like stimulations (e.g. inflammatory stress). Testing these
39	predictions experimentally in the rat, our results show that the steroidogenic
40	regulatory network architecture is sufficient to respond to both small and large ACTH
41	perturbations, but that coupling this regulatory network with the immune pathway is
42	necessary to explain the dissociated dynamics between ACTH and glucocorticoids
43	observed under conditions of inflammatory stress.
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55 Significance

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57 Our ability to respond to stress depends on a remarkably dynamic process of 58 hormone secretion. The rapid release of glucocorticoid hormones from the adrenal 59 glands is critical to mount such an efficient response to stress, particularly during 60 inflammation. Though many key factors involved in this process have been studied, 61 the way in which these factors interact dynamically with one another to regulate 62 glucocorticoid secretion has not been investigated. Here, we develop a mathematical 63 model of the regulatory network controlling glucocorticoid synthesis and, by 64 combining this with in vivo experiments, show how this network governs changes in 65 adrenal responsiveness in basal unstressed physiological conditions and under 66 exposure to inflammatory stress.

67 \body

68 Introduction

69

70 The hypothalamic-pituitary-adrenal (HPA) axis is a stress-responsive neuroendocrine 71 system that controls circulating levels of the vital glucocorticoid (CORT) hormones 72 corticosterone (in rodents) and cortisol (in humans). These are steroids synthesised 73 by the adrenal gland in response to stimulation by adrenocorticotropic hormone 74 (ACTH), which is secreted by the anterior pituitary in response to corticotrophin-75 releasing hormone (CRH) and arginine vasopressin released from the hypothalamic 76 paraventricular neurons (PVN). These neurons receive circadian inputs from the 77 suprachiasmatic nucleus (SCN), and are activated in response to stress. Via the 78 bloodstream, CORT accesses target tissues where it mediates metabolic, cognitive, 79 and immune responses. CORT also regulates its own production through negative 80 feedback inhibition of ACTH and CRH secretion from the pituitary and the PVN, 81 respectively. In order to mount an effective response to stress, CORT must be 82 secreted rapidly by the adrenal glands. However, because of its lipophilic nature,

CORT cannot be pre-stored into vesicles, and must therefore be rapidly synthesised *de novo* in response to ACTH stimulation.

85

86 In basal, unstressed conditions, ACTH and CORT exhibit ultradian oscillations. 87 Although there is some evidence for pulsatility of CRH (1, 2), our recent work 88 suggests that ACTH and CORT pulsatility is predominantly generated by a sub-89 hypothalamic oscillator within the pituitary-adrenal system (3, 4). The amplitude of 90 these pulses varies in a circadian manner, with larger pulses occurring at the start of 91 the active phase (morning in humans; evening in rodents). In normal physiological 92 conditions, CORT secretion is tightly correlated with ACTH (5). However, there are a 93 number of conditions where a dynamic dissociation between these hormones occurs 94 (reviewed in (6)). For example, there is evidence that pro-inflammatory cytokines 95 released during inflammation can potentiate adrenal responsiveness to ACTH, and 96 can activate the adrenal steroidogenic pathway directly (7). Recent examples of this 97 are the hormonal stress responses observed during cardiac surgery in humans and 98 in a rodent model of inflammation (8). It has been hypothesised that this dynamic 99 dissociation is due to increased adrenal sensitivity to ACTH, presumably as a result 100 of the effect of circulating pro-inflammatory cytokines. Thus, a better understanding 101 of the mechanisms that regulate steroidogenesis is necessary to explain the dynamic 102 response of CORT to both physiological and pathological ACTH perturbations.

103

Within cells of the adrenal cortex *zona fasciculata*, steroidogenesis is regulated by several processes operating over a range of timescales: there are rapid non-genomic processes, as well as slower processes that depend on gene expression. Together, these form a complex pathway activated when ACTH binds the melanocortin type-2 receptor (MC2R), resulting in rapid phosphorylation of proteins involved in cholesterol metabolism—the substrate for CORT synthesis. These include hormonesensitive lipase (HSL) and steroidogenic acute regulatory (StAR) proteins, which

111 control levels of intracellular cholesterol and its transport within the mitochondrial 112 matrix, respectively (9, 10). In addition to these rapid, non-genomic events, ACTH 113 simultaneously triggers a slower genomic response involving the expression of 114 steroidogenic genes such as StAR, CYP11A, MC2R and MRAP. These genes are 115 transcriptionally regulated by a number of nuclear receptors, whose activity and 116 expression is also under the control of ACTH. Specifically, ACTH induces the 117 "positive regulators" steroidogenic factor 1 (SF-1) (11) and Nur77 (NR4A1) (12), and 118 inhibits the "negative regulator" DAX-1 (dosage sensitive sex-reversal (DSS), adrenal 119 hypoplasia congenita (AHC) locus on the X-chromosome) (13). These interacting 120 cascades of genomic and non-genomic events work in combination to activate and 121 maintain optimal levels of steroidogenic proteins, ultimately leading to mitochondrial 122 import of cholesterol and CORT synthesis. In addition, our recent mathematical 123 modelling of the CORT response to ACTH suggests that a rapid, intra-adrenal CORT 124 negative feedback loop constitutes an additional control mechanism of 125 steroidogenesis (14-17). The biological processes underlying this self-inhibition are 126 not known, but the glucocorticoid receptor (GR) is expressed in the adrenal cortex 127 (15, 18). Furthermore, the synthetic glucocorticoid dexamethasone—a specific GR 128 agonist—has been shown to inhibit ACTH-induced corticosterone secretion from rat 129 adrenocortical cells (19), as well as the transcription of steroidogenic genes through 130 a mechanism that involves GR- and SF-1-mediated induction of DAX-1 expression 131 (20).

132

Though many of the components involved in the steroidogenic pathway have been identified, it remains unclear how their mutual interaction regulates CORT dynamics. In this study, we considered the core components of the steroidogenic regulatory network (SRN) and investigated how the cross-talk among these components underlies ACTH-responsive CORT dynamics. To understand the time evolution of steroidogenesis following ACTH stimulation, we developed a mathematical model of

139 the SRN based on the complex molecular interactions within the SRN, including the 140 intra-adrenal GR-CORT connection that feeds back on the genomic pathway. We 141 used this model to characterise dynamic responses to a variety of perturbations, and 142 to make predictions that we then tested experimentally. Our results show that it is the 143 complex interactions between components of the SRN that governs dynamic 144 steroidogenic responses in adrenocortical cells. Importantly, the SRN architecture 145 enables cells to respond to both small and large ACTH perturbations, but is not 146 sufficient to explain the dynamic response to inflammatory stress; in this case, we 147 found that the CORT response can only be explained by coupling the SRN with the 148 immune pathway. This work provides a theoretical framework to study adrenal CORT 149 responses to a range of physiological and pathological perturbations, building upon 150 our previous work on the role of glucocorticoid-mediated negative feedback loops 151 regulating the ultradian dynamics of the HPA axis (3, 4).

152

153 **Results**

154

155 A short pulse of ACTH dynamically activates the SRN and CORT biosynthesis

156 To explore systematically how dynamic responses result from complex interactions 157 within the SRN, we developed a mathematical model of the regulatory network. To 158 aid analysis, we restrained the complexity of the model by stripping off non-essential 159 regulatory components from the network architecture. This was performed by 160 including only those nodes within the SRN that have been shown to be involved in 161 CORT-mediated feedback loops and participate in cross-talk with StAR (Fig. 1; see 162 SI Appendix, Fig. S1 for the full SRN) (20-25). Specifically, the model consists of a 163 set of delay differential equations (DDEs) that describe the dynamics of intra-adrenal 164 levels of CORT (A-CORT) and phosphorylation of GR (pGR; a marker of GR 165 activation), and the expression of DAX-1, SF-1 and StAR following "virtual" (in silico) 166 stimulation by ACTH (SI Appendix, Fig. S2 and SI Appendix, Supp. Text).

168 To assist the calibration of our mathematical model, we first considered the response 169 of the SRN to a single pulse of ACTH, similar to an endogenous ultradian pulse of 170 ACTH (Fig. 2a). To do this, we performed an in vivo experiment in the rat to 171 characterise the dynamic response of the adrenal SRN to an i.v. injection of ACTH. 172 We measured both those factors in the core SRN (Fig. 1), as well as some additional 173 factors that are also involved in steroidogenesis but were not considered in the 174 model. The i.v. ACTH injection resulted in a rapid and transient increase in plasma 175 ACTH (Fig. 2a), with a peak concentration comparable to the peak of endogenous 176 ultradian ACTH pulses (26, 27). This was accompanied by a rapid and dynamic 177 increase in CORT secretion, detectable in both adrenal tissue (A-CORT; Fig 2b) and 178 blood plasma (P-CORT; SI Appendix, Fig S3b). To assess the dynamics of GR 179 activation following the i.v. ACTH injection, we measured phosphorylation of GR 180 (pGR) at Serine 211 (28). pGR demonstrated a rapid, significant increase (Fig. 2c), 181 and its pattern of activation was similar to the observed pattern of A-CORT (Fig. 2b). 182 183 ACTH also regulates steroidogenesis over a slightly slower time frame via 184 modulation of the expression of genes involved in CORT synthesis. We therefore 185 determined the effects of the i.v. ACTH injection on the dynamic expression of key 186 steroidogenic genes in the adrenal SRN. Since many ACTH-regulated genes require

187 activation of CREB in order to be transcribed, we first assessed the dynamics of

188 CREB phosphorylation (pCREB). Consistent with previous reports (29, 30), there

189 $\,$ was a rapid and transient increase in pCREB (SI Appendix, Fig. S3d). We then

190 investigated the dynamic transcription of steroidogenic genes by measuring their

191 hnRNA and mRNA levels. As expected, the dynamics of pCREB was paralleled by a

192 transient activation of StAR hnRNA, followed by a slower StAR mRNA accumulation

193 (Fig. 2f,i). In response to the i.v. ACTH injection, there was a rapid, transient

decrease in DAX-1 hnRNA, and a dynamic increase in SF-1 hnRNA (Fig. 2d,e). In

- 195 contrast, no significant changes in DAX-1, SF-1, and StARp37 protein levels were
- 196 detected within two hours of ACTH administration (Fig. 2j-l).
- 197
- 198 With respect to those SRN factors that were not considered in the model, we found
- 199 that the i.v. ACTH injection also had a rapid effect on the activity of hormone-
- 200 sensitive lipase (HSL), a protein that plays a key role in the CORT SRN by regulating
- 201 intracellular metabolism of cholesterol (SI Appendix, Fig. S3c). We assessed this by
- 202 measuring the phosphorylation dynamics at sites known to increase HSL activity
- 203 (pHSL at Serine 660 and 563) (31), and observed a dynamic activation of both
- 204 pHSL-(Ser660) and pHSL-(Ser563). This suggests that a rapid, transient increase in
- 205 intracellular free cholesterol precedes CORT synthesis, and supports previous
- 206 findings that this is a crucial step in the steroidogenic process. Furthermore, ACTH
- 207 rapidly activated hnRNA and mRNA levels of MC2R, MRAP, Nur77 and CYP11A1,
- 208 but had no effect on HSL hnRNA and mRNA (SI Appendix, Fig. S3e). However, no
- 209 significant changes in CYP11A1, StARp32, StARp30 or HSL protein levels were
- 210 detected within two hours of ACTH administration (SI Appendix, Fig. S3f). Taken
- 211 together, these results show that dynamic changes within both non-genomic and
- 212 genomic processes are involved in the dynamic regulation of steroidogenesis.
- 213
- 214 Next, we fitted the model to the experimental data (Fig. 2). In addition to recalibrating 215 model parameters, we further considered a potential post-transcriptional regulatory 216 mechanism; namely, ACTH dose-dependent control of DAX-1 mRNA stability. This 217 proposed mechanism accounts for DAX-1 mRNA levels observed following ACTH 218 stimuli of varying magnitude (32) (SI Appendix, Supp. Text and SI Appendix, Fig. 219 S4). The time-evolution of the SRN was generated by driving the mathematical 220 model with a virtual ACTH pulse given during the nadir of the circadian rhythm (SI 221 Appendix, Fig. S2 and SI Appendix, Supp. Text). These simulations showed that the
- 222 A-CORT dynamics exhibited a rising phase (<5 min), a peak ~10 min, and a falling

- 223 phase (>15 min) (Fig. 2b). Similarly, the pGR dynamics predicted by the model
- displayed a rising phase (<5 min), a peak ~12 min, and a falling phase (>15 min)
- 225 (Fig. 2c). The model also predicted a transient induction of the SF-1 and StAR
- 226 genes, as judged from their rapidly-inducible primary transcripts (heteronuclear RNA,
- 227 hnRNA), mRNA, and protein dynamics (Fig. 2e,f,h,i,k,l). In contrast, the model
- 228 predicted a transient inhibition of DAX-1 hnRNA, but a transient induction of its
- 229 mRNA and protein levels (Fig. 2d,g,j). Taken together, these simulations closely
- 230 matched the *in vivo* data for A-CORT and pGR dynamics, as well as gene induction
- 231 (Fig. 2b-f). Whilst the changes in mRNA and protein levels measured in vivo were not
- 232 statistically significant, our model still reproduced the transient dynamics of DAX-1
- 233 mRNA and protein (Fig. 2g,j), and predicted a gradual increase of SF-1 and StAR
- 234 mRNA and protein (Fig. 2h,i,k,l) that is consistent with the transient increase of their
- 235 corresponding hnRNAs. Discrepancies between the model and experimental data
- 236 may originate from unexplored post-transcriptional or post-translational regulatory
- 237 mechanisms of SF-1 and StAR expression that the model does not account for.
- 238
- 239
- 240

241 The SRN underlies the ACTH/CORT dynamic response to large perturbations

242 comparable to those observed during inflammatory stress

243 Our data show that an ACTH stimulus comparable to an endogenous ultradian pulse

- 244 leads to dynamic changes within the SRN, ultimately driving the tightly-correlated
- 245 dynamic release of CORT. However, the adrenal gland is also subjected to more
- forceful perturbations (e.g. those associated with severe stressors), and in some
- cases the tight correlation between ACTH and CORT is lost. This "dissociation" has
- been implicated in a change of the adrenal's sensitivity to ACTH (reviewed in (6)),
- suggesting that the SRN has the ability to decode ACTH signals in a context-
- 250 dependent manner. Motivated by this, we used our model to investigate how the

SRN responds to a larger ACTH perturbation, comparable to that measured in
response to an inflammatory stress (8), and tested our model predictions
experimentally.

254

255 We have previously developed an experimental methodology to apply a large ACTH 256 stimulus to the rat adrenal glands (8). This consists of four sequential sub-cutaneous 257 (s.c.) injections of ACTH at 35 min intervals, which results in a large surge in ACTH 258 plasma concentrations comparable to that observed during inflammatory stress (Fig. 259 3a). To explore how such a substantial ACTH perturbation affects the dynamics of 260 components within the adrenal SRN, we drove our model with this same pattern of 261 ACTH stimulation and simulated the transient dynamics of CORT and pGR, as well 262 as DAX-1, SF-1, and StAR. Our model predicted a rapid, transient increase in A-263 CORT consisting of a rising phase (<30 min), plateau (45 to 120 min), and falling 264 phase (>120 min) (Fig. 3b). As expected, the pGR dynamic profile predicted by the 265 model was consistent with that of A-CORT (Fig. 3c). In terms of gene expression, the 266 model predicted a transient inhibition of DAX-1 gene expression, as well as a 267 transient induction of SF-1 and StAR genes (Fig. 3d-i). Regarding protein levels, our 268 model produced a transient activation of SF-1 and StARp37, and a gradual decline in 269 DAX-1 (Fig. 3j-I).

