



# **Buffering and Amplifying Transcriptional Noise During Cell Fate Specification**

Elizabeth A. Urban and Robert J. Johnston Jr.\*

Department of Biology, Johns Hopkins University, Baltimore, MD, United States

The molecular processes that drive gene transcription are inherently noisy. This noise often manifests in the form of transcriptional bursts, producing fluctuations in gene activity over time. During cell fate specification, this noise is often buffered to ensure reproducible developmental outcomes. However, sometimes noise is utilized as a "bet-hedging" mechanism to diversify functional roles across a population of cells. Studies of bacteria, yeast, and cultured cells have provided insights into the nature and roles of noise in transcription, yet we are only beginning to understand the mechanisms by which noise influences the development of multicellular organisms. Here we discuss the sources of transcriptional noise and the mechanisms that either buffer noise to drive reproducible fate choices or amplify noise to randomly specify fates.

Keywords: transcriptional bursting, expression noise, cell fate, Waddington landscape, MS2 hairpin, smFISH

### **OPEN ACCESS**

### Edited by:

Alexander William Shingleton, University of Illinois at Chicago, United States

#### Reviewed by:

Aishwarya Swaminathan, University of Massachusetts Medical School, United States Zong Wei, Salk Institute for Biological Studies, United States

#### \*Correspondence:

Robert J. Johnston Jr. robertjohnston@jhu.edu

### Specialty section:

This article was submitted to Epigenomics and Epigenetics, a section of the journal Frontiers in Genetics

Received: 18 August 2018 Accepted: 15 November 2018 Published: 29 November 2018

#### Citation

Urban EA and Johnston RJ Jr (2018) Buffering and Amplifying Transcriptional Noise During Cell Fate Specification. Front. Genet. 9:591. doi: 10.3389/fgene.2018.00591

### INTRODUCTION

Cell fate specification during development is often thought of as a highly reproducible process where tight regulation of gene expression determines precise cell fates. These robust, reproducible fates are driven by cell lineage history and signaling. A beautiful example of lineage-driven cell fate specification occurs in the nematode Caenorhabditis elegans. Nearly all of the cells of the worm derive from stereotypical division patterns and gene expression that have been very well-mapped (Sulston et al., 1983; Maduro, 2010). For example, the ASEL/R neurons are derived from a distinct lineages that are regulated by a network of transcription factors and microRNAs (Hobert et al., 2002; Johnston and Hobert, 2003, 2005; Chang et al., 2004; Johnston et al., 2005, 2006; Poole and Hobert, 2006; Sarin et al., 2007; Cochella et al., 2014). Conversely, one of the best understood paradigms for signaling-driven development is observed in the eye of the fruit fly Drosophila melanogaster. In the fly eye, precise and progressive signaling cues determine retinal cell fates, generating a near-crystalline pattern of ommatidia (Wolff and Ready, 1991; Kumar, 2011, 2012). All photoreceptors develop from the same pool of undifferentiated progenitor cells (Kumar, 2012). The final photoreceptor to develop, the R7, is generated through combinatorial Notch, RAS, and EGFR signaling from the other photoreceptor subtypes (Kumar, 2012). The transformation of a pool of undifferentiated progenitor cells into 800 ommatidia arranged in a crystalline pattern across the retina highlights the importance of signaling as a mechanism to determine robust cell fates.

Lineage and signaling cues provide a framework for the energy landscape of cell fate specification first described by Waddington (1957). In Waddington's energy landscape, "hills" and "valleys" represent developmental energy potential. These geographical landmarks are used to guide cells

1

toward terminal differentiation. Lineage and signaling inputs push cells into valleys of low potential energy, thereby restricting them to specific fates (Waddington, 1957).

The road to differentiation isn't always smooth. Lineage and signaling must overcome molecular noise to drive cell fates. Gene expression noise is characterized by differences in the level of gene expression between cells of the same type. It arises due to random fluctuations in the level of mRNA or protein expressed at a given time in an individual cell. Noise roughens the road in Waddington's developmental landscape, generating "bumps" in gene expression that lineage and signaling cues often override (Balazsi et al., 2011) (Figure 1). However, sometimes these bumps are employed during development to generate a "fork" in the road, causing a cell to randomly fall into one of two fates. Slight variations in the level of noise change the contours of the fork, steering the cell toward one of the fates at a particular frequency. This random choice between fates is called stochastic cell fate specification (Figure 1). Together, stochastic fate specification complements lineage- and signaling-based mechanisms to further diversify cell types during development (Johnston and Desplan, 2010).