270

271 To test these modelling predictions, we measured the dynamic response of the rat 272 SRN to this ACTH challenge. The increase in plasma ACTH levels was paralleled by 273 a robust increase in both A-CORT (Fig. 3b) and P-CORT (SI Appendix, Fig. S5b). As 274 observed in the i.v. ACTH pulse experiment, there was a dynamic increase in pGR 275 that was consistent with the observed pattern of A-CORT (Fig. 3c). Consistent with 276 the ACTH i.v. pulse data, there was a significant decrease in DAX-1 hnRNA and 277 mRNA (Fig. 3d,g), and a significant increase in SF-1 and StAR hnRNA and mRNA 278 (Fig. 3e,f,h,i). With regard to protein levels, the high dose of ACTH also induced a

279 significant increase in StARp30 (Fig. S5f), but not in StARp37 (Fig. 3I), StARp32 (SI 280 Appendix, Fig. S5f), or SF-1 (Fig. 3k). There was a non-significant decrease in DAX-281 1 protein (Fig. 3j). The dynamic effects of a high dose of ACTH on A-CORT was 282 paralleled by an increase in both pHSL-(Ser660) and pHSL-(Ser563) (SI Appendix, 283 Fig. S5c). Furthermore, dynamic transcription of StAR was accompanied by dynamic 284 changes in pCREB (Fig. S5d), as well as a rapid increase in MRAP, Nur77 and 285 CYP11A1 hnRNA and mRNA. Unexpectedly, we observed a decrease in MC2R 286 hnRNA and mRNA, and small dynamic changes were also observed for HSL hnRNA 287 and mRNA (SI Appendix, Fig. S5e). In contrast, there was no effect on CYP11A1 288 and HSL protein (SI Appendix, Fig. S5f).

289

290 Overall, there was a good qualitative fit between our model predictions and our

291 experimental data for both small and large ACTH perturbations (Fig. 2 and Fig. 3,

respectively). Importantly, the network architecture and parameter values in our SRN

293 model were identical in both cases. This suggests that, for the majority of the

294 molecular species considered in the model, no new mechanisms controlling the SRN

295 dynamic response have to be introduced to explain the dynamic SRN responses to

296 ACTH stimuli of varying magnitude.

297

298 Dynamics of adrenal glucocorticoid synthesis and SRN activity in response to

299 LPS

Our mathematical model of the SRN suggests that the adrenal response to ACTH is primarily dependent on the specific pattern of ACTH stimulation. However, in several stress scenarios, the adrenal may not only be driven by ACTH, but may also be influenced by other factors that modulate the SRN response. For instance, activation of the immune system by administration of the bacterial endotoxin LPS is known to induce CORT secretion through a robust activation of the HPA axis, as well as through a direct effect at the level of adrenocortical steroidogenic cells (33).

- 307 Therefore, we investigated the dynamic pattern of adrenal activation in response to
- 308 such a stimulus. As previously shown (8), LPS administration induced an elevated
- 309 and prolonged ACTH response (Fig. 4a) that was followed by a long-lasting A-CORT
- 310 (Fig. 4b) and P-CORT (SI Appendix, Fig. S6b) response. Unexpectedly, despite the
- 311 high levels of A-CORT measured after the LPS injection, there was no significant
- 312 effect of LPS on pGR (Fig. 4c). An elevated and long-lasting effect of LPS was
- 313 observed on the genomic adrenal steroidogenic pathway, including StAR hnRNA and
- 314 mRNA (Fig. 4f,i), and a multiphasic effect on SF-1 hnRNA (Fig. 4e) with a small non-
- 315 significant increase in SF-1 mRNA (Fig. 4h). In contrast to the decrease in DAX-1
- 316 observed in the high ACTH dose experiment, we observed a biphasic dynamic effect
- 317 on DAX-1 hnRNA and mRNA (Fig. 4d,g). We also measured the effects of LPS on
- 318 protein levels and found that it increased the levels of StARp37 (Fig. 4I) and
- 319 StARp32 (SI Appendix, Fig. S6f), and caused a non-significant increase in StARp30
- 320 (SI Appendix, Fig. S6f). There was no effect of LPS on SF-1 protein (Fig. 4k), but we
- 321 observed a biphasic effect of LPS on DAX-1 protein (Fig. 4j) that was consistent with
- 322 DAX-1 hnRNA and mRNA.
- 323
- 324 In parallel to the effect on CORT, there was a robust and prolonged effect on pHSL-
- 325 (Ser660) and pHSL-(Ser563), as well as dynamic changes in pCREB (SI Appendix,
- Fig. S6c,d). Consistent with the findings from the high ACTH dose experiment, LPS
- 327 induced dynamic changes in MRAP, Nur77 and CYP11A1 hnRNA and mRNA, and a
- 328 significant decrease in MC2R and HSL hnRNA and mRNA (SI Appendix, Fig. S6e).
- 329 Despite its effects on transcription, there was no effect of LPS on CYP11A1 and HSL
- 330 protein (SI Appendix, Fig. S6f).
- 331
- 332 Injection of LPS in the rat is also known to increase plasma cytokines in a time-
- dependent manner (34). In addition, because LPS can also increase intra-adrenal
- and expression of cytokines, we investigated the dynamics of these in LPS-treated rats.

As expected, LPS induced a rapid and transient increase in IL-1β hnRNA and
mRNA, IL-6 hnRNA and mRNA, and TNF-α hnRNA and mRNA (SI Appendix, Fig.
S7a-c).

338

339 Whilst LPS increases both ACTH and cytokine levels, it is not clear whether these 340 cytokines modulate the SRN response to ACTH and the mechanisms through which 341 this may occur. To investigate this, we drove our model of the SRN with the ACTH 342 pattern generated by the LPS injection, and compared its output with the 343 experimental data (Fig. 4a). Our model roughly approximated the dynamics of A-344 CORT and the DAX-1 hnRNA (Fig. 4b,d). In contrast, the expression dynamics of 345 SF-1 and StAR showed some discrepancies in their hnRNA dynamics (Fig. 4e, f), 346 where the model failed to reproduce the biphasic induction of these genes. Parallel to 347 A-CORT, the model predicted induction and sustained activation of pGR. This was in 348 stark contrast with our experimental data showing a lack of pGR activation despite 349 high A-CORT (Fig. 4c). Taken together, these results suggest that a model of the 350 SRN lacking cytokine regulation is not sufficient to explain its response to LPS. 351

352 **Dynamic effects of LPS-induced adrenal cytokine expression on the SRN**

353 The lack of GR activation in response to LPS, together with the fact that LPS has

354 been shown to repress GR function in other tissues (reviewed in (35), points to an

355 interaction between cytokines and the GR. Indeed, previous findings show that the

356 LPS-induced pro-inflammatory cytokine TNF-α inhibits GR phosphorylation in Serine

357 211 in human airway cells (36). In light of this, we extended our model to account for

358 cross-talk between components of the SRN and the immune pathway, and used it to

359 test whether these interactions could account for the observed response to LPS. To

360 do this, we constructed a map of the network architecture in order to elucidate its

361 effects upon steroidogenesis (SI Appendix, Fig. S8 and SI Appendix, Supp. Text). As

362 was the case for the full steroidogenic network, we simplified the map of interactions

363	between the immune pathway and the SRN (Fig. 5 and SI Appendix, Supp. Text).
364	Further, our time course data on cytokine mRNA levels measured in the adrenal
365	exhibits a timescale similar to cytokine protein dynamics in plasma following LPS
366	stimulation (34). Thus, we accounted for the effects that cytokines TNF- α , IL-1 β , and
367	IL-6 have upon specific components of the SRN by using their mRNA time course
368	data. In other words, in addition to the high ACTH response elicited by LPS we used
369	the cytokines' dynamic profiles as additional, selective input to specific targets within
370	the SRN (SI Appendix, Fig. S9).

372 We modified the mathematical model to account for these cytokine effects (SI

Appendix, Supp. Text), and carried out computer simulations to explore the

374 steroidogenic response to LPS. The mathematical model not only predicted the

375 sustained induction of A-CORT (Fig. 6b), but also the inhibition of pGR that was

observed following LPS injection (Fib. 6c; compare with Fig. 4c). The model also

377 predicted the transient, multiphasic induction of SF-1 and StAR genes (Fig. 6e,f), as

well as the transient inhibition of DAX-1 gene expression (Fig. 6d), and their

379 associated mRNA and protein dynamics (Fig. 6g-I).

380

381 Discussion

382

383 Previous studies of adrenal steroidogenesis have focused on specific interactions 384 between genes and proteins involved in glucocorticoid synthesis. Whilst these 385 studies have provided important biological insight into the structure and regulation of 386 key steroidogenic factors, the dynamics of the interactions of these factors across 387 different time scales has not been considered. Indeed, using static data on individual 388 interactions within the steroidogenic pathway has not been sufficient to explain how 389 specific patterns of ACTH stimulation are translated by the adrenal into specific 390 patterns of CORT secretion.

392 To understand this, we explored how some of these factors interact with one another 393 in the context of a complex dynamic regulatory network. Because of the dynamic 394 nature of steroidogenesis, we developed a mathematical model of the SRN that 395 accounted for regulatory processes occurring at different time scales. Importantly, we 396 also considered an intra-adrenal CORT-mediated feedback loop whose effects were 397 previously described through mathematical modelling (17). We used our model to 398 predict the time evolution of steroidogenesis following a series of physiological 399 perturbations relevant to stress and inflammation. By testing these predictions 400 experimentally, we show how the SRN can respond to both small and large ACTH 401 perturbations, and how the coupling with the immune pathway may be necessary to 402 elicit an ACTH/CORT dissociated response during inflammation. 403 404 Mathematical models of CORT dynamics have been developed to study the origin of

405 ultradian oscillations within the HPA axis (4, 17) and the inflammatory response at 406 the adrenal level (37). However, these models do not consider the dynamic 407 molecular interactions within adrenocortical cells that occur during steroidogenesis. 408 In this study, we used a mathematical modelling approach that focuses on the 409 adrenal cell's network architecture, the interplay between fast and slow processes, 410 and post-transcriptional and post-translational control of steroidogenic genes. 411 Predictions from this model can only be of a qualitative nature, which reflects our 412 limited knowledge on the regulatory processes that we modelled phenomenologically 413 (e.g. using Michaelis-Menten and Hill type functions) and the uncertainties 414 associated with some parameter values. Another limitation relates to the fact that our 415 model did not consider all the components within the full SRN. This reflects the need 416 to simplify our computational analysis by reducing the network architecture to its core 417 components. Nonetheless, our results show that the SRN operates with the same 418 core components, network architecture, and parameter values to respond to both

small and large ACTH perturbations; in other words, no new mechanisms need to beintroduced to explain the adrenal's response to a range of ACTH stimuli.

421

422 We observed that rapid and transient CORT responses to ACTH stimulation are 423 accompanied by dynamic changes in several components within the adrenal SRN. 424 Previously, we have shown that ultradian secretion of corticosterone in the rat is 425 paralleled by pulsatile transcription of steroidogenic genes including StAR, CYP11A, 426 and MRAP, and pulsatile phosphorylation of CREB (29, 30, 38) suggesting that 427 dynamic activation of the adrenal steroidogenic pathway is important for maintaining 428 ultradian glucocorticoid secretion. In this study, we aimed to further elucidate how 429 ACTH regulates adrenocortical dynamics in response to a pulse of ACTH. To our 430 knowledge, we provide the first evidence of ultradian responsiveness at the level of 431 non-genomic mechanisms within the adrenal steroidogenic pathway (i.e. 432 phosphorylation of HSL) and within the transcriptional regulation of steroidogenic 433 gene regulators (i.e. transcription of SF-1, Nur77 and DAX-1). These findings 434 suggest that ultradian secretion of ACTH is important not only for a rapid "acute" 435 adrenal response, but also for long-term maintenance of the steroidogenic pathway 436 in basal conditions. Overall, our data suggest that dynamic events occurring within 437 the adrenal SRN are important in maintaining ultradian secretion of glucocorticoids in 438 a normal, healthy state.

439

440 While there was a good fit between the model and the hormone, hnRNA and pGR

441 data, some discrepancies were observed for mRNA and protein. The levels of mRNA

442 and protein are typically thought to depend on the transcription of the gene, but

443 additional regulatory factors such as RNA binding proteins and micro RNAs can

444 affect the stability of transcripts. Because a role for ACTH-dependent regulation of

445 mRNA stability in the adrenal cortex has been proposed (reviewed in Desroches-

446 Castan et al. 2017), we hypothesise that mechanisms such as these could explain
447 the discrepancies we have observed.

448

449 Recent data from our group suggests that, in addition to the well-known CORT-450 mediated negative feedback at the pituitary and brain regulating HPA axis activity, an 451 intra-adrenal negative feedback involving activation of the GR can rapidly inhibit 452 ACTH-induced glucocorticoid synthesis (17). Consistent with this, there is evidence 453 that glucocorticoids can inhibit the transcription of steroidogenic genes through a 454 mechanism that involves the synergistic association of SF-1 and GR to modulate 455 DAX-1 expression (20). The present study provides further evidence to support this 456 hypothesis by showing that a rapid, transient activation of adrenal GR tightly follows 457 the dynamics of intra-adrenal corticosterone. To our knowledge, this is the first 458 proposal of a model accounting for intra-adrenal feedback regulation of 459 steroidogenesis.

460

461 Further implications of an intra-adrenal CORT negative feedback loop relate to the 462 role that the DAX-1, SF-1, and StAR genes play in controlling CORT synthesis. For 463 instance, though DAX-1 expression is critical for normal HPA organ development 464 (39-41), its role in regulating steroidogenesis in the adult is still unclear. SF-1 and GR 465 have been shown to synergistically modulate DAX-1 expression (20), which in turn 466 inhibits SF-1-mediated induction of StAR (42). ACTH is known to disrupt this SF-467 1/GR synergistic effect on DAX-1 (20), and GR is activated by intra-adrenal CORT. 468 This suggests that DAX-1 may be acting as an organising "hub" by simultaneously 469 integrating and decoding hormone signals from both the pituitary and the adrenal to 470 control steroidogenesis. If this is the case, it is possible that changes in the way this 471 "hub" functions could lead to alterations in the way the adrenal SRN responds to 472 ACTH perturbations.

473

474 We also studied the steroidogenic response to an LPS injection. In this case, not only 475 the SRN but also the immune pathway was activated, as shown by the induction of 476 cytokines TNF- α , IL-6, and IL-1 β . The cross-talk between the steroid ogenic and 477 immune pathways has been modelled previously, though only at the system level 478 (37). The mathematical model we present here is the first to incorporate a detailed 479 description of the SRN and its interactions with the immune pathway. Notably, our 480 model reproduced the LPS response until we explicitly accounted for the effects that 481 cytokines have upon specific components within the SRN. This suggests that 482 cytokines interact with ACTH to mediate the inflammatory response at the level of the 483 adrenal SRN. Although it has been shown that a severe stressor can increase 484 cytokine levels in the brain and pituitary even in the absence of inflammation 485 (O'Connor K et al. 2003), and that ACTH can increase cytokine secretion in the 486 adrenal gland in vitro (Judd and MacLeod, 1992; Call GB et al. 2000), we did not see 487 such an increase in the level of adrenal cytokines following a high ACTH stimulation 488 alone.

489

490 A surprising finding of our study is the biphasic change in DAX-1 hnRNA, mRNA and

491 protein observed in the LPS experiment. While we know that IL-6 down-regulates

492 DAX-1 gene expression in bovine zona fasciculate cells, studies in different

493 steroidogenic tissues have shown that TNF- α can increase the expression of DAX-1

494 protein (Sadasivam 2015). Indeed, a rapid increase in TNF-α levels in plasma has

495 been observed shortly after LPS stimulation in the rat (Givalois, 1994), and we have

- 496 observed a rapid increase in TNF- α hnRNA and mRNA in this study. It is therefore
- 497 tempting to speculate that a transient enhancement of DAX-1 in response to
- 498 increased TNF-α may occur in the zona fasciculata of rats injected with LPS, prior to

499 the decrease induced by IL-6.

501 The sustained CORT response to transient ACTH following an LPS injection 502 suggests a disruption in the GR-mediated intra-adrenal feedback loop. This could be 503 a factor underlying the dissociated dynamics between ACTH and CORT observed in 504 stress scenarios. Uncovering the origin of this dissociated dynamics is key to 505 understanding a number of physiological and pathological conditions where 506 sustained high levels of CORT have been observed even after ACTH has returned to 507 normal (8). Consistent with this, our model predicts that cytokine effects on the DAX-508 1 regulatory "hub" of the SRN may explain the dynamic dissociation between ACTH 509 and CORT during inflammation.

- 510
- 511 With regard to other measured factors that were not included in the model, on the
- 512 whole, we observed a dynamic activity that was consistent with the ACTH and/or

513 LPS stimulation. However, we did observe some discrepancies with previous studies;

514 for example, it has been shown that ACTH induces the expression of MC2R (Winnay

515 and Hammer, 2006). However, in our study we found that, while a small pulse of

516 ACTH increased MC2R hnRNA and mRNA, administration of LPS decreased MC2R

517 hnRNA and mRNA. Similarly, while a pulse of ACTH had no effect on HSL

518 expression, LPS decreased both HSL hnRNA and mRNA. This suggests that

519 inflammatory stress can lead to a decrease in ACTH signalling and cholesterol

520 availability, which may in turn lead to adrenal hypo-responsiveness in the longer-term

- 521 that is often observed in chronic inflammation and sepsis (Boonen et al. 2015).
- 522

523 In summary, we have presented a mathematical model of steroidogenesis that

524 makes qualitative predictions in response to a range of ACTH stimuli. Further *in vitro*

- 525 experiments are required to shed light on the specific molecular mechanisms
- regulating the cross-talk between steroidogenic genes, which would enable us to
- 527 refine our models to make quantitative predictions. Nonetheless, we have shown that
- 528 our mathematical model of the adrenal SRN delivers valuable insight about the

529	transient and rapid adrenal dynamics observed in response to ACTH perturbations
530	during both basal and acute stress scenarios. In future work, the model could be
531	further exploited to study several ACTH-independent mechanisms of adrenal
532	sensitivity, such as splanchnic nerve-mediated effects on steroidogenesis (43), the
533	effect of metabolic factors such as leptin and insulin (6), and the effects on SRN
534	dynamics caused by acute and chronic exposure to synthetic glucocorticoids.
535	
536	Materials and Methods
537	
538	Animals and surgery
539	All experiments were conducted on adult male Sprague–Dawley rats (Harlan
540	Laboratories, Inc., Blackthorn, UK) weighting 220-250g at the time of arrival.
541	Rats were anaesthetised using isoflurane and an indwelling catheter was inserted in
542	the right jugular vein as previously described (44).
543	
544	Experiments and tissue collection
545	All experiments started at 9 AM and were performed 5–7 days after the surgery. Rats
546	were given: <i>i</i>) an intravenous injection of synthetic ACTH (10 ng per 0.1 ml; ACTH
547	<i>pulse experiment); ii</i>) four sub-cutaneous injections of ACTH (2 μg/kg; <i>high dose</i>
548	ACTH experiment; iii) an intravenous injection of LPS (25 µg/rat in 0.1 mL of sterile
549	saline; <i>LPS experiment</i>).
550	
551	After decapitation, trunk blood was collected and plasma processed for ACTH and
552	corticosterone measurement as previously described (Spiga et al., 2007). Adrenal
553	glands were collected and the inner zones (comprising the zona fasciculata and the
554	zona reticularis of the cortex and the adrenal medulla) were separated from the outer
555	zone (containing the zona glomerulosa and the capsula). Individual inner zones were
556	immediately frozen until processing for isolation of RNA for real-time quantitative

557	polymerase chain reaction (RTqPCR; left adrenal), and for protein extraction for
558	Western immunoblotting and corticosterone measurement (right adrenal) as
559	previously described (38, 45).
560	
561	Statistical analysis
562	Data were analyzed using the non-parametric Kruskal-Wallis Test. Statistical
563	significance was set at P < 0.05.
564	
565	Mathematical model
566	We developed a mathematical model of the adrenal SRN that accounts for the time
567	evolution of core components of the full network, namely: A-CORT, pGR, and genes
568	DAX-1, SF-1, and StAR. The model postulates a network architecture based on
569	known biological interactions, with parameter values estimated from the literature
570	and calibrated using the ACTH i.v. pulse data. The network architecture and kinetic
571	parameter values remained invariant for generating predictions about the high s.c.
571 572	parameter values remained invariant for generating predictions about the high s.c. ACTH and LPS challenge experiments.
571 572 573	parameter values remained invariant for generating predictions about the high s.c. ACTH and LPS challenge experiments.
571 572 573 574	parameter values remained invariant for generating predictions about the high s.c. ACTH and LPS challenge experiments. Because of the involvement of a slow genomic pathway and a fast non-genomic one,
571 572 573 574 575	parameter values remained invariant for generating predictions about the high s.c. ACTH and LPS challenge experiments. Because of the involvement of a slow genomic pathway and a fast non-genomic one, we wrote the model in terms of a set of DDEs, where the delays were associated to
571 572 573 574 575 576	parameter values remained invariant for generating predictions about the high s.c. ACTH and LPS challenge experiments. Because of the involvement of a slow genomic pathway and a fast non-genomic one, we wrote the model in terms of a set of DDEs, where the delays were associated to the gene transcription and translation processes. Our set of non-linear, coupled
571 572 573 574 575 576 577	parameter values remained invariant for generating predictions about the high s.c. ACTH and LPS challenge experiments. Because of the involvement of a slow genomic pathway and a fast non-genomic one, we wrote the model in terms of a set of DDEs, where the delays were associated to the gene transcription and translation processes. Our set of non-linear, coupled DDEs was numerically integrated using XPPAUT (46) and PyDDE using a fixed time
571 572 573 574 575 576 577 578	parameter values remained invariant for generating predictions about the high s.c. ACTH and LPS challenge experiments. Because of the involvement of a slow genomic pathway and a fast non-genomic one, we wrote the model in terms of a set of DDEs, where the delays were associated to the gene transcription and translation processes. Our set of non-linear, coupled DDEs was numerically integrated using XPPAUT (46) and PyDDE using a fixed time step of 0.001 min and a 4 th order Runge-Kutta integrator with adaptive steps. The
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571 572 573 574 575 576 577 578 579 580 581 581 582 583	parameter values remained invariant for generating predictions about the high s.c. ACTH and LPS challenge experiments. Because of the involvement of a slow genomic pathway and a fast non-genomic one, we wrote the model in terms of a set of DDEs, where the delays were associated to the gene transcription and translation processes. Our set of non-linear, coupled DDEs was numerically integrated using XPPAUT (46) and PyDDE using a fixed time step of 0.001 min and a 4 th order Runge-Kutta integrator with adaptive steps. The model development and complete set of model equations are detailed in the Supp. Text.

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587

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591

592 Author contributions

593

- 594 FS, JJW, JRT and SLL designed research. FS and ZZ performed research. FS, EZ
- and JJW analysed data. EZ and JJW developed and analysed mathematical models.
- 596 FS, EZ, and JJW wrote the paper with revision by all authors.

597

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Figure legends

Figure 1. The simplified SRN represented in the mathematical model accounts for both genomic and non-genomic processes occurring at different time

scales. Also considered in the model are the A-CORT/GR-mediated intra-adrenal feedback loop, and post-transcriptional and post-translational processes, including the ACTH-mediated stabilisation of StARp37.

Figure 2. The calibrated model reproduces the SRN dynamics following a short **pulse of ACTH.** Fitting the model to data successfully reproduced the rapid and transient synthesis of A-CORT (a) and GR activation (b). The model also made qualitative predictions about the induction of DAX-1, SF-1, and StAR genes (d-f) that closely matched the *in vivo* data. Similarly, the model approximated the expression of the gene products, as judged from the dynamics of mRNA (g-i) and protein levels (j-I). Red dashed lines indicate the time of ACTH injection. Representative Western immunoblotting images for pGR, StARp37, SF-1 and DAX-1 are shown in SI Appendix, Fig. S3.

Figure 3. The model reproduces the SRN dynamics following a high dose of **ACTH.** The model successfully reproduced the long and transient surges of A-CORT (a) and GR activation (b). The model also made qualitative predictions about the induction of DAX-1, SF-1, and StAR genes (d-f) that closely matched the *in vivo* data. Similarly, the model approximated the expression of the gene products, as judged from the dynamics of mRNA (g-i) and protein levels (j-I). Red dashed lines indicate the time of ACTH injections. Representative Western immunoblotting images for pGR, StARp37, SF-1 and DAX-1 are shown in Fig. SI Appendix, S5.

Fig. 4. Before considering the cross-talk with the immune pathway, the model partially reproduces the dynamic effects of an LPS-induced pulse of ACTH on components of the SRN. While ignoring cytokine effects, the model successfully reproduced the long and sustained surge of A-CORT (a), but not of GR activation

(b). The model also made qualitative predictions about the gene expression dynamics that closely matched the DAX-1 *in vivo* data (d, g, j), but failed to reproduce the SF-1 (e, h, k) and StAR (f, i, l) gene expression dynamics. Red dashed lines indicate the time of LPS injection. Representative Western immunoblotting images for pGR, StARp37, SF-1 and DAX-1 are shown in SI Appendix, Fig. S6.

Fig. 5. Cytokine effects upon targets within the adrenal SRN considered in the model. The transient cytokine pulses elicited by LPS were used as additional input functions to ACTH.

Figure 6. The extended model reproduces the SRN dynamics following administration of LPS. After accounting for cytokine effects, the model successfully reproduced the long and sustained surge of A-CORT (a) and approximated the pGR dynamics (b). The model also made qualitative predictions about the transient induction of DAX-1, SF-1, and StAR genes (d-f) that approximated the *in vivo* data. Similarly, the model approximated the expression of the gene products, as judged from the dynamics of mRNA (g-i) and protein levels (j-I). Red dashed lines indicate the time of LPS injection. Representative Western immunoblotting images for pGR, StARp37, SF-1 and DAX-1 are shown in SI Appendix, Fig. S6.













SI Appendix

Supplementary Figures



Fig. S1. The steroidogenic regulatory network (SRN). The synthesis of glucocorticoids in adrenocortical cells is governed at multiple levels by both genomic and non-genomic components. In addition to the high degree of cross-talk between these components, an intra-adrenal feedback loop mediated by A-CORT/GR may underlie adrenal responsiveness and maintain appropriate levels of glucocorticoids.



Fig. S2. Virtual (*in silico***) computer simulations of ACTH endogenous oscillations and perturbations. (a)** The model was driven by oscillating levels of ACTH that represented endogenous fluctuations. **(b)** A simple function representing a "virtual" pulse of ACTH. **(c)** The pulse was of a size comparable to the maximum endogenous levels in normal physiological conditions. **(d)** The "virtual" pulse was given near the nadir of the circadian ACTH rhythm, at approximately the same time corresponding to the *in vivo* experiments (between 8 and 9 AM).


Fig. S3. A short pulse of ACTH dynamically activates the SRN and CORT biosynthesis. (a) Representative Western immunoblotting of the effect of a pulse of ACTH on intra-adrenal phospho-GR (quantification is shown in Fig. 2c in the main text). **(b)** A pulse of ACTH dynamically affects plasma CORT (P-CORT). **(c)** Representative Western immunoblotting and quantification of the dynamic effect of a pulse of ACTH on phosphorylation of HSL (pHSL-(Ser660) and pHSL-(Ser563)). **(d)** Representative Western immunoblotting and quantification of the dynamic effect of a pulse of ACTH on phosphorylation of CREB in Ser133. **(e)** A pulse of ACTH dynamically affects plasma CORT steroidogenic genes hnRNA and mRNA. **(f)** Representative Western immunoblotting and quantification of a pulse of ACTH on steroidogenic protein expression (quantification of StARp37, SF-1 and DAX-1 proteins is shown in Fig. 2 in the main text). For Western immunoblotting data, optical density was normalised to Vinculin; for RTqPCR, data are expressed as fold induction of time 0 (n=4-7/time point). Details of asymptotic significances are reported in Table S2.



Fig. S4. Modelling ACTH dose-dependent effects on DAX-1 mRNA stability. While an increasing function g⁺ is supported by previous experiments performed at high doses of ACTH, only a decreasing function g⁻ fits our *in vivo* experiments at low doses of ACTH. We hypothesise a non-monotonous –decreasing, then increasing– response mechanism that successfully reproduces our data at both low and high doses of ACTH.



Fig. S5. Dynamic synthesis of adrenal glucocorticoids and activity of the SRN following a high dose of ACTH. (a) Representative Western immunoblotting of the effect of a high dose of ACTH on intra-adrenal phospho-GR (quantification is shown in Fig. 3c in the main text). (b) A high dose of ACTH dynamically affects plasma CORT (P-CORT). (c) Representative Western immunoblotting and quantification of the dynamic effect of a high dose of ACTH on phosphorylation of HSL (pHSL-(Ser660) and pHSL-(Ser563)). (d) Representative Western immunoblotting and quantification of the dynamic effect of a high dose of ACTH on phosphorylation of CREB in Ser133. (e) A high dose of ACTH dynamically affects plasma CORT steroidogenic genes hnRNA and mRNA. (f) Representative Western immunoblotting and quantification of StARp37, SF-1 and DAX-1 proteins is shown in Fig. 3 in the main text). For Western immunoblotting data, optical density was normalised to Vinculin; for RTqPCR, data were normalised to GAPDH mRNA levels; Western immunoblotting and RTqPCR data are expressed as fold induction of time 0 (n=4/time point). Details of asymptotic significances are reported in Table S2.



Fig. S6. Dynamic synthesis of adrenal glucocorticoids and activity of the SRN following administration of LPS. (a) Representative Western immunoblotting of the effect of LPS on intra-adrenal phospho-GR (quantification is shown in Fig. 4c in the main text). (b) Administration of LPS dynamically affects plasma CORT (P-CORT). (c) Representative Western immunoblotting and quantification of the dynamic effect of LPS on phosphorylation of HSL (pHSL-(Ser660) and pHSL-(Ser563)). (d) Representative Western immunoblotting and quantification of the dynamic effect of LPS on phosphorylation of CREB in Ser133. (e) Administration of LPS dynamically affects plasma CORT steroidogenic genes hnRNA and mRNA. (f) Representative Western immunoblotting and quantification of the dynamic effect of LPS on steroidogenic protein expression (quantification of StARp37, SF-1 and DAX-1 proteins is shown in Fig. 4 in the main text). For Western immunoblotting data, optical density was normalised to Vinculin; for RTqPCR, data were normalised to GAPDH mRNA levels; Western immunoblotting and RTqPCR data are expressed as fold induction of time 0 (n=4-12/time point). Details of asymptotic significances are reported in Table S2.



Fig. S7. Effect of LPS vs direct administration of a high dose of ACTH on intra-adrenal cytokines gene expression. The effect of LPS and high s.c. ACTH was evaluated by measuring the hnRNA and mRNA dynamics by RTqPCR (n=4-5/time point). Data are expressed as fold induction of time 0. LPS administration dynamically increased (a) IL-1 β hnRNA (P<0.0001) and mRNA (P<0.0001), (b) IL-6 hnRNA (P<0.0001) and mRNA (P<0.0001), and (c) TNF- α hnRNA (P=0.006) and mRNA (P<0.0001). In contrast, administration of a high dose of ACTH significant decreased (d) IL-1 β hnRNA (P=0.005) and mRNA (P=0.003), whereas only a small but significant increase was observed in (e) IL-6 hnRNA (P=0.03) but not on mRNA (P=0.171), and a trend toward significant increase and a significant decrease was observed in (f) TNF- α hnRNA (P=0.067) and mRNA (P=0.02), respectively.



Fig. S8. Cross-talk between the SRN and the immune pathway. During the inflammatory response elicited by LPS, the synthesis of glucocorticoids in adrenocortical cells is modulated by the immune pathway through cytokines. The SRN, in turn, also feeds back upon these cytokines, thus exhibiting cross-talk between both regulatory networks.



Fig. S9. ACTH and cytokine "virtual" pulse input functions used in the computer simulations and their associated sensitivities. (a) Small pulse (low dose) of ACTH, (b) large pulse of ACTH integrated by four cumulative doses, (c) large pulse of ACTH elicited by LPS, (d-f) cytokine pulses elicited by LPS.

Supplementary Tables

Supplementary Table 1. Primer Sequences.