In single-celled organisms, stochastic cell fate choices generate cellular diversity and facilitate survival in adverse conditions. In the bacterium *Bacillus subtilis*, about 10% of cells transiently enter a competent state in which they are able to take up DNA, while the other 90% remain in a dormant state. This diversity allows the competent population to survive under

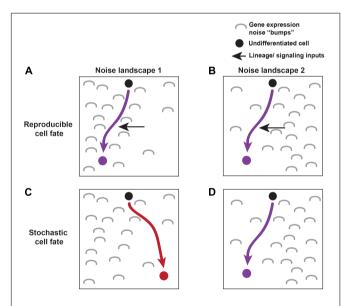


FIGURE 1 | Lineage, signaling, and noise make up the molecular environment driving cell fate specification. An undifferentiated cell (black) moves towards its terminal cell fate based on its molecular landscape (described by Waddington's energy landscape). Gene expression noise effects the landscape through which cells differentiate. Two different noise landscapes are shown (A,C vs. B,D). Noise is depicted by gray "bumps." Reproducible fates are able to overcome noise in both landscapes by utilizing lineage and signaling cues to push them towards a particular fate (A,B). Other cells choose their fate stochastically, where noisy inputs shape the molecular environment driving the stochastic fate decision (C,D).

stressful conditions (Maamar and Dubnau, 2005; Maamar et al., 2007). These two cell populations are genetically identical and are exposed to the same external cues, indicating that this stochastic fate choice is independent of initial genetic and environmental conditions. Rather, intrinsic gene expression noise causes isogenic populations of cells to exhibit varying levels of expression of numerous genes, tipping the balance toward one fate or the other within individual cells (Losick and Desplan, 2008; Raj and van Oudenaarden, 2008; Enver et al., 2009; Balazsi et al., 2011).

Although noise plays a role in stochastic cell fate specification, the mechanisms driving these decisions are poorly understood. In addition to studies performed in bacteria and cultured cells, recent advances in imaging allow the study of stochastic fate choices in multicellular organisms (Box 1 and Figure 2). In this review, we discuss gene expression noise and transcriptional bursting, and examine how these sources of noise are buffered in deterministic fate choices and amplified in stochastic cell fate decisions. Understanding the role of gene expression noise in cell fate decisions is crucial for determining when and how these systems go awry.

### SOURCES OF NOISE IN GENE EXPRESSION

Gene expression noise arises in many different ways, and the extent of noise varies dramatically among different genes and organisms (Balazsi et al., 2011; Suter et al., 2011). Gene expression noise has two major types: **extrinsic** and **intrinsic** noise (Elowitz et al., 2002). Extrinsic noise arises from environmental perturbations surrounding the cell, and within the cell, such as changes in the local distribution and concentration of general

**BOX 1** Studying the dynamics of transcriptional bursting. With recent advances in imaging technologies, we can now study transcriptional bursting in greater detail using techniques such as single molecule RNA fluorescent *in situ* hybridization (smFISH) and the MS2/MCP system (Bertrand et al., 1998; Gregor et al., 2014; Lenstra et al., 2016) (**Figure 2**). Each of these techniques provides unique insight into the kinetic parameters regulating transcriptional bursting.

smFISH uses fluorescent DNA probes to label nascent RNA transcripts. The amount of RNA produced at the nascent site of transcription is reflected in the fluorescence intensity. Therefore, the elongation rate, length of a transcript, and probe density are used to calculate the exact number of RNA molecules produced (Little et al., 2013; Zoller et al., 2018). Even more information can be extracted from multi-color FISH experiments. For example, the 5' and 3' end of a transcript can be labeled in two different colors, or introns and exons can be differentially labeled, allowing the temporal state of transcription to be analyzed in fixed tissues (Little et al., 2013; Zoller et al., 2018) (Figure 2A).

The MS2/MCP system provides a complementary system to study transcriptional bursting parameters. Using this system, multiple copies of a sequence coding for MS2 RNA hairpins are incorporated into a gene of interest (Bertrand et al., 1998) (Figure 2B). Upon transcription, these hairpin sequences are recognized by the MS2 coat protein (MCP). MCP is directly tagged with GFP and expressed at low levels in the cells or tissue of interest. When the hairpins are expressed, MCP-GFP binding to the transcript allows nascent transcription to be monitored in real time in living cells and tissues (Gregor et al., 2014). Thus, dynamics of transcription, such as the burst frequency, duration, amplitude, and total amount of RNA can be determined.