Target Forward Primer		Reverse Primer		
CYP11A1 hnRNA	TGTGTGTGTGACCCCAGGAGAC	CCCAGGTCCTGCTTGAGAGGCT		
CYP11A1 mRNA	TGCGAGGGTCCTAACCCGGA	ACCTTCCAGCAGGGGCACGA		
DAX-1 hnRNA	TCCAGGCCATCAAGAGTTTC	AAGCTCACCCACTTGACCAC		
DAX-1 mRNA	TCCAGGCCATCAAGAGTTTC	GTGCTCAGTGAGGATCTGC		
GAPDH mRNA	CCATCACTGCCACCCAGAAGA	GACACATTGGGGGTAGGAACA		
HSL hnRNA	AGGTAGGAGCTGTACCCCTG	CTGCAAAGACGTTGGACAGC		
HSL mRNA	TATCCGCTCTCCGGTTGA	CGAGCACTGGAGGAGTGTTT		
IL-1β hnRNA	AGTTGTCCGTGTGTATGGGATG	GCCAGGCAGAAAGGTTTTTGTT		
IL-1β mRNA	ACCTATGTCTTGCCCGTGGA	AGGTCGTCATCATCCCACGA		
IL-6 hnRNA	CCCAGAGCACTCCACAAGG	TCTTGGTCCTTAGCCACTCCT		
IL-6 mRNA	AGCCACTGCCTTCCCTACTT	GCCATTGCACAACTCTTTTCTCA		
MC2R hnRNA	GAAGTCCGTGAGGTTGSACA	TTGTGCGGAAGGATCCAGTTT		
MC2R mRNA	GCTTTTGATCCCTGCTTTGAGTG	CATCTGTTAAAGAAGGAAAGGCTGG		
MRAP hnRNA	ACCTCATTCCTGTGGACGAG	ACCCGCCATATTATCACTGC		
MRAP mRNA	CCTCCCGGTGTGTGGCCTCT	GGGGACTATGCCTTACCTGTGGGG		
NR4A1 hnRNA	CTTGTGGGGTCCCTGCCTGC	ACGTGGAGAAGGGGCGGTCT		
NR4A1 mRNA	GCGGAACCGCTGCCAGTTCT	GCATCTGGGGGCTGCTTGGG		
SF-1 hnRNA	AGAGGGTGATGGGCTGCT	ACCTCCACCAGGCACAATAG		
SF-1 mRNA	CGCCAGGAGTTTGTCTGTCT	ACCTCCACCAGGCACAATAG		
StAR hnRNA	GCAGCAGCAACTGCAGCACTAC	GTGCCCCCGGAGACTCACCT		
StAR mRNA	CTGGCAGGCATGGCCACACA	GGCAGCCACCCCTTGAGGTC		
$TNF-\alpha$ hnRNA	CCGTGACTGTAATCGCCCTAC	CTTTAGGAGGCTGCAGAGAGAC		
TNF- α mRNA	AAATGGGCTCCCTCTATCAGTTC	TCTGCTTGGTGGTTTGCTACGAC		

Supplementary Table 2. Statistical information (P values). Related to Figs. 2–4, 6, and to Figs. S3,5,6.

0	P value				
Component	Fig. 2 and Fig. S3	Fig. 3 and Fig. S5	Figs. 4, 6 and Fig. S6		
ACTH	<0.0001	0.002	<0.0001		
A-CORT	<0.0001	<0.0001	<0.0001		
P-CORT	<0.0001	<0.0001	<0.0001		
pHSL-(Ser660)	<0.0001	0.001	<0.0001		
pHSL-(Ser563)	<0.0001	0.003	<0.0001		
pCREB	0.001	0.028	<0.0001		
StAR hnRNA	<0.0001	0.001	<0.0001		
StAR mRNA	0.067	<0.0001	<0.0001		
CYP11A1 hnRNA	<0.0001	0.002	0.007		
CYP11A1 mRNA	0.013	0.051	<0.0001		
MRAP hnRNA	<0.0001	<0.0001	<0.0001		
MRAP mRNA	<0.0001	<0.0001	<0.0001		
MC2R hnRNA	0.016	0.051	0.002		
MC2R mRNA	<0.0001	0.002	<0.0001		
HSL hnRNA	0.517	0.16	0.006		
HSL mRNA	0.221	0.021	<0.0001		
Nur77 hnRNA	<0.0001	<0.0001	<0.0001		
Nur77 mRNA	<0.0001	<0.0001	<0.0001		
SF-1 hnRNA	0.008	0.003	0.005		
SF-1 mRNA	0.361	0.002	0.091		
DAX-1 hnRNA	0.001	0.001	0.001		
DAX-1 mRNA	0.154	0.002	<0.0001		
StARp37	0.755	0.441	<0.0001		
StARp32	0.967	0.457	<0.0001		
StARp30	0.354	0.007	0.433		
CYP11A1	0.482	0.814	0.971		
HSL	0.777	0.993	0.146		
SF-1	0.385	0.492	0.76		
DAX-1	0.728	0.301	0.003		
pGR	0.032	0.003	0.528		

Supplementary Materials and Methods

Animals

All experiments were conducted on adult male Sprague–Dawley rats (Harlan Laboratories, Inc., Blackthorn, UK) weighting 220-250g at the time of arrival. Animals were given a 1-week acclimatization period prior to the start of the experiments, they were maintained under a 14 h light, 10 h dark schedule (lights on at 0500 h), and housed four per cage with ad libitum access to food and water. All animal procedures were approved by the University of Bristol Ethical Review Group and were conducted in accordance with Home Office guidelines and the United Kingdom Animals (Scientific Procedures) Act, 1986.

Surgery

Rats were anaesthetised using isoflurane and an indwelling catheter was inserted in the right jugular vein as previously described (Spiga et al., 2007). In brief, the right jugular vein was exposed, and a silastic-tipped (Merck, Whitehouse, NJ) polythene cannula (Portex, Hythe, UK) was inserted into the vessel to allow substance infusion. Cannula was prefilled with pyrogen-free heparinized (10 IU/ml) isotonic saline; the free end was exteriorized through a scalp incision and then tunnelled through a protective spring that was anchored to the parietal bones using two stainless steel screws and self-curing dental acrylic. For the high dose ACTH experiment, in addition to the intravenous cannula, during the same surgery a subcutaneous cannula was implanted under the skin between the shoulder blades. After recovery, animals were housed in individual cages in a soundproof room. The end of the protective spring was attached to a mechanical swivel that rotated through 360° in a horizontal plane and 180° through a vertical plane, allowing the rats to maximize freedom of movement. The cannula was flushed daily with heparinized saline to maintain patency.

Experiments and tissue collection

All experiments started at 9 AM and were performed 5–7 days after the surgery. *ACTH pulse experiment.* To investigate the dynamic adrenal response to an ACTH pulse rats were given an intravenous injection of synthetic ACTH (10 ng per 0.1 ml, i.v.; Synachten, Alliance Pharma, Cheltenam, United Kingdom). *High dose ACTH experiment.* To investigate the effect of a high dose of ACTH, that is able to produce similar plasma ACTH levels as observed after LPSs treatment, rats were given four injections of ACTH (2 µg/kg, sc; Synachten Depot, AlliancePharma, Cheltenam, United Kingdom) at 35-minute intervals, as previously described in (Gibbison et al., 2015). *LPS experiment.* To investigate the dynamic

adrenal response to an inflammatory stress rats were given an intravenous injection of LPS (Escherichia coli, clone 055:B5; 25 µg/rat in 0.1 mL of sterile saline; Sigma, Dorset, United Kingdom).

At the end of each experiment, rats were overdosed with 0.2 mL of sodium pentobarbitone (Euthatal, 200 mg/mL; Merial, Harlow, United Kingdom) at specific time points shown in the figures. Trunk blood was collected in ice-cold tubes containing EDTA (0.5 M; pH 7.4) and Trasylol (Aprotinin, 500,000 KIU/mL, Roche Diagnostics). Plasma was separated by centrifugation and then stored at –80°C until processed for ACTH and corticosterone measurement.

Adrenal glands were collected and the inner zones (comprising the *zona fasciculata* and the *zona reticularis* of the cortex and the adrenal medulla) were separated from the outer zone (containing the *zona glomerulosa* and the *capsula*). Individual inner zones were immediately frozen until processing for isolation of RNA for real-time quantitative polymerase chain reaction (RTqPCR; left adrenal), and for protein extraction for Western immunoblotting and corticosterone measurement (right adrenal) as previously described (Park et al., 2013; Spiga et al., 2011b).

RNA isolation and RT-qPCR

Total RNA was extracted from the inner zone of individual adrenals using TRIzol reagent (Invitrogen, Hopkinton, MA, USA), followed by purification using RNeasy mini kit reagents, and column DNase digestion (Qiagen, Valencia, CA, USA) to remove genomic DNA contamination. Complementary DNA was reverse transcribed from 1 µg of total RNA using Cloned AMV First-Strand cDNA synthesis kit (Invitrogen). Primers were designed to specifically detect primary transcript (hnRNA) or mRNA (Table S1). Fast SYBRGreen Master (Applied Biosystems, Foster City, CA, USA) was used for the amplification mixture with each primer at a final concentration of 200 nm and 2 µl of cDNA for a total reaction volume of 25 µl. PCR reactions were performed on a spectrofluorometric thermal cycler. The expression of each target gene was normalized to glyceraldehyde 3-phosphate dehydrogenase (GAPDH) mRNA as determined in a separate real-time PCR reaction. The absence of RNA detection when the reverse transcription step was omitted indicated the lack of genomic DNA contamination in the RNA samples.

Western immunoblotting

Whole cell lysate from the inner zone of individual adrenals were prepared using RIPA buffer (Sigma) supplemented with 0.2 mM Na orthovanadate, 2 mM NaF, and Complete Protease

Inhibitor (Roche Diagnostics Ltd., Burgess Hill, UK). Protein concentration was quantified by spectrophotometry using the Pierce BCA protein assays, (Thermo Fisher Scientific Inc. Rockford, IL, USA). Aliquots of each sample (10–15 µg of protein) were loaded and separated in a 10% or 4–15% Tris–Glycine gel (BioRad, Hercules, CA, USA), transferred to a PVD membrane (GE Amersham Biosciences, Piscataway, NJ, USA), blocked with 5% non-fat milk or 1% bovine serum albumin (BSA, sigma) in 1 × Tris-buffered saline plus 0.05% Tween 20 (TBST) and incubated overnight with Aliquots of each sample (10–15 µg of protein) were loaded and separated in a 10% or 4–15% Tris–Glycine gel (BioRad, Hercules, CA, USA), transferred to a PVD membrane (GE Amersham Biosciences, Piscataway, NJ, USA), blocked with 5% non-fat milk or 1% bovine serum albumin (BSA, sigma) in 1 × Tris-Glycine gel (BioRad, Hercules, CA, USA), transferred to a PVD membrane (GE Amersham Biosciences, Piscataway, NJ, USA), blocked with 5% non-fat milk or 1% bovine serum albumin (BSA, sigma) in 1 × Tris-buffered saline plus 0.05% Tween 20 (TBST) and incubated overnight with antibodies to StAR, DAX-1, CYP11A1, HSL, GR (Santa Cruz Biotechnologies, Inc., Dallas, TX, US), SF-1 (Upstate Biotechnologies Inc., Lake Placid, NY, US); CREB, phospho-CREB(Ser133), phospho-HSL(Ser660), phospho-HSL(Ser563); phospho-GR(Ser211) (Cell Signalling Technology, Danvers, MA, US).

After washing with TBST, the membranes were incubated with a horseradish peroxidaseconjugated donkey anti-rabbit IgG or donkey anti-goat IgG (Santa Cruz Biotechnologies). Immunoreactive bands were visualized using ECL Plus TM reagents (GE Amersham Biosciences) followed by exposure to BioMax MR film (Eastman Kodak; Rochester, NY, USA). After film exposure, blots were stripped and assayed for anti-goat vinculin (Santa Cruz Biotechnology). The intensity of the protein targets bands integrated with the area was quantified using a computer image analysis system, Image J (developed at the National Institutes of Health and freely available at: http://rsb.info.nih.gov). Data points for each gene were then normalized relative to the vinculin band in the respective sample.

Hormone measurement

Adrenal CORT was measured in adrenal whole cell extract prepared for Western blotting and CORT levels were normalized to the total protein content. Total plasma and adrenal CORT was measured by radioimmunoassay (RIA) using a citrate buffer (pH 3.0) to denature the binding globulin as previously described (Spiga et al., 2007). Antiserum was kindly supplied by Professor Gabor Makara (Institute of Experimental Medicine, Budapest, Hungary) and [125I] CORT was purchased from Izotop (Budapest, Hungary). ACTH in plasma was measured by RIA using a commercially available assay (MP Biomedicals, Santa Ana, California, USA) in accordance with the manufacturer's instructions.

Statistics

Sample sizes in each experiment were determined on the basis of pilot studies and previous experience with similar experimental design. Animals were allocated to each experimental group (time point after treatment) by simple randomization. Experimenters were blinded to the experimental group at the time of hormones, RNA and protein measurements.

Data are represented as the mean \pm SEM, hnRNA, mRNA and protein data are expressed as fold induction of basal (time 0). No animals or samples were excluded from statistical analysis. To check data for normality and equality of variance, we used Shapiro-Wilk and Leven tests, respectively. The overall effect of treatments was analysed using the Independent-Samples Kruskal-Wallis Test. Asymptotic significances for each experiment are shown in Table S2. Statistical significance was set at P < 0.05.

Mathematical Model

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1 Model Development

The adrenal SRN controlling the CORT biosynthetic pathway is very complex (Fig. S1). In order to develop a useful yet, manageable mathematical model, we simplified the network by reducing the number of components to a minimum (Fig. 1). Our model predictions support these gene components as the core regulatory clockwork of the SRN. A list of source references to the connections between nodes in both the full and reduced network is shown in Table A.

The mathematical model is representative of a single adrenocortical cell stimulated by an ACTH input. Specifically, ACTH stimulates adrenocortical cells for controlling the biosynthesis of CORT, which, respectively, represent the 'input' and 'output' of the SRN. Since our model is decoupled from other components of the HPA axis, we reproduced endogenous ACTH levels originated from the brain by using an ultradian oscillatory input (T = 75 min) with circadian modulated amplitude. Thus, in basal, non-stressed conditions, the model dynamics is not static but exhibits an ultradian and circadian rhythmicity driven by the ACTH input. Some model assumptions and biological considerations are the following:

- i) The model doesn't focus on the specifics of the CORT biosynthetic pathway within the mitochondria or its release into the bloodstream. Instead, we consider A-CORT as a better readout of the network since its increase occurs earlier than P-CORT, thus reflecting the very first product of CORT biosynthesis.
- ii) The network architecture emphasises the crosstalk between StAR, DAX-1 and SF-1 genes. For each of these genes, their hnRNA, mRNA and protein species are modelled explicitly. Though MC2R, MRAP, Nur77 and HSL genes are included in the full network map (Fig. S1), the network architecture suggests they're not directly involved in feedback regulation within the genomic pathway.
- iii) Likewise, even though the dynamics of the different phosphorylation states of the HSL protein (pHSL-S565, pHSL-S563 and pHSL-S660) is important to understand cholesterol synthesis, it is StARp37 the one controlling the mitochondrial import of cholesterol, which is the ratelimiting step for ACORT biosynthesis. Accordingly, we didn't include the pHSL dynamics in the present model.
- iv) To decouple the intra-adrenal system from the whole HPA axis, we assumed ACTH as an externally controlled input that targets specific components within the adrenal SRN. This took the form of either an endogenous ultradian input with circadian amplitude, a single pulse perturbation, or a combination of both.

Furthermore, we simplified the model by reducing the number of model equations. For instance, cAMP, PKA and pCREB intermediary species in the signalling cascade ending in promoter regulation are not modelled explicitly. Though pCREB is known to regulate transcription of StAR [Manna et al., 2009; Sugawara et al., 1997a,b; Zhao et al., 2005] as well as of other steroidogenic genes, the components of the cascade are known to be downstream of the ACTH pathway, which is already considered in the model.

To reproduce experimental conditions, we administered "virtual" ACTH pulses near the circadian nadir of the endogenous ACTH oscillations. Specifically, we simulated 7 pulses, one every 10 min, distributed between 8 AM and 9 AM of the day. The dynamics during the 120 min following the stimuli was recorded and averaged for the ACTH i.v. pulse experiments, and during 240 min for the high s.c. ACTH and LPS challenge experiments.

1.1 DAX-1 Gene Expression

The dynamics of gene activation, represented by the newly synthesised, non-edited hnRNA $dax1^1$, can be represented by the following equation:

$$dax1 = \sigma_{dax1} + k_{dax1} f_{dax1} (SF1, pGR) g_{dax1}^{-} (ACTH) - \gamma_{dax1} dax1 \tag{1}$$

where σ_{dax1} is the basal transcription rate, k_{dax1} is the maximum transcription rate, γ_{dax1} is the dax1 hnRNA degradation rate, and f_{dax1} is a function accounting for dax1 transcription initiation by the synergistic action of SF1 and pGR, which is modulated by ACTH. Since the specific molecular mechanisms governing the Dax1² promoter activation are still poorly understood, we take an heuristic approach and propose a phenomenological function that captures the effects observed in the study by [Gummow et al., 2006], in which SF1 and GR were cotransfected with a Dax1-Luc reporter into JEG3 cells and stimulated with varying doses of dexamethasone (thus mimicking steroid activation of the GR). Subsequently, a similar experiment was performed in Y1 cells which were concomitantly stimulated with ACTH to assess the level of induction when SF1 and GR were present separately or simultanouesly.

CORT-activated GR, which we measured via its phosphorylated form pGR, is known to synergise with SF1 to enhance Dax1 gene expression (up to 30-fold activation in a dose-dependent dexamethasone stimulation experiment). In addition, SF1 is known to also activate the promoter in a pGR-independent way (11-fold activation). Moreover, Glucocorticoid Responsive Elements (GREs) have been identified in the Dax1 promoter and they are dose-responsive to dexamethasone stimulation in the absence of SF1 [Gummow et al., 2006]. However, we will not consider SF1-independent effects on f_{dax1} as the elicited fold activation is not only very close to baseline but also small when compared against SF1-pGR synergistic activation and SF1-dependent induction. Further experiments performed by [Gummow et al., 2006] suggest that SF1 and pGR form a molecular complex at the Dax1 promoter which enhances its expression upon steroid stimulation. Considering this and the role that SF1 has in regulating the expression of the StAR gene after binding to the DAX1 protein (thus forming another molecular complex) [Sugawara et al., 1997a,b; Fan et al., 2004; Gummow et al., 2006; Manna et al., 2009], we propose that f_{dax1} takes the form:

$$f_{dax1}(SF1, pGR) = \frac{\frac{SF1}{K_{SF1}} + \frac{SF1 \cdot pGR}{K_{SF1} \cdot K_G}}{1 + \frac{SF1}{K_{SF1}} + \frac{SF1 \cdot pGR}{SF1 \cdot FGR}}$$
(2)

In Eq. 2, it can be seen that SF1 is necessary for activating the Dax1 promoter, either independently (i.e. in its free form) or forming a complex with pGR. Thus, pGR can be viewed as an enhancer which acts upon Dax1 promoter through the SF1-pGR complex. The non-linearity observed in Eq. 2 is expected from the pGR-independent SF1 binding to any of its three binding sites within the Dax1 promoter and from the synergy between the central SF1 binding site and the proximal GRE binding site that enhances expression [Gummow et al., 2006]. As the specific mechanisms of regulation are unknown, we chose the simplest empirical representation of the Dax1 promoter regulation by SF1 and pGR transcription factors.

Furthermore, [Gummow et al., 2006] also show that both the SF1 and SF1-pGR synergistic activation of the Dax1 promoter are decreased upon ACTH stimulation. We modelled this by multiplying the second term in Eq. 1 by a decreasing function $g_{dax1}^-(ACTH)$. As the specific mechanisms and molecular intermediaries of ACTH-dependent modulation are unknown, it is convenient to choose a non-linear Hill type function (see Eq. 3). The K's in Eqs. 1 and 2 account for half-maximum constants which were determined by fitting the model to data.

$$g_{dax1}^{-}(ACTH) = \frac{K_{dax1}^{ACTH^{4}}}{K_{dax1}^{ACTH^{4}} + ACTH^{4}}$$
(3)

¹The hyphen in dax-1, Dax-1, DAX-1, sf-1, Sf-1, SF-1, and A-CORT labels was dropped to avoid confusion with a minus sign.

²In what follows, the non-italicised nomenclature will be used for gene names unless otherwise specified, whereas the italicised will be reserved for state variables used in the mathematical model (e.g. dax1, Dax1, and DAX1 refer, respectively, to hnRNA, mRNA, and protein concentrations).

The mature transcript dynamics is given by

$$Dax1 = k_{Dax1} dax1_{\tau_{Dax1}} - \gamma_{Dax1} Dax1 \cdot g_{Dax1}^{\mp} (ACTH)$$

$$\tag{4}$$

where k_{Dax1} stands for the mRNA maturation rate and γ_{Dax1} is the mRNA degradation rate. RT-PCR experiments aimed at detecting hnRNA can't distinguish the transcriptional stage at which the RNA is within the nucleus. However, we know that edition processes such as splicing have finished once mature mRNAs are detected. Hence, we assumed Eq. 4 is delayed by τ_{Dax1} , which accounts for the elapsed time since transcription initiation up to the emergence of a hnRNA plus the maturation time after completion of transcription (splicing, capping, polyadenylation and mRNA nuclear export processes).

In addition, high doses of ACTH are known to destabilise Dax1 transcripts [Ragazzon et al., 2006], which in a first instance was modelled by multiplying the turnover term in Eq. 4 by a positively increasing function of ACTH $(g_{Dax1}^+(ACTH))$. However, when re-calibrating the model by fitting it to our ACTH i.v. pulse experiments, we observed a mismatch between its predictions and our DAX1 mRNA and protein data. We realised that our assumption of a monotonously-increasing function was based on experimental results by [Ragazzon et al., 2006] that were performed using very high levels of ACTH. Thus, to fit our model to data from low ACTH i.v. pulse experiments, but keeping consistency with previous observations by [Ragazzon et al., 2006] at high ACTH levels, we hypothesised an ACTH dose-dependent control mechanism of Dax1 mRNA stability that was modelled assuming its turnover rate is modulated by the function $g_{Dax1}^{\mp}(ACTH)$. Though it's not clear how ACTH actually promotes or inhibits degradation of Dax1, we propose that a nonmonotonous regulatory function would provide the best fit of the model to our experimental data in both the low and high ACTH stimuli experiments. We model this through a decreasing-thenincreasing function containing two half-maximum constants: $K_{Dax1}^{ACTH-lo}$ for inhibition at low doses of ACTH, and $K_{Dax1}^{ACTH-hi}$ for activation at high doses of ACTH (Fig. S4). This function takes the form:

$$g_{Dax1}^{\mp}(ACTH) = \frac{K_{Dax1}^{ACTH-lo}}{K_{Dax1}^{ACTH-lo} + ACTH} + \frac{ACTH}{K_{Dax1}^{ACTH-hi} + ACTH}$$
(5)

Finally, the DAX1 protein dynamics is governed by:

$$DAX1 = k_{DAX1} Dax1_{\tau_{DAX1}} - \gamma_{DAX1} DAX1 \tag{6}$$

where k_{DAX1} stands for the protein translation rate, τ_{DAX1} is the delay associated to translation, and γ_{DAX1} is the protein degradation rate.

1.2 SF-1 Gene Expression

The dynamics of gene activation, represented by the newly synthesised, non-edited sf1 hnRNA, can be represented by the following equation:

$$sf1 = \sigma_{sf1} + k_{sf1}g_{sf1}^+(ACTH) - \gamma_{sf1}sf1 \tag{7}$$

where σ_{sf1} is the basal transcription rate, k_{sf1} is the maximum transcription initiation rate, and γ_{sf1} is the sf1 degradation rate. To the best of our knowledge, the SF1 gene is not subject to transcriptional regulation by any of the proteins considered in the model. Thus, conversely to Eq. 1, here we have no function accounting for sf1 transcription modulation by any transcription factors other than the effects induced by ACTH. This is modelled by a function $g_{sf1}^+(ACTH)$, which takes the form:

$$g_{sf1}^{+}(ACTH) = \frac{ACTH^4}{K_{sf1}^{ACTH^4} + ACTH^4}$$
(8)

Though in our model we consider that SF1 gene expression is regulated by ACTH alone [Ragazzon et al., 2006], regulation by other transcription factors within the steroidogenic pathway cannot be ruled out. This poses interesting questions about the role of this gene in controlling the adrenal

response. Moreover, evidence suggests that its mRNA and protein levels remain constant after variations in cAMP concentration, which is a known mediator of the ACTH pathway [Fan et al., 2004].

The mature transcript dynamics is given by:

$$Sf1 = k_{Sf1}sf1_{\tau_{Sf1}} - \gamma_{Sf1}g_{Sf1}(Sf1)$$
(9)

where k_{Sf1} stands for the mRNA maturation rate and γ_{Sf1} is the mRNA degradation rate. By similar arguments as in Eq. 4, we assumed Eq. 9 is delayed by τ_{Sf1} . Importantly, we found out that the best fit of the model to our data was achieved when assuming a Michaelis-Menten degradation of Sf1 mRNA. Thus, the function $h_{Sf1}(Sf1)$ takes the form:

$$h_{Sf1}(Sf1) = \frac{Sf1}{K_{Sf1} + Sf1} \tag{10}$$

Finally, the SF1 protein dynamics is governed by:

$$SF1 = k_{SF1}Sf1_{\tau_{SF1}} - \gamma_{SF1}SF1 \cdot g_{SF1}^- (ACTH) \tag{11}$$

where k_{SF1} stands for the protein translation rate, τ_{SF1} is the delay associated to translation, and γ_{SF1} is the protein degradation rate. Since ACTH is suspected to stabilise SF1, probably through phosphorylation or ubiquitination of the proteasome [Æsøy et al., 2002], we assume its turnover rate is modulated by a function $g_{SF1}^-(ACTH)$, which takes the form:

$$g_{SF1}^{-}(ACTH) = \frac{K_{SF1}^{ACTH}}{K_{SF1}^{ACTH} + ACTH}$$
(12)

1.3 StAR Gene Expression

The dynamics of StAR gene activation, represented by the newly synthesised, non-edited stAR hnRNA, can be represented by the equation:

$$stAR = \sigma_{stAR} + k_{stAR} f_{stAR} (SF1, DAX1) g^+_{stAR} (ACTH) - \gamma_{stAR} stAR$$
(13)

where σ_{stAR} is the basal transcription rate, k_{stAR} is the maximum transcription initiation rate, γ_{stAR} is the stAR degradation rate, and f_{stAR} is a function accounting for stAR transcription initiation controlled positively by SF1 and negatively by DAX1. This DAX1-mediated steroidogenic inhibition follows from the finding that DAX1 inhibits SF1 transactivation upon binding it [Babu et al., 2002; Fan et al., 2004], thus preventing StAR gene activation. In addition, ACTH is known to control the StAR gene activation directly through pCREB [Sugawara et al., 1997a,b; Manna et al., 2009]. However, since we're not modelling pCREB explicitly, we employ the function $g_{stAR}^+(ACTH)$ to model the StAR gene activation by ACTH. This function takes the form:

$$g_{stAR}^{+}(ACTH) = \frac{ACTH^4}{K_{stAR}^{ACTH^4} + ACTH^4}$$
(14)

In summary, an SF1 dependent activation mechanism, similar to the one modelled in Eq. 2, has been shown to be responsible for StAR gene activation [Sugawara et al., 1997b; Manna et al., 2003; Fan et al., 2004; Xu et al., 2009]. This follows from cotransfection experiments carried out by [Fan et al., 2004] to explore the effects of SF1 (a.k.a. Ad4BP) and DAX1 concomitantly with Forskolin stimulation of the PKA pathway.

These considerations were taken into account to assume that f_{stAR} takes the form:

$$f_{stAR}(SF1, DAX1) = \frac{\frac{SF1}{K_{SF1}}}{1 + \frac{SF1}{K_{SF1}} + \frac{DAX1}{K_D}}$$
(15)

In Eq. 13, it can be seen that both ACTH and SF1 are necessary for activating stAR, whereas DAX1 modulates the promoter activity by binding to SF1 at the promoter site (the gene is inactive

if either ACTH or SF1 are absent, but not if DAX1 is). Importantly, this empirical approach facilitates modifying Eq. 15 if we wish to account for DAX1 dosage dependent effects observed in diseased states [Xu et al., 2009; Sadasivam et al., 2015]. Just as in Eq. 2, the K's in Eq. 15 account for half-maximum constants (often interpreted as sensitivities) and were determined by fitting the model to data.

The mature transcript dynamics is given by:

$$StAR = k_{StAR} stAR_{\tau_{StAR}} - \gamma_{StAR} h_{StAR} (StAR)$$
(16)

where k_{StAR} stands for the mRNA maturation rate and γ_{StAR} is the mRNA degradation rate. By similar arguments as in Eq. 4, we assumed Eq. 16 is delayed by τ_{StAR} . Importantly, we found out that the best fit of the model to data was achieved when assuming a Michaelis-Menten degradation of StAR mRNA. Thus, the function $h_{StAR}(StAR)$ takes the form:

$$h_{StAR}(StAR) = \frac{StAR}{K_{StAR} + StAR}$$
(17)

Finally, for the StAR protein dynamics we will account for the 37 kDa precursor, as it has been shown to be the one responsible for cholesterol import before being itself cleaved and imported to the mitochondria [Arakane et al., 1998; Bose et al., 2002; Manna et al., 2009]. It's dynamics is given by:

$$StARp37 = k_{StARp37}StAR_{\tau_{StARp37}} - \mu_{StARp37}StARp37g_{StARp37}^{-}(ACTH)$$
(18)

where $k_{StARp37}$ stands for the protein translation rate, $\tau_{StARp37}$ is the delay associated to translation, and $\mu_{StARp37} = \gamma_{StARp37} + \epsilon_{StARp37}$ is the StARp37 turnover rate. This value accounts for the proteasome-mediated degradation rate of the active precursor StARp37, $\gamma_{StARp37}$, and the import rate of the StARp37 cleaved byproducts into mitochondria, $\epsilon_{StARp37}$ [Arakane et al., 1997; Bose et al., 2002; Manna et al., 2009]. As it has been show that StARp37 activity is proportional to its residence time in the cytosol [Bose et al., 2002; Granot et al., 2003], and that PKA-mediated phosphorylation stabilises the precursor [Clark and Hudson, 2015], we have included the function $g_{StARp37}(ACTH)$ in the turnover term in Eq. 18, which takes the form:

$$\bar{g_{StARp37}}(ACTH) = \frac{K_{StARp37}^{ACTH}}{K_{StARp37}^{ACTH} + ACTH^4}$$
(19)

1.4 A-CORT Dynamics

As mentioned before, the CORT biosynthetic pathway within the mitochondria won't be modelled explicitly at this stage. Instead, we assume that *ACORT* dynamics is governed by:

$$ACORT = k_{ACORT} f_{ACORT} (StARp37) - \mu_{ACORT} ACORT$$
(20)

where k_{ACORT} is the maximum synthesis rate of ACORT, and the turnover term $\mu_{ACORT} = \gamma_{ACORT} + \epsilon_{ACORT}$ accounts for the ACORT degradation rate γ_{ACORT} and the export rate from the adrenal cell into the bloodstream ϵ_{ACORT} . The function f_{ACORT} expresses the dependence of ACORT synthesis on the precursor StARp37 and takes the form:

$$f_{ACORT}(StARp37) = \frac{StARp37}{K_{StARp37} + StARp37}$$
(21)

where the big K has the same meaning as in previous functions and was determined by fitting the model to data. As mentioned before, StARp37 controls the import of cholesterol into the mitochondria [Bose et al., 2002], which is the rate limiting step for corticosteroid biosynthesis. We assumed this process can be represented as a Michaelis-Menten reaction (see Eq. 21) [Arakane et al., 1997; Manna et al., 2009; Spiga et al., 2014]. Importantly, although StARp37 is downstream of ACTH (through regulation of the StAR gene), ACORT is known to respond to ACTH stimulation on a much shorter timescale than the time needed for ACTH to exert its effects on StARp37 through the cAMP/PKA/pCREB pathway. This likely happens due to PKA mediated StARp37 stabilisation in the cytosol upon ACTH stimulation [Arakane et al., 1997; Manna et al., 2009], a process that is already accounted for in the turnover term in Eq. 18.

1.5 pGR Dynamics

Lastly, instead of modelling the concentration of the glucocorticoid receptor, we model the measured levels of its phosphorylated form pGR, a marker of its activity. We propose the dynamics of the pGR is governed by:

$$pGR = k_{GR} f_{pGR} (ACORT) - \gamma_{pGR} pGR$$
(22)

where k_{pGR} stands for the pGR activation rate by ACORT and γ_{pGR} is it's deactivation rate. The function f_{pGR} takes the form:

$$f_{pGR}(ACORT) = \frac{ACORT}{K_{ACORT} + ACORT}$$
(23)

where K_{ACORT} stands for the half-maximum constant of pGR activation by ACORT.