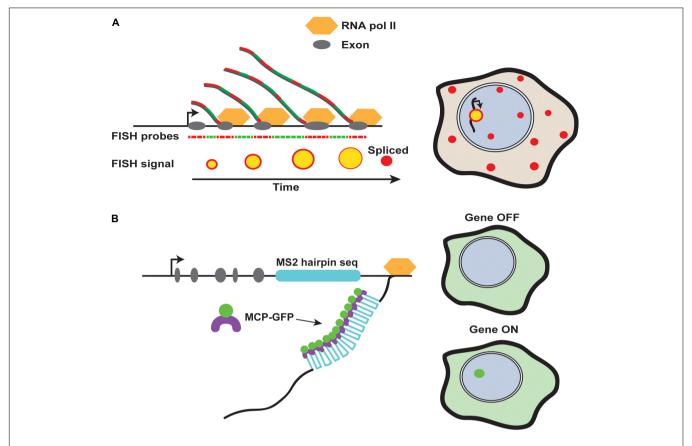


FIGURE 2 | Methods for detection of nascent RNA transcripts. (A) Fluorescence in situ hybridization tracks transcriptional dynamics. Different colored fluorescent probes against intronic and exonic regions of a gene can be alternated to monitor the progression of transcription in fixed cells. Sites of nascent transcription will contain both fluorophores, while mature transcripts will be labeled by only one fluorophore due to splicing. (B) The MS2 system allows for detection of transcripts in living or fixed tissues. The nascent site of transcription is detected by binding of MS2 coat protein (MCP) to MS2 RNA hairpins, which can be incorporated into transgenes and endogenous loci in multiple copies.

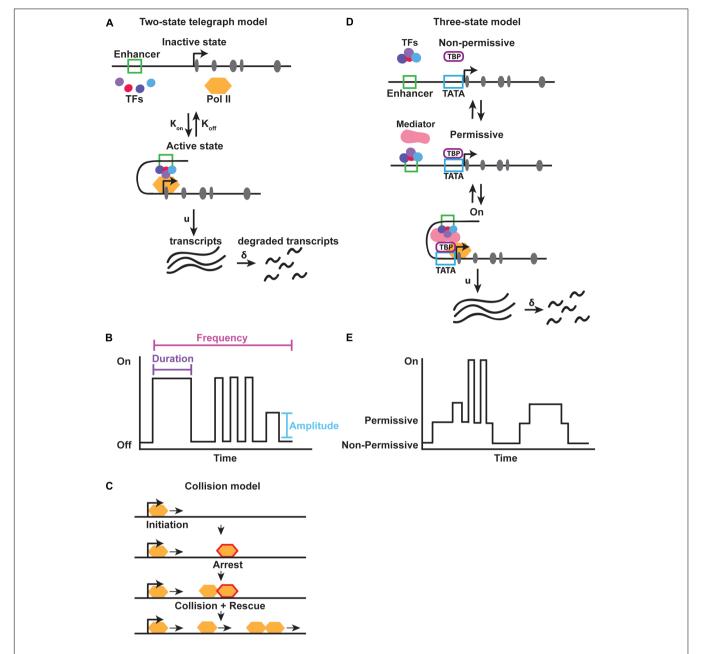
transcription factors or other proteins. Extrinsic noise varies from cell to cell but has the potential to affect all genes, and is therefore a **gene-independent characteristic**. Conversely, intrinsic noise can be attributed to things such as variation in transcription factor and chromatin modifier binding at individual gene loci. Thus, intrinsic noise is considered a **gene-dependent characteristic** because its effects vary from gene-to-gene and from cell-to-cell (Gregor et al., 2014; Fujita et al., 2016).

The majority of gene expression noise arises from stochasticity in mRNA production due to the random binding of transcription factors and other transcriptional machinery to the DNA, as well as due to the turnover of mRNA and protein (McAdams and Arkin, 1997; Elowitz et al., 2002; Ozbudak et al., 2002; Swain et al., 2002; Raser and O'Shea, 2004). Binding and dissociation events at the enhancer or promoter occur with particular rates,  $K_{on}$  and  $K_{off}$  respectively (**Figure 3A**). These rates are largely dependent on protein availability and the presence of other bound proteins. Extrinsic fluctuations produce variability in the local concentrations of proteins, leading to alterations in the probability of protein binding over time. Gene expression noise also arises from variation in the transcription initiation rate  $(\mu)$  and mRNA degradation rate  $(\delta)$  (**Figure 3A**). For highly

regulated genes, noise is strongly buffered. For example, noise is reduced by controlling the degradation rate of mRNA. mRNAs with faster turnover rates have lower noise, because the amount of protein that can be translated before the transcript is degraded is limited (Swain, 2004; Cai et al., 2006; Singh, 2011). While gene expression noise also arises from translation, we focus here on the role of intrinsic transcriptional noise, in particular how transcription occurs in non-continuous bursts, referred to as **transcriptional bursting** (Ross et al., 1994; Peccoud and Ycart, 1995; Newlands et al., 1998; Takasuka et al., 1998; Golding et al., 2005).

## Two-State Models of Transcriptional Bursting

Transcriptional bursting occurs when a gene promoter fluctuates between an "on" and "off" state for different periods of time. Each time the promoter switches to an "on" state, a burst of transcription is produced (Peccoud and Ycart, 1995; Gregor et al., 2014; Kumar et al., 2015). The frequency, duration, and amplitude of bursts determine the total amount of RNA produced from a particular gene (**Figure 3B**). In this manner, transcriptional bursting is a mechanism that produces gene



**FIGURE 3** | Models for transcriptional bursting. **(A)** The two-state bursting model suggests that a gene flips between an "inactive" and "active" state at a particular rate, K<sub>on</sub> and K<sub>off</sub>. In the "active" state RNA transcripts are produced at a rate (u) and degraded at a rate (8). **(B)** Transcriptional bursting parameters. Each burst occurs for a particular duration (i.e., period of time) with a distinct amplitude (i.e., strength) and at a particular frequency. Adapted from Raj et al. (2006). **(C)** A modified two-state collision model suggests that intrinsic interactions between RNA polymerase and DNA are sufficient to create transcriptional bursts. Polymerase arrest, followed by collision and rescue allow for individual bursts to occur. Adapted from Tantale et al. (2016). **(D,E)** The three-state bursting model suggests that a slow burst cycle and fast burst cycle are controlled by the binding of TATA binding protein (TBP) to the promoter, and subsequent recruitment of the co-activator Mediator, allowing a gene to be in either a non-permissive, permissive, or active state of transcription.

expression noise. This noise is often described by the amount of variation in gene expression seen among cells of a population.

There are two major models of transcriptional bursting: the two-state and three-state models (Figure 3). Multiple versions of the two-state model have been proposed, but the two-state telegraph model is most commonly used. In the two-state telegraph model first described by Peccoud and Ycart,

a promoter's on/off state is controlled by intrinsic variability in transcription initiation (Peccoud and Ycart, 1995; Lenstra et al., 2016). In this model, the promoter for a gene is in an active or inactive state for different periods of time (**Figure 3A**). While in the "on" state, the gene produces bursts of transcripts.

*In vitro* single molecule imaging of RNA polymerase and mRNA production has provided evidence for a second two-state

model known as the **two-state collision model**. In the collision model, the "off" state is determined by RNA polymerase arrest. Transcription is then rescued back to the "on" state by collision of the next polymerase, indicating that bursting can be produced by general transcriptional machinery dynamics alone (Fujita et al., 2016) (**Figure 3C**). Although arrest and release can be visualized *in vitro*, whether this pausing and collision actually occurs *in vivo* is unknown. It is possible that promoter proximal pausing could provide a long enough arrest to allow collision and bursts.

Evidence for the two-state collision model stems from in vitro work, but a related mechanism for bursting has been observed in bacteria. Bacterial DNA is packaged into topologically constrained loops (Postow et al., 2004; Hardy and Cozzarelli, 2005). As RNA polymerase transcribes these loops, it induces positive supercoiling ahead of the transcription machinery and negative supercoiling behind the transcription machinery. As positive supercoiling accumulates ahead of the polymerase, initiation is blocked, and elongation is slowed. Transcription remains arrested until gyrase (topoisomerase II) relieves the positive supercoiling, producing a burst of transcription until the next arrest. These observations suggest that the binding of gyrase to DNA promotes transcriptional bursts by inducing RNA polymerase II pause release (Chong et al., 2014). By combining the in vitro findings about the collision model, with in vivo work in bacteria, it is interesting to consider the possibility of a similar mechanism of pause and release creating bursts in eukaryotes as