1.6 Model Equations for the SRN

The complete set of model equations is shown below:

$$\begin{split} d\dot{x}1 &= \sigma_{dax1} + k_{dax1} f_{dax1}(SF1, pGR) g_{dax1}^{-}(ACTH) - \gamma_{dax1} dax1, \\ D\dot{a}x1 &= k_{Dax1} dax1_{\tau_{Dax1}} - \gamma_{Dax1} Dax1 g_{Dax1}^{+}(ACTH), \\ D\dot{A}X1 &= k_{DAX1} Dax1_{\tau_{DAX1}} - \gamma_{DAX1} DAX1, \\ s\dot{f}1 &= \sigma_{sf1} + k_{sf1} g_{sf1}^{+}(ACTH) - \gamma_{sf1} sf1, \\ S\dot{f}1 &= k_{Sf1} sf1_{\tau_{Sf1}} - \gamma_{Sf1} h_{Sf1}(Sf1), \\ S\dot{F}1 &= k_{SF1} Sf1_{\tau_{SF1}} - \gamma_{SF1} SF1 g_{SF1}^{-}(ACTH), \\ s\dot{t}AR &= \sigma_{stAR} + k_{stAR} f_{stAR}(SF1, DAX1) g_{stAR}^{+}(ACTH) - \gamma_{stAR} stAR, \\ S\dot{t}AR &= k_{StAR} stAR_{\tau_{StAR}} - \gamma_{StAR} h_{StAR}(StAR), \\ StARp37 &= k_{StARp37} StAR_{\tau_{StAR}} - \mu_{StARp37} StARp37 g_{stARp37}^{-}(ACTH), \\ ACORT &= k_{ACORT} f_{ACORT}(StARp37) - \mu_{ACORT} ACORT, \\ p\dot{G}R &= k_{GR} f_{pGR}(ACORT) - \gamma_{pGR} pGR. \end{split}$$

where:

$$g_{dax1}^{-}(ACTH) = \frac{K_{dax1}^{ACTH^4}}{K_{dax1}^{ACTH^4} + ACTH^4},$$

$$g_{Dax1}^{\mp}(ACTH) = \frac{K_{Dax1}^{ACTH-lo}}{K_{Dax1}^{ACTH-lo} + ACTH} + \frac{ACTH}{K_{Dax1}^{ACTH-hi} + ACTH},$$

$$ACTH^4$$

$$K^{ACTH}$$

$$g_{sf1}^{+}(ACTH) = \frac{ACTH^{4}}{K_{sf1}^{ACTH^{4}} + ACTH^{4}}, \qquad g_{SF1}^{-}(ACTH) = \frac{K_{SF1}^{BTH}}{K_{SF1}^{ACTH} + ACTH},$$

$$g_{stAR}^{+}(ACTH) = \frac{ACTH^{4}}{K_{stAR}^{ACTH^{4}} + ACTH^{4}}, \qquad g_{stARp37}^{-}(ACTH) = \frac{K_{SF1}^{BTH}}{K_{StARp37}^{ACTH} + ACTH^{4}},$$

$$h_{Sf1}(Sf1) = \frac{Sf1}{K_{Sf1} + Sf1}, \qquad h_{StAR}(StAR) = \frac{StAR}{K_{StAR} + StAR},$$

$$f_{ACORT}(StARp37) = \frac{StARp37}{K_{StARp37} + StARp37}, \qquad f_{pGR}(ACORT) = \frac{ACORT}{K_{ACORT} + ACORT},$$

$$f_{dax1}(SF1, pGR) = \frac{\frac{SF1}{K_{SF1}} + \frac{SF1 \cdot pGR}{K_{SF1} \cdot K_{G}}}{1 + \frac{SF1}{K_{SF1}} + \frac{SF1 \cdot pGR}{K_{SF1} \cdot K_{G}}}, \qquad f_{stAR}(SF1, DAX1) = \frac{\frac{SF1}{K_{SF1}} + \frac{DAX1}{K_{D}}.$$

Node	Target	Sign	Effect	References
ACTH dax1		_	Inhibits SF1-dependent activation	[Gummow et al., 2006]
Dax1		Ŧ	Modulates mRNA stability	[Ragazzon et al., 2006]
sf1		+	Promoter activation	[Ragazzon et al., 2006]
	SF1	_	Protein stabilisation	[Æsøy et al., 2002]
	stAR	+	pCREB-mediated promoter activation [Sugawara et al., 1	
				$[{\rm Sugawara~et~al.,~1997b}]$
				[Zhao et al., 2005]
				[Manna et al., 2009]
	StARp37	+	PKA-mediated protein stabilisation	$[Arakane \ et \ al., \ 1997]$
				[Manna et al., 2009]
	HSL	+	PKA-mediated activation (S563 and S660) $$	[Manna et al., 2013]
		+	AMPK-mediated activation $(S565)$	[Watt et al., 2006]
	mc2r	+	Promoter activation	[Winnay and Hammer, 2006]
	mrap	+	Promoter activation	[Gorrigan et al., 2011]
ACORT	pGR	+	Activation through phosphorylation	
pGR	dax1	+	SF1-dependent enhancement of Dax1	[Gummow et al., 2006]
DAX1	stAR	+/-	Inhibits SF1-dependent activation	[Sugawara et al., 1997a]
				[Sugawara et al., 1997b]
				[Fan et al., 2004]
			Activation at high doses	[Xu et al., 2009]
	mc2r	+	Promoter activation	[Xu et al., 2009]
	nur77	_	Promoter inhibition	[Song et al., 2004]
	NUR77	_	Protein inhibition	[Song et al., 2004]
SF1	dax1	+	Promoter activation	[Gummow et al., 2006]
	stAR/mc2r	+	Promoter activation	[Babu et al., 2002]
				[Gummow et al., 2006]
				[Winnay and Hammer, 2006]
StARp37	StARp30/Chol	+	Cholesterol import to mitochondria	[Arakane et al., 1997]
				[Manna et al., 2009]
	$\mathrm{HSL}\text{-}\mathrm{S660}/\mathrm{Chol}$	+	Cholesterol biosynthesis	[Shen et al., 2003]
HSL	S565/S563/S660	+/-	Context-dependent phosphorylation	[Kraemer and Shen, 2002]
				[Watt et al., 2006]

Note: Obvious connections (e.g. transcription, translation, ACORT biosynthesis) are not included in this table.

Table A: Supporting references for cross-talk connections within the adrenal SRN.

2 Parameter Estimation

The temporal resolution of our experiments is determined by the timescale of the regulatory processes involved, which was in the order of minutes. We set this as the timescale for our kinetic parameters, so that all parametric units were expressed in terms of molar concentrations (M) and minutes (min). Moreover, even though our model equations are considering a single cell system, we must not forget that our experimental measurements were carried out at the tissue level. This means our model predictions could be extrapolated to the total number of cells in the adrenal cortex, provided that the appropriate cellular and nuclear volumes for the different adrenal cell types are accounted for. Though we couldn't find measurements of these volumes in Sprague Dawley rats, we approximated them from a study of the ultrastructure of Wistar rat adrenal cells, which estimates the cell and nuclear volume of steroidogenic ZF cells as ~ $3600 \,\mu m^3$ and ~ $260 \,\mu m^3$, respectively [Nickerson, 1976]. Thus, to deal with Molar units in a simpler way, we used the nuclear volume $V_{nuc} = 2.6 \times 10^{-13} \, lt$ as it is in this cellular enclosure where gene regulatory reactions take place.

Also, note that the regulatory functions have either sigmoidal or Michaelian shapes and that they're normalised (including the multivariate functions). Hence, the overall flux contribution of these terms is to modulate the weight of kinetic rates (small k's), whereas their sensitivity to activation by the independent variable is represented by the half-maximum constants (big K's), i.e. the concentration at which the *i*-th factor reaches $\frac{1}{2}$.

In what follows, we estimate the kinetic rates and other parameter values. A comprehensive summary of these values is shown in Table B.

2.1 Basal Transcription Rates

The basal transcription rates, denoted by the σ 's, are the rates at which the gene promoters are activated without the influence of any of the above-mentioned transcription regulators (thought other factors may be involved), or the minimum level of transcription that these regulators may exert on the gene (see for instance the influx term in Eq. 7 when ACTH is absent). For the sake of simplicity, we assumed that basal transcription rates amounted to 5% of the maximum transcription rates. That is:

$$\sigma_{dax1} = 0.05 * k_{dax1}, \ \sigma_{sf1} = 0.05 * k_{sf1}, \ \sigma_{stAR} = 0.05 * k_{stAR}.$$

2.2 Transcription, Translation and Activation Rates

We calculated the maximum rate of gene expression, assessed by the synthesis of the nascent transcript, as:

$$k_{gene} = \frac{k_{pol}^{max}}{L_{hnRNA}N_A V_{nuc}} D_{gene} \tag{24}$$

where k_{gene} is the maximum transcription rate of the gene, measured in Molar concentration per minute; k_{pol}^{max} is the maximum elongation speed (processivity rate) of the RNA polymerase, measured in number of bases synthesised per gene copy per cell per minute and neglecting pauses in polymerase activity; L_{hnRNA} is the hnRNA transcript length, measured in number of bases per transcript (including introns); N_A is Avogadro's constant; V_{nuc} is the nuclear volume of a Zona Fasciculata (ZF) cell, measured in litres; and D_{gene} is the gene dosage, or number of gene copies within each ZF cell.

The maximum processivity rate of the RNA Pol II in mammalian tissue culture cells is estimated at 71.6 nt/sec [Darzacq et al., 2007], corresponding to a maximum elongation speed of $k_{pol}^{max} \approx 4.3 \, kb \, min^{-1}$. Regarding the gene dosage, we only have information for human Dax1 present as a duplicate, with both copies active. Thus, we will assume for all genes that $D_{gene} = 2$. The only parameter that remains undetermined in Eq. 24 is the transcript length, which depends on the gene in question and we calculate below. For the Dax1 gene, which in rats is known to encode a hnRNA of 4129 b in length ³, we estimate a maximum transcription rate of $k_{dax1} = 1.33 \times 10^{-11} M \min^{-1}$ (after splicing, the mature mRNA is reduced to 1.8 kb in length ⁴). Similarly, the SF1 gene in rats is known to encode a hnRNA of 20825 b in length ⁵, which we estimate will have a maximum transcription rate of $k_{sf1} = 2.64 \times 10^{-12} M \min^{-1}$ (after splicing, the mature mRNA is barely 2182 b in length ⁶). Lastly, the StAR gene is known to encode a hnRNA of 4643 b in length ⁷, for which we estimate a maximum transcription rate of $k_{stAR} = 1.18 \times 10^{-11} M \min^{-1}$ (after splicing, the predominant transcript variant is a 3.5 kb mRNA [Ariyoshi et al., 1998]). Thus:

$$k_{dax1} = 1.33 \times 10^{-11} M \min^{-1}, \quad k_{sf1} = 2.64 \times 10^{-12} M \min^{-1}, \quad k_{stAR} = 1.18 \times 10^{-11} M \min^{-1}.$$

which then implies the basal transcription rates are:

$$\sigma_{dax1} = 6.65 \times 10^{-13} \, M \, min^{-1}, \ \ \sigma_{sf1} = 1.32 \times 10^{-13} \, M \, min^{-1}, \ \ \sigma_{stAR} = 5.91 \times 10^{-13} \, M \, min^{-1}.$$

The maturation rates of hnRNA into mRNA are likely subject to high variability given the postprocessing of transcripts depends on multiple, complex reaction steps (e.g. capping, polyadenylation, splicing, nuclear export) [Keene, 2007]. For the sake of simplicity, we'll assume their values are half the above maximum rates of gene expression. Thus, after halving their values and removing Molar units by multiplying by the product of Avogadro's constant and the nuclear volume, we arrive to:

$$k_{Dax1} = 1.04 \min^{-1}, \ k_{Sf1} = 0.21 \min^{-1}, \ k_{StAR} = 0.93 \min^{-1}.$$

We calculate the translation rate (protein synthesis) for any given transcript in a single ZF cell as:

$$k_{protein} = \frac{k_{rib}}{L_{poly}} \tag{25}$$

where $k_{protein}$ is the translation rate of the protein per minute per cell (keeping in mind this rate should be multiplied by the Molar concentration of mRNAs, so that protein concentration changes in units of $M \min^{-1}$); k_{rib} is the polypeptide elongation speed (processivity rate) of the ribosome, measured in number of codons translated (or aa's synthesised) per transcript per cell per minute and assuming ribosome stalling is already accounted for; and L_{poly} is the polypeptide length, measured in number of aa's (we account for the polypeptide length instead of the mRNA length as UTR regions do not contribute to protein synthesis). We will use the average ribosome progression rate for translation in *Mus musculus* as a proxy for that of rat, which was unavailable. This is estimated at 5.6 ± 0.5 codons per second, which equals to $k_{rib} \approx 336 \ aa \min^{-1}$ [Ingolia et al., 2011].

Considering the rat DAX1 polypeptide is 472 *aa* in length ⁸, we estimate a translation rate of $k_{DAX1} = 0.71 min^{-1}$. Similarly, the SF1 polypeptide in rats is 462 *aa* in length ⁹, which corresponds to a translation rate of $k_{SF1} = 0.73 min^{-1}$. Lastly, for the StAR protein precursor (the 37 kDa form StARp37), we can assume as a "rule of thumb" a molecular weight of 100-110 Da per average amino acid [Phillips et al., 2009]; assuming the newly synthesised StARp37 is ~ 336 *aa* in length (the 32 kDa form is 284 *aa* long ¹⁰), we can estimate a translation rate of $k_{StARp37} = 1 min^{-1}$. Thus:

$$k_{DAX1} = 0.71 \, min^{-1}, \ k_{SF1} = 0.73 \, min^{-1}, \ k_{StARp37} = 1 \, min^{-1}$$

³http://www.ensembl.org/Rattus_norvegicus/Transcript/Exons?db=core;g=ENSRN0G0000003765;r=X: 54734385-54738513;t=ENSRN0T00000005023

⁴http://www.ncbi.nlm.nih.gov/nuccore/NM_053317.1

⁵http://www.ensembl.org/Rattus_norvegicus/Transcript/Exons?db=core;g=ENSRN0G00000012682;r=3: 22999616-23020441;t=ENSRN0T00000017651

⁶http://www.ncbi.nlm.nih.gov/nuccore/NM_001191099.1

⁷http://www.ensembl.org/Rattus_norvegicus/Transcript/Exons?db=core;g=ENSRNOG00000015052;r=16:

^{71036204-71040847;}t=ENSRN0T00000020606

⁸http://www.uniprot.org/uniprot/P70503

⁹http://www.uniprot.org/uniprot/P50569

¹⁰http://www.uniprot.org/uniprot/P97826

Lastly, we chose the parameters k_{ACORT} and k_{GR} arbitrarily and allow them to vary during our model fitting to data as a means to probe the negative feedback strength. For these values, we assumed

$$k_{ACORT} = 6.39 \times 10^{-12} \, M \, min^{-1}, \ k_{GR} = 1 \times 10^{-12} \, M \, min^{-1}$$

where the chosen value for k_{ACORT} is just 1 divided by the product of the nuclear volume and Avogadro's constant.

2.3 Delays

From the HPA model proposed in [Walker et al., 2010], the overall delay for ACTH signal transduction within the adrenal has been estimated to be $\tau_{Adrenal} \in [6.5, 20] min$. However, recent detailed experiments suggest two different time scales for the intra-adrenal network responsiveness, one for the slow genomic pathway and another for the fast, non-genomic one. Specifically, an *intra venous* (i.v.) pulse of ACTH produced rapid, transient increases in plasma CORT, with maximal responses achieved after 5 to 15 min (though some *in vitro* studies show ACTH can trigger CORT synthesis within ~ $3 \min$), and a decrease to almost basal levels at ~ $30 \min$. In contrast, StAR and P450scc hnRNA levels increased at 15 min following ACTH and decreased towards basal values after 30 min [Spiga et al., 2011a]. Hence, we can assume $\tau_{Adrenal}^{Non-genomic} \leq 5 \min$ and $\tau_{Adrenal}^{Genomic} \approx 15 \min$.