Eukaryotic DNA is packaged into topologically associated domains (TADS) (Jackson and Pombo, 1998; Ma et al., 1998; Dixon et al., 2012; Nora et al., 2012; Sexton et al., 2012). Although the direct mechanisms driving TAD formation are still being investigated, one common model is the loop extrusion model, where DNA is fed through cohesin as a loop, and is halted at boundaries containing CTCF proteins. Simulation work by the group of Racko and colleagues has shown that transcription within a region of DNA, constrained by cohesin on each end, can create the spatial constraint necessary to promote negative supercoiling accumulation within intra-TAD structures (Nasmyth, 2011; Sanborn et al., 2015; Dixon et al., 2016; Fudenberg et al., 2016; Stigler et al., 2016; Racko et al., 2018). Whether this negative intra-TAD supercoiling is seen in vivo is unknown, but with similar physical constraints to bacterial chromosomes, it will be interesting to see if this mechanism of bursting is conserved and how it relates to TAD formation.

While the telegraph and collision models describe general mechanisms controlling transcriptional bursting across all genes (i.e., gene-independent), they do not explain how genes in the same nucleus exhibit different kinetics. Promoter architecture, chromatin state, enhancer-promoter interactions, transcription factor copy number, and binding kinetics control different bursting parameters such as amplitude, duration, and frequency of bursting from individual loci (Skupsky et al., 2010; Suter et al., 2011; Dar et al., 2012; Larson et al., 2013; Senecal et al., 2014; Bartman et al., 2016; Fujita et al., 2016) (Figure 3B). These different parameters ultimately determine the levels of RNA for each gene and thus the amount of noise within a population, which are detected by techniques such as single molecule

Fluorescent *in situ* hybridization and the MS2 hairpin system (**Box 1** and **Figure 2**). We explore these different contributions to bursting in more detail below.

## Three-State Model of Transcriptional Bursting: Promoter Contributions

As the landing site for the general transcription machinery and RNA polymerase, the core promoter plays a key role in regulating transcriptional bursting. Promoter architecture varies across the genome. Studies in mouse B-cell culture elegantly showed that unique core promoter elements differentially and precisely regulate either the burst size or frequency (Hendy et al., 2017). For example, the TATA box of the MHC class 1 gene PD1 regulates both burst size and frequency, the Initiator element regulates only frequency, and the transcription factor Sp1 binding site regulates only size. These differences in promoter composition allow for tissue-specific bursting effects, which can be modulated by *trans*-acting factor concentrations and binding affinities (Senecal et al., 2014; Hendy et al., 2017).

The TATA box is a critical determinant of gene expression noise. Genes lacking a TATA box are associated with lower noise, while the presence of a TATA box promotes higher noise (Figure 4). In general, promoters of housekeeping genes lack TATA boxes, suggesting a mechanism by which constitutively and highly expressed genes buffer gene expression noise throughout development (Blake et al., 2006; Hornung et al., 2012; Larson et al., 2013; Zabidi et al., 2015; Tantale et al., 2016; Hendy et al., 2017).

How does the presence of a TATA box promote higher gene expression noise? A study examining the roles of the TATA binding protein (TBP) and Mediator complex suggested a **three-state bursting model** (Tantale et al., 2016) (**Figure 3D**). Binding of TBP to the promoter induces the promoter to switch

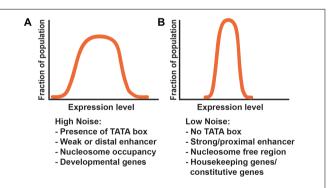


FIGURE 4 | Gene architecture contributes to noise. (A,B) Graphs represent theoretical frequency distributions of gene expression levels across a population of cells. Common components associated with differential effects on noise are depicted below each distribution, and include promoter architecture, enhancer strength and position, and chromatin state. (A) Genes with high noise generally show a broad distribution of expression levels. Common factors associated with high noise include the presence of a TATA box, weak or distal enhancer elements, and high nucleosome occupancy. (B) Low noise-producing genes have a tighter distribution of expression. Common factors associated with low noise include the absence of a TATA box, strong or proximal enhancers, and nucleosome free regions.

from a non-permissive to a permissive state. This generates long periods of inactivity followed by periods of high activity (Figures 3D,E). The binding of the co-factor Mediator to TBP and the promoter then enables rapid on/off cycling of the promoter. Retention of TBP at the promoter may allow rapid re-initiation of transcription by Mediator. This multi-state model has been used to describe bursting in mammalian cells where a non-permissive, or refractory, state is seen between bursts that is potentially driven by chromatin remodeling events (Harper et al., 2011; Suter et al., 2011). In some cells, a gene may be poised for activation, while in other cells, the gene may remain in a non-permissive refractory state (Harper et al., 2011). This refractory period increases noise, creating heterogeneities within a population.

Additional work studying the role of promoters in transcriptional bursting has brought up conflicting data about whether promoter elements buffer or foster noise. In studies by Hendy and colleagues, mutation of core promoter elements increased noise, indicating that promoter elements buffer and tune noise naturally (Hendy et al., 2017). Since many factors bind the promoter, it makes sense that promoters would be regions where noise is buffered by higher binding affinities (i.e., K<sub>on</sub>). In contrast, other experiments suggest that promoters have naturally evolved to produce noise. For example, synthetic Escherichia coli promoters were generated in the lab through directed evolution. These artificially evolved promoters drove gene expression at similar levels to their endogenous promoters (Wolf et al., 2015). However, the promoters evolved in the lab produced lower noise than promoters naturally found in E. coli (Bury-Mone and Sclavi, 2017). Taken together, these studies suggest that evolutionary pressures may both promote and reduce noise at the level of the promoter sequence, which in turn tunes noise for specific genes (Wolf et al., 2015).

#### **Enhancer Contributions**

In addition to promoter architecture, enhancer-promoter interactions act as gene-dependent regulators of transcriptional bursting. Enhancers are bound by gene-specific combinations of transcription factors to control spatial and temporal expression (Levine, 2010; Buecker and Wysocka, 2012). In addition to binding specific transcription factors, enhancers also assist in recruiting the general transcriptional machinery. What we know about enhancer function has mainly been assessed through reporter assays and transgenes. These approaches give a broad view of expression, but the role of endogenous enhancers in transcriptional bursting at the single cell level is less understood (Bulger and Groudine, 2011; Spitz and Furlong, 2012).

One mechanism by which enhancers affect bursting is through enhancer-promoter looping (Lewis, 1978; Lettice et al., 2003; Fukaya et al., 2016). Kinetic fluctuations in the cellular environment, such as changes in the concentration and distribution of transcription factors, change the probability of enhancer-promoter contact from cell-to-cell. These changes in enhancer-promoter interactions lead to differences in bursting between cells (Lenstra et al., 2016; Tantale et al., 2016).

The effect of a particular enhancer on bursting is also dependent upon enhancer strength, or how well it can find

and loop to its promoter. Weak enhancers drive a lower burst frequency than strong enhancers (Fukaya et al., 2016) (Figure 5A). Additionally, enhancers that are closer to promoters have higher burst frequencies compared to distal enhancers, indicating that proximity increases the probability of an enhancer finding its promoter (Fukaya et al., 2016) (Figure 5B). DNA looping co-factors known as insulators block interactions between particular regions of DNA (Geyer and Corces, 1992; Dorsett, 1993; Bell et al., 2001). An insulator element inserted between an enhancer and promoter reduces bursting by blocking the enhancer-promoter interaction (Fukaya et al., 2016) (Figure 5C). In contrast, tethering an enhancer to a promoter increased the frequency of bursts (Bartman et al., 2016). Thus, enhancer-promoter interactions make critical contributions to transcriptional noise by controlling burst frequency.

While the study of simple enhancer-promoter interactions provides important insight into the effects of enhancer looping on bursting, many genes have a more complex regulatory logic. For example, some enhancers activate multiple genes and some genes are regulated by multiple enhancers, raising two questions: can an enhancer simultaneously activate multiple promoters, and can multiple enhancers simultaneously regulate a single promoter? To address the first question, Fukaya and colleagues examined the activity of an enhancer placed between two promoters (Fukaya et al., 2016) (Figure 5D). The enhancer simultaneously activated both promoters in the *Drosophila* embryo, leading to coordinated bursting (Fukaya et al., 2016) (Figure 5D).

While an enhancer can act on multiple promoters at once, can multiple enhancers act on the same promoter simultaneously? To investigate this second question, Bothma and colleagues looked at how primary and shadow enhancers regulate bursting of a single gene and found that these enhancers do not act simultaneously, indicating that the overall bursting parameters of a gene are a result of an additive effect of combined enhancer inputs (Bothma et al., 2015) (**Figure 5E**).