The case of the StAR gene is of particular interest since it is known to control the limiting step in CORT biosynthesis, which occurs very rapidly during the adrenal response to ACTH stimulation. From [Miller, 2013], we know that on a $15 - 60 \min$ time scale, ACTH rapidly stimulates both the activation of pre-existing StAR protein and its *de novo* the synthesis. ACTH/cAMP doubles the activity of StAR proteins almost immediately [Arakane et al., 1997] and induces transcription of the StAR gene within minutes [Manna et al., 2009]. *StARp37* then interacts with a complex macromolecular machine in the Outer Mitochondrial Membrane (OMM) that increases the flow of cholesterol from the OMM to the Inner Mitochondrial Membrane (IMM), where it becomes the substrate for CORT biosynthetic enzymes. Thus, we can assume that CORT synthesis is triggered almost instantaneously, whereas StAR gene expression kicks in until a few minutes later.

We can estimate the delays in the genomic pathway by taking into account the gene and hnRNA lengths, together with the processivity rate of the RNA Pol II and the ribosome. The StAR gene, for instance, is known to encode a hnRNA which is $4643 b \log b$. As mentioned before, the maximum processivity rate of the RNA Pol II in mammalian tissue culture cells is estimated at $71.6 \, nt/sec =$ $4.3 \, kb/min$ [Darzacq et al., 2007], which means the StAR hnRNA would take ~ $1.08 \, min$ to be transcribed. This transcriptional delay would be unusually short, and we should note that additional processes occur during transcript elongation, such as cumulative pauses of $\sim 4 \min$ in average for polymerases on genes. After accounting for stalling effects on a variety of genes, the same study reports an average processivity rate for RNA Pol II as small as 6.3 nt/sec. However, we must keep in mind that these measurements are subject to high variability, depending on the gene, the physiological conditions, and the cell type. For instance, in rat kidney cells the reactivation of serum responsive genes following serum deprivation suggests a synthetic rate of $1.1-1.4 \, kb/min$ [Femino et al., 1998], whereas a more recent study performed on mouse ES cells determined that the RNA Pol II elongation rates ranged between 0.5 to 4 kb/min [Jonkers et al., 2014]. In summary, several studies where measurements were taken by different techniques and in different conditions place the RNA Pol II elongation speed between 18 and 72 nt/sec [Swinburne and Silver, 2008], with most studies reporting values in the lower half of this range. Here, we will assume an average value of $30 \, nt/sec = 1.8 \, kb/min.$

Using this average processivity rate, the stAR transcriptional delay would be of 2.58 min. Regarding transcript editing, we find the duration of an average splicing event has been estimated to last ~ 30 sec [Hnilicová and Staněk, 2011]. Considering the NCBI reports the StAR gene in Rattus norvegicus has 6 introns ¹¹, we can estimate it takes ~ 3 min for stAR hnRNA to mature into StARmRNA, assuming introns are removed sequentially instead of simultaneously and neglecting transcript export to the translation site. Regarding translation, a ribosome progression rate of 5.6 ± 0.5

¹¹http://www.ncbi.nlm.nih.gov/gene/25557

codons per second in Mus musculus has been reported [Ingolia et al., 2011]. Considering a newly synthesised StARp37 precursor protein would have $\sim 336 aa$ in length (the 32 kDa form is 284 aa long ¹²), we can estimate a translation delay of $\sim 1 \min$ assuming a "rule of thumb" of 100-110 Da per average amino acid [Phillips et al., 2009]. Following this, we estimated the delays for StAR gene expression as

$$\tau_{StAR} = 5.58 \, min, \ \tau_{StARp37} = 1 \, min.$$

Likewise, the SF1 gene in rats is known to encode a hnRNA which is 20825 b long. Using the same average RNA Pol II processivity rate as before, we estimate it would take $\sim 11.57 min$ to be transcribed. As the transcript has 6 introns in total 13 , we can estimate ~ $3 \min$ for hnRNA to mature into mRNA, assuming introns are spliced sequentially and not simultaneously. Regarding translation, and considering the rat SF1 polypeptide is 462 aa in length ¹⁴, we can estimate a translation delay of $\sim 1.38 \, min$. Following this, we estimated the delays for the SF1 gene expression as

$$\tau_{Sf1} = 14.57 \, min, \ \tau_{SF1} = 1.38 \, min.$$

Lastly, the Dax1 gene in rats is known to encode a hnRNA which is 4129b long. Using the same average RNA Pol II processivity rate as before, we estimate it would take $\sim 2.29 min$ to be transcribed. As the transcript has only 1 intron in total 15 , we can estimate ~ 0.5 min for dax1 hnRNA to mature into Dax1 mRNA. Regarding translation, and considering the rat DAX1polypeptide is 472 aa in length ¹⁶, we can estimate a translation delay of ~ 1.4 min. Following this, we estimated the delays for Dax1 gene expression as

$$\tau_{Dax1} = 2.79 \, min, \ \tau_{DAX1} = 1.4 \, min.$$

$\mathbf{2.4}$ **Degradation Rates**

For simplicity, we'll assume that immature, uncapped, non-polyadenylated hnRNAs will have a degradation rate an order of magnitude faster than their mature mRNA counterparts. This results in half-lives for hnRNAs of $\sim 20 min$, comparable to those estimated previously for primary transcripts in a model of RNA metabolism in mammalian cells [Jackson et al., 2000]. However, after comparing the model predictions with our data, we required to slightly adjust some of our estimations. This likely arises from previously unaccounted context-dependent degradation processes, as we will see in what follows.

Dax1 mRNA half-life is reported by [Ragazzon et al., 2006] to be 3 hrs. This value corresponds to a degradation rate of $\gamma_{Dax1} = 3.85 \times 10^{-3} min^{-1}$. After fitting the model to data regarding the hnRNA half-life and considering a protein half-life of 6 hrs [Osman et al., 2002], we arrive to the following estimates for the Dax1 gene:

 $\gamma_{dax1} = 1.5 \times 10^{-2} \min^{-1}, \ \gamma_{Dax1} = 3.85 \times 10^{-3} \min^{-1}, \ \gamma_{DAX1} = 1.93 \times 10^{-3} \min^{-1}.$

To the best of our knowledge, there is no data available regarding Sf1 mRNA stability. Nonetheless, we can assume that, because its mRNA is ~ 400 bases longer than the Dax1 transcript, it degrades at a slower rate. Thus, after calculating a longer half-life in proportion to its length as compared to the Dax1 transcript, we estimated a degradation rate for the Sf1 mRNA of $3.18 \times 10^{-3} \min^{-1}$. This rate, however, had to be expressed in different units after the model fitting to data suggested that this degradation was better represented through a Michaelis-Menten mechanism. Thus, the Sf1 degradation rate was set to $\gamma_{Sf1} = 2.03 \times 10^{-14} M \min^{-1}$. The sf1 hnRNA degradation rate, in contrast, was assumed an order of magnitude larger than its mRNA counterpart. In the proper units, this was set as $\gamma_{sf1} = 3.18 \times 10^{-2} \min^{-1}$. Regarding the SF1 protein, previously [Jacob et al., 2001] and [Chen et al., 2007] have estimated its half-life between 2.5 - 4 hrs, corresponding to

¹²http://www.uniprot.org/uniprot/P97826

¹³http://www.ncbi.nlm.nih.gov/gene/83826

¹⁴http://www.uniprot.org/uniprot/P50569 ¹⁵http://www.ncbi.nlm.nih.gov/gene/58850

¹⁶http://www.uniprot.org/uniprot/P70503

a degradation rate within the range $2.89 \times 10^{-3} min^{-1}$ to $4.62 \times 10^{-3} min^{-1}$. However, we must take into account the PKA-mediated stabilisation effects of ACTH upon SF1. For instance, [Æsøy et al., 2002] found a significant reduction in SF1 protein amount in transfected COS-1 cells in the absence of PKA-C α overexpression measured 4hrs post chase (mean $\pm SEM = 23.7 \pm 4.64\%$, P = 0.0022), compared with the amount of SF1 protein after 1hr, which was set to 100%. In contrast, no significant reduction in SF1 protein amount was observed after 4hrs in cells cotransfected with PKA-C α (mean $\pm SEM = 77.2 \pm 9.5\%$, P = 0.143). Thus, it seems that the turnover of transiently expressed SF1 was decreased by coexpression of PKA-C α . Solving the differential equation for first order degradation, we find the degradation rate is given by:

$$\gamma = \frac{1}{t} \ln \frac{x_0}{x(t)} \tag{26}$$

which we can use to calculate $\gamma_{SF1}^{PKA-} = 8 \times 10^{-3} \min^{-1}$ and $\gamma_{SF1}^{PKA+} = 1.44 \times 10^{-3} \min^{-1}$. Here, we will use the first value, as by including the function $g_{SF1}^-(ACTH)$ in the turnover term in Eq. 11 we account for the PKA-mediated stabilising effects of ACTH stimuli. Thus, for the SF1 gene, in the absence of ACTH stimulation, we will have:

$$\gamma_{sf1} = 3.18 \times 10^{-2} \min^{-1}, \ \gamma_{Sf1} = 2.03 \times 10^{-14} M \min^{-1}, \ \gamma_{SF1} = 8 \times 10^{-3} \min^{-1},$$

The available information regarding StAR mRNA stability is a bit controversial, especially since two isoforms have been reported, each with different half-lives and the possibility of selective degradation in steroidogenic cells [Duan and Jefcoate, 2007]. As before, we could make an estimation based on its transcript length being twice as long as the Dax1 mRNA. Following this, we estimated a degradation rate of $1.98 \times 10^{-3} min^{-1}$. However, at the moment of fitting the model to data from the ACTH i.v. pulse experiment, we found that this degradation rate had to be approximately twice as large. Accordingly, we have modified it to $\gamma_{StAR} = 3.96 \times 10^{-3} min^{-1}$. Likewise, the value of the stAR hnRNA degradation was adjusted after fitting the model to data and was fixed at $\gamma_{stAR} = 1 \times 10^{-1} min^{-1}$. The increase in the γ_{StAR} degradation rate, when compared to the value we originally estimated for StAR mRNA is also supported by experimental evidence showing this gene contains three conserved AU-rich (AURE) element motifs known to mediate fast mRNA turnover [Zhao et al., 2005; Duan and Jefcoate, 2007]. Moreover, the measured half-life for the 3.5 kb StAR mRNA isoform corresponds to a degradation rate of $3.3 \times 10^{-3} min^{-1}$, which is close to our estimated value.

The estimation of the StARp37 protein turnover rate requires careful examination. According to [Arakane et al., 1997], the StAR protein has a short half-life, but a specific value is not reported. The half-life reported by [Ragazzon et al., 2006] is ~ 5 hrs, with the primary source being [Granot et al., 2003]. However, this value refers to mitochondrial 30 kDa StAR, which lacks the N-terminus mitochondrial targeting sequence and it's not involved in cholesterol import into mitochondria. This sequence is cleaved upon the StAR 37 kDa import into mitochondria, a process that influences the rate of the StARp37 precursor cytosolic proteasome degradation, according to [Granot et al., 2003].

Both [Arakane et al., 1998] and [Granot et al., 2003] estimate that the StARp37 precursor has a half-life of ~ 15 min. However, this estimate accounts only for proteasome-assisted degradation, and the authors suggest that a fast import of the precursor into mitochondria would decrease the time it remains active. This would effectively increase the precursor's lability when both degradation and translocation processes are accounted for, thus decreasing the half-life to $\leq 5 min$ [Manna et al., 2009; Clark and Hudson, 2015]. This is consistent with previous estimations of StARp37 exhibiting a half-life of 5 min in mouse Y1 cells [Artemenko et al., 2001], and of 3 - 4 min in rat adrenal cells incubated at 37 °C [Epstein and Orme-Johnson, 1991].

How then is it possible for adrenal cells to exhibit such a fast steroidogenic response while depending upon a precursor protein that is so labile? As it turns out, PKA-mediated ACTH stimulation induces rapid phosphorylation of the precursor StARp37, thus stabilising it in the cytosol and enhancing its activity. In other words, ACTH stimulation increases the residence time of StARp37 in the cytosol, thus increasing the rate of cholesterol import into mitochondria and enabling the fast response of the adrenal to synthesise CORT.

In summary, we have that when mitochondrial import is blocked, the half-life of the StAR precursor is 15 min, corresponding to $\gamma_{StARp37} = 4.62 \times 10^{-2} min^{-1}$; but when both mechanisms are blocked the precursor accumulates in the cytosol and its half-life increases up to 6.5 hrs, corresponding to a much lower turnover rate of $\mu_{StARp37} = \gamma_{StARp37} + \epsilon_{StARp37} = 1.78 \times 10^{-3} min^{-1}$ [Granot et al., 2003]. In contrast, when both protein degradation and mitochondrial import are fully functional, and in the absence of ACTH stimulation, the protein is extremely labile, with a half-life of $\sim 3.5 \, min$, corresponding to $\mu_{StARp37}^{ACTH-} = \gamma_{StARp37} + \epsilon_{StARp37} = 1.98 \times 10^{-1} \, min^{-1}$ [Epstein and Orme-Johnson, 1991]. However, upon PKA-mediated phosphorylation of the precursor after ACTH stimulation, the StARp37 stabilises and its effective turnover rate drops by two orders of magnitude down to $\mu_{StARp37}^{ACTH+} = \gamma_{StARp37}^{-1} + \epsilon_{StARp37} = 1.49 \times 10^{-3} min^{-1}$ [Clark and Hudson, 2015]. Thus

$$\begin{aligned} \mu_{StARp37}^{ACTH-} &= 1.98 \times 10^{-1} \, min^{-1} \\ \mu_{StARp37}^{ACTH+} &= 1.49 \times 10^{-3} \, min^{-1} \end{aligned}$$

The transition from a very high to a very low turnover rate for StARp37 is represented in the second term in Eq. 18 by introducing the function $g_{StARp37}^{-}(ACTH)$ (Eq. 19), which modulates the removal rate $\mu_{StARp37}$. As the decrease in the turnover rate upon ACTH stimulation is quite large, we can fix the parameter as $\mu_{StARp37} = 1.98 \times 10^{-1} min^{-1}$ when no ACTH stimulation is present and let the function $g_{StARp37}^{-}(ACTH)$ stabilise StARp37. Summarising for the StAR gene, the turnover rates are:

$$\gamma_{stAR} = 1 \times 10^{-1} \min^{-1}, \ \gamma_{StAR} = 3.96 \times 10^{-3} \min^{-1}, \ \gamma_{StARp37} = 1.98 \times 10^{-1} \min^{-1},$$

Though the half-life of intra-adrenal corticosterone (ACORT) has not been measured directly. given its rapid export to the bloodstream we can assume it has a faster removal rate than its plasma counterpart. Assuming a half-life of $1 \min$, we estimate a turnover rate of

$$\mu_{ACORT} = 6.93 \times 10^{-1} \, min^{-1}$$

Lastly, for the phosphorylated glucocorticoid receptor (pGR), we find from [Bodwell et al., 1998] that the mouse GR is very stable, with a half-life of $\sim 18 hrs$ (reduced down to 8-9 hrs upon dexamethasone stimulation). However, we're interested in the half-life of the active, phosphorylated form of GR. This is likely to be much more short-lived, but nonetheless has not been measured. For the sake of simplicity, we assume it's half-life is the same as that of ACORT, which implies a turnover rate of

$$\gamma_{pGR} = 6.93 \times 10^{-1} \, min^{-1}$$

2.5Half-Maximum Constants

The half-maximum constants (big K's) are arbitrary parameters that nonetheless capture the sensitivity of a process (e.g. synthesis or degradation) affecting the dynamics of a molecular species Y as a function of the concentration of another species X. As the molecular mechanisms underlying these processes are often unknown, we model them as Michaelis-Menten and Hill type functions. In the latter case, when the Hill coefficient is high enough, these K's could also be interpreted as activation thresholds. Though arbitrary, their relative values are informative of how sensitive different nodes within the adrenal SRN are to common external *stimuli* (e.g. ACTH and cytokines).

Half-maximum constants were estimated manually after observing the system's time evolution and correcting our model for the timescales at which the peak of the transient response, together with the rising and decreasing phases, followed the experimental data. The estimated K's are listed in Table B and represented graphically in Fig. S9. The half-maximum constants that are neither dependent on ACTH nor cytokines, but are rather involved in Dax1, Sf1 and StAR mRNA degradation, in ACORT and pGR synthesis, and in regulation of Dax1 and StAR gene promoters, were also fixed manually to values close to the range observed during their circadian dynamics. Of particular interest are the constants $K_{Dax1}^{ACTH-lo}$ and $K_{Dax1}^{ACTH-hi}$ which, respectively, represent

the lower and upper near half-maximum constants for a non-monotonous regulatory function (Eq. 5)

that models dose-dependent effects of ACTH upon $Dax1$ mRNA stability (Fig. S4). We hypothesise
that this dose-dependent degradation of $Dax1$ mRNA could be a post-transcriptional regulatory
mechanism such as stress-induced non-sense mediated mRNA decay [He and Jacobson, 2015].

Basal Transcription Rates						
$\sigma_{dax1} = 6.65 \times 10^{-13} M min^{-1}$	$\sigma_{sf1} = 1.32 \times 10^{-13} M min^{-1}$	$\sigma_{stAR} = 5.91 \times 10^{-13} M min^{-1}$				
Transcription, Translation and Activation Rates						
$k_{dax1} = 1.33 \times 10^{-11} M min^{-1}$	$k_{sf1} = 2.64 \times 10^{-12} M min^{-1}$	$k_{stAR} = 1.18 \times 10^{-11} M min^{-1}$				
$k_{Dax1} = 1.04 min^{-1}$	$k_{Sf1} = 0.21 min^{-1}$	$k_{StAR} = 0.93 min^{-1}$				
$k_{DAX1} = 0.71 min^{-1}$	$k_{SF1} = 0.73 min^{-1}$	$k_{StARp37} = 1 min^{-1}$				
$k_{ACORT} = 6.39 \times 10^{-12} M min^{-1}$	$k_{GR} = 1 \times 10^{-12} M min^{-1}$					
	Degradation Rates					
$\gamma_{dax1} = 1.5 \times 10^{-2} min^{-1}$	$\gamma_{sf1} = 3.18 \times 10^{-2} min^{-1}$	$\gamma_{stAR} = 1 \times 10^{-1} min^{-1}$				
$\gamma_{Dax1} = 3.85 \times 10^{-3} min^{-1}$	$\gamma_{Sf1} = 2.03 \times 10^{-14} M min^{-1}$	$\gamma_{StAR} = 3.96 \times 10^{-3} min^{-1}$				
$\gamma_{DAX1} = 1.93 \times 10^{-3} min^{-1}$	$\gamma_{SF1} = 8 \times 10^{-3} min^{-1}$	$\mu_{StARp37} = 1.98 \times 10^{-1} min^{-1}$				
$\mu_{ACORT} = 6.93 \times 10^{-1} min^{-1}$	$\gamma_{pGR} = 6.93 \times 10^{-1} min^{-1}$					
	Delays					
$\tau_{Dax1} = 2.79 \min$	$\tau_{Sf1} = 14.57 min$	$ au_{StAR} = 5.58 \min$				
$ au_{DAX1} = 1.4 \min$	$ au_{SF1} = 1.38 min$	$ au_{StARp37} = 1 \min$				
Half-Maximum Constants						
$K^{ACTH} = 00 \ mg \ ml^{-1}$		$K_{dax1}^{IL6} = K_{Dax1}^{IL6} = K_{DAX1}^{IL6} = 500 AU$				
$K_{dax1} = 50 pg mi$ $V^{ACTH-lo} = 20 ng ml^{-1}$	$K_{ACORT} = 4.5 AU$	$K_{sf1}^{IL6} = 1000 AU$				
$K_{Dax1} = 50 pg m t$ $K^{ACTH-hi} = 1000 m s m l^{-1}$	$K_G = 0.70 AU$	$K_{Sf1}^{IL6} = K_{SF1}^{IL6} = 100 AU$				
$K_{Dax1} = 1000 pg mt$ $K^{ACTH} = 20 mm m^{-1}$	$K_D = 5.7 \times 10^{-8} M$	$K^{IL6}_{StAR} = 2500 AU$				
$\kappa_{sf1} = 30 pg m i$ $\kappa^{ACTH} = 70 n_{S} m l^{-1}$	$\kappa_{Sf1} = 1 \times 10 M$ $K_{T-} = 6.2 \times 10^{-7} M$	$K_{StARp37}^{IL6} = 500 AU$				
$K_{SF1}^{ACTH} = 10 pg ml^{-1}$	$K_{SF1} = 0.2 \times 10^{-8} M$	$K_{stAR}^{IL1\beta} = 200 AU$				
$K_{stAR}^{ACTH} = 70 \ pg \ ml^{-1}$	$\kappa_{StAR} = 1 \times 10 M$ $V = -4 \times 10^{-8} M$	$K_{stAR}^{TNF\alpha} = K_{StAR}^{TNF\alpha} = 60 AU$				
$K_{StARp37} = 70 pg ml^{-1}$	$M_{StARp37} - 4 \times 10$ M	$K_{nGR}^{TNF\alpha} = 20 AU$				

Table B: Kinetic rates and other estimated parameter values.

3 Crosstalk Between the Adrenal SRN and the Immune Pathway

Stimulation of rats with the bacterial lipopolysaccharide (LPS) endotoxin elicits a strong response not only at the HPA level but also on the immune system. In particular, LPS induces cytokines $TNF\alpha$, $IL1\beta$ and $IL6^{17}$, which are known to act upon targets of the adrenal SRN and among themselves.

A previous mathematical model by [Malek et al., 2015] suggested a way in which LPS, $TNF\alpha$ and IL6 may interact with ACTH and CORT to mediate the adrenal response to inflammation. However, such a model assumed direct interactions of these cytokines upon ACTH, ignoring the steroidogenic pathway within adrenal cells. LPS effects on the immune system are more complex and start with the recruitment of macrophages to the adrenal cortex. This increases levels of $TNF\alpha$ in a first stage (peak at 60 min), and of $IL1\beta$ and IL6 (in T-lymphocytes) in a second stage (peaks at 120 min) [Givalois et al., 1994]. While it is believed that the first stage is amplified through a positive feedback, in the second one $IL1\beta$ and IL6 are known to inhibit rising levels of $TNF\alpha$. Moreover, though CORT is known to inhibit both $IL1\beta$ and IL6, it has been proposed that $IL1\beta$ is involved in the initial activation of CORT production, whereas IL6 sustains it [Givalois et al., 1994; Kanczkowski et al., 2013]. These interactions between LPS and cytokines, together with the crosstalk between the immune pathway and the adrenal SRN (Table C), is summarised in Fig. S8.

Even though the interactions among cytokines are complex, we know that the output of the internal crosstalk within the immune pathway upon an LPS challenge delivers the measured levels of $TNF\alpha$, $IL1\beta$ and IL6 (Fig. S7a-c). Thus, we can ignore the interactions within the immune pathway as long as we account for the effects of cytokines upon specific targets of the adrenal SRN. This effectively means that, during an LPS challenge, in addition to an ACTH input we will now use the time course of these cytokines as input functions. The task becomes complex as soon as we realise that the experimental evidence available comes from different cell types and animal models (Table C).

3.1 Model Equations for the SRN with Cytokine Interactions

We explored different network architectures and arrived to the scenario depicted in Fig. 5, which is associated to the following set of model equations, where the cytokine effects are shown in red:

 $\begin{aligned} d\dot{a}x1 &= \sigma_{dax1} + k_{dax1}f_{dax1}(SF1, pGR)g_{dax1}^{-}(ACTH)\phi_{dax1}(IL6) - \gamma_{dax1}dax1, \\ D\dot{a}x1 &= k_{Dax1}dax1_{\tau_{Dax1}}\phi_{Dax1}(IL6) - \gamma_{Dax1}Dax1g_{Dax1}^{\mp}(ACTH), \\ D\dot{A}X1 &= k_{DAX1}Dax1_{\tau_{DAX1}}\phi_{DAX1}(IL6) - \gamma_{DAX1}DAX1, \\ s\dot{f}1 &= \sigma_{sf1} + k_{sf1}g_{sf1}^{+}(ACTH)\phi_{sf1}(IL6) - \gamma_{sf1}sf1, \\ S\dot{f}1 &= k_{Sf1}sf1_{\tau_{Sf1}}\phi_{Sf1}(IL6) - \gamma_{Sf1}h_{Sf1}(Sf1), \\ S\dot{F}1 &= k_{SF1}Sf1_{\tau_{SF1}}\phi_{SF1}(IL6) - \gamma_{SF1}SF1g_{SF1}^{-}(ACTH), \\ s\dot{t}AR &= \sigma_{stAR} + k_{stAR}f_{stAR}(SF1, DAX1)g_{stAR}^{+}(ACTH)\phi_{stAR}(TNF\alpha)\phi_{stAR}(IL1\beta) - \gamma_{stAR}stAR, \\ S\dot{t}AR &= k_{StAR}stAR_{\tau_{StAR}}\phi_{StAR}(TNF\alpha)\phi_{StAR}(IL6) - \gamma_{StAR}h_{stAR}(StAR), \\ StARp37 &= k_{StARp37}StAR_{\tau_{StARp37}}\phi_{StARp37}(IL6) - \mu_{StARp37}StARp37g_{stARp37}^{-}(ACTH), \\ ACORT &= k_{ACORT}f_{ACORT}(StARp37) - \mu_{ACORT}ACORT, \\ \dot{\alpha}D &= k_{ACORT}f_{ACORT}(StARp37) - \mu_{ACORT}ACORT, \end{aligned}$

 $p\dot{G}R = k_{GR}f_{pGR}(ACORT)\phi_{pGR}(TNF\alpha) - \gamma_{pGR}pGR.$

¹⁷The hyphen in $TNF-\alpha$, $IL-1\beta$, and IL-6 labels was dropped to avoid confusion with a minus sign.

where

$$\begin{split} \phi_{dax1}(IL6) &= \frac{K_{dax1}^{IL6}}{K_{dax1}^{IL6} + IL6}, & \phi_{stAR}(TNF\alpha) = 1 + \frac{TNF\alpha}{K_{stAR}^{TNF\alpha} + TNF\alpha}, \\ \phi_{Dax1}(IL6) &= \frac{K_{Dax1}^{IL6}}{K_{Dax1}^{IL6} + IL6}, & \phi_{stAR}(IL1\beta) = 1 + \frac{IL1\beta}{K_{stAR}^{IL1\beta} + IL1\beta}, \\ \phi_{DAX1}(IL6) &= \frac{K_{DAX1}^{IL6}}{K_{DAX1}^{IL6} + IL6}, & \phi_{StAR}(TNF\alpha) = 1 + \frac{TNF\alpha}{K_{StAR}^{TNF\alpha} + TNF\alpha}, \\ \phi_{sf1}(IL6) &= \frac{K_{sf1}^{IL64}}{K_{sf1}^{IL64} + IL64}, & \phi_{StAR}(IL6) = \frac{K_{StAR}^{IL6}}{K_{StAR}^{IL6} + IL6}, \\ \phi_{Sf1}(IL6) &= \frac{K_{Sf1}^{IL64}}{K_{Sf1}^{IL64} + IL64}, & \phi_{stARp37}(IL6) = \frac{K_{StARp37}^{IL6}}{K_{StARp37}^{IL6} + IL6}, \\ \phi_{SF1}(IL6) &= \frac{K_{SF1}^{IL6}}{K_{SF1}^{IL6} + IL6}, & \phi_{pGR}(TNF\alpha) = \frac{K_{pGR}^{TNF\alpha}}{K_{pGR}^{TNF\alpha} + TNF\alpha}. \end{split}$$

As before, the half-maximum constants for the cytokine effects upon targets of the adrenal SRN can be understood as sensitivities. These were estimated by manually fitting the model predictions to experimental data from the LPS challenge experiments. The values are listed in Table B and represented graphically in Fig. S9d-f.

Cytokine	Target	\mathbf{Sign}	Cell type/species	Reference
$TNF\alpha$	$F\alpha$ StAR mRNA/protein		Rat testis Leydig tumor cells (LC-540)	[Sadasivam et al., 2015]
	StAR mRNA/protein	_	Mouse macrophage cell line (RAW264.7)	[Ma et al., 2007]
	DAX1 protein	+	Rat testis Leydig tumor cells (LC-540)	[Sadasivam et al., 2015]
	StAR mRNA	+	Human adrenocortical cells (NCI-H295R) $$	[Tkachenko et al., 2011]
	Cortisol	_	Bovine Zone Fasciculata cells	[Barney et al., 2000]
	pGR	_	Human Airway Smooth Muscle cells	$[\mathrm{Bouazza\ et\ al.,\ }2012]$
$IL1\beta$	StAR mRNA	+	Human adrenocortical cells (NCI-H295R)	[Tkachenko et al., 2011]
IL6	StAR mRNA/protein	+	Bovine Zona Fasciculata cells	[McIlmoil et al., 2016]
	Dax1 mRNA/protein	_	Bovine Zona Fasciculata cells	[McIlmoil et al., 2016]
SF1 mRNA/protein		+	Bovine Zona Fasciculata cells	[McIlmoil et al., 2016]
	StAR mRNA/protein	_	Bovine Zona Reticularis cells	[McIlmoil et al., 2015]
	Dax1 mRNA/protein	+	Bovine Zona Reticularis cells	[McIlmoil et al., 2015]
	$\rm SF1~mRNA/protein$	_	Bovine Zona Reticularis cells	[McIlmoil et al., 2015]
	StAR mRNA	+	Human adrenocortical cells (NCI-H295R) $$	[Tkachenko et al., 2011]
	Cortisol	+	Bovine Zona Fasciculata cells	[Barney et al., 2000]
LPS	StAR 30 kDa	_	Rat testis Leydig cells	[Allen et al., 2004]
				[Held Hales et al., 2000]
	StAR 30 kDa	+	Rat adrenal cells	[Held Hales et al., 2000]
	StAR 37 kDa	+	Rat testis Leydig tumor cells	[Allen et al., 2004]
	StAR	+	Mouse Y1 cells	[Calejman et al., 2011]
	Corticosterone	+	Rat adrenal cells	[Calejman et al., 2011]

Table C: Cytokine effects upon targets within the adrenal SRN.

3.2 Scaling Model Outputs of Cytokine Targets

Even though our model predicts the time evolution of hnRNA, mRNA, and protein concentrations, these predictions are inherently qualitative. This is mainly because we have approximated the mechanisms underlying gene regulation through Hill type functions and, to a lesser degree, because our model doesn't account for regulatory processes affecting mRNA and protein stability.

However, in order to make our model predictions comparable to the experimental data (expressed in $pg/\mu g$ for ACORT and as fold induction for all other variables), we have implemented a variable change to express the model predictions in terms of dimensionless arbitrary units. To do this consistently, all x state variables in the model were transformed according to the formula:

$$\hat{x} = \frac{x - x_{t=0}}{x_{max} - x_{min}} \sigma_x + o \textit{ffset}$$

where $x_{t=0}$ is the value of x at the time the pulse is given, x_{max} and x_{min} are the maximum and minimum value of the response, and σ_x is a scaling factor. The offset value was set to 1 for all state variables with data reported as fold induction, except for ACORT, which experiments reported as $pg/\mu g$ and for which we set an offset of 25, following the baseline observed in ACTH i.v. pulse experiments.

The scaling factors were adjusted to reproduce the data of the ACTH i.v. pulse experiment, and these remained invariant when the model was used to predict the high s.c. ACTH pulse and LPS challenge data without cytokines (Table D). The only scenario where scaling factors were allowed to vary was after modifying the existing model to account for cytokine effects. Interestingly, in this latter case most scaling factors remain invariant, except for those multiplying variables targeted by cytokines. As before, this is because we have used Hill type functions to approximate the mechanisms by which these pro-inflammatory cytokines interact with targets of the SRN.

Variable	ACTH small	ACTH high	LPS challenge	LPS challenge
Variable	i.v. pulse	s.c. pulse	without cytokines	with cytokines
ACORT	300	300	300	300
pGR	3.5	3.5	3.5	0.4
dax1 hnRNA	0.9	0.9	0.9	0.9
Dax1 mRNA	0.4	0.4	0.4	0.8
DAX1 protein	0.1	0.1	0.1	0.6
sf1 hnRNA	0.6	0.6	0.6	0.6
Sf1 mRNA	0.5	0.5	0.5	0.1
SF1 protein	0.2	0.2	0.2	0.2
stAR hnRNA	3	3	3	3
StAR mRNA	2	2	2	2
StARp37 protein	0.1	0.1	0.1	2

Table D: Scaling factors σ_x . Values in red correspond to non-statistically significant data. All scaling factors remain invariant in the model, except when the model is modified to account for cytokine inputs (last column) and only on molecular species targeted by cytokines (Fig. 5).

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