Complex networks of enhancers, such as super-enhancers, regulate genes involved in cell type specification and development (Hnisz et al., 2017). Super-enhancers consist of a collection of enhancers with high densities of transcription factors that regulate a gene. Simulation studies of super-enhancer regulation of transcription by Hnisz and colleagues has suggested that the increased strength of super-enhancers should lead to high burst frequencies, creating constitutive gene expression. One potential mechanism by which super-enhancers could increase transcriptional activity is through phase transitions, bringing all enhancers in close proximity to their promoter through networks of cooperative binding. This may explain a mechanism by which gene expression noise is reduced at genes involved in processes regulating cell type, as these are the genes often associated with super-enhancers (Hnisz et al., 2017). Whether all enhancers are acting simultaneously or independently is still unknown, but it will be interesting to see if super-enhancers actually form phase transitions in vivo, and how each enhancer contributes directly to transcriptional activity. Together, this model of phase transition of super-enhancers, in addition to the findings of Fukaya et al. (2016) suggest that beyond the role of enhancers, chromosome topology plays an integral role

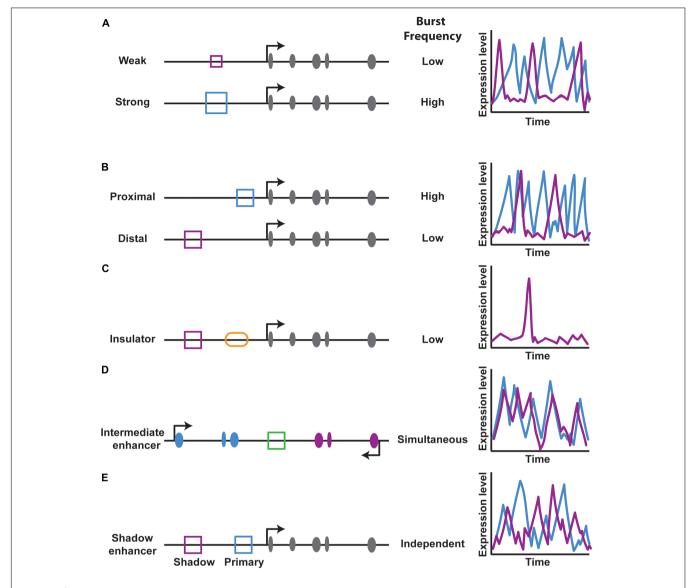


FIGURE 5 | Enhancer effects on transcriptional burst frequency. (A,B) Enhancer strength and distance regulate transcriptional burst frequency. Schematics on the left of gene loci depict exons in gray, the enhancer type (blue or purple square) and other cis-regulatory elements. Graphs on the right show theoretical burst frequencies from each enhancer over time. (C) The presence of an insulator element (orange oval) between an enhancer-promoter pair reduces transcriptional burst frequency. (D,E) An enhancer placed between two promoters acts on both promoters simultaneously, while the presence of two enhancers that act on the same promoter act independently, as in the case of shadow enhancers.

in the regulation of gene expression and enhancer-promoter interactions.

### **Chromatin Contributions**

The studies described above indicate that enhancer-promoter looping is a critical regulator of transcriptional bursting and thus gene expression noise. Higher-order chromatin studies have revealed that, beyond enhancer-promoter contacts, chromosomes are organized into topologically associated domains, or TADs (Dixon et al., 2016). Within these TADs, DNA contacts are made and facilitated by insulator and architectural proteins such as CTCF and cohesin (Dowen et al., 2014). Depletion of CTCF or cohesin subunits result in intra-TAD

contact changes, associated with misregulation of gene expression (Handoko et al., 2011; Seitan et al., 2013; Sofueva et al., 2013; Zuin et al., 2014; Ing-Simmons et al., 2015; Dixon et al., 2016). Studies in mammalian cells have revealed that intra-TAD CTCF binding promotes and stabilizes enhancer-promoter interactions to regulate gene expression. These stabilized interactions result in a decrease in transcriptional bursting and noise (Ren et al., 2017).

The role of the architectural insulator protein CTCF in controlling enhancer-promoter interactions suggests that chromatin state and topology are also important regulators of transcriptional bursting. Hendy et al. (2017) have shown that the level of chromatin compaction and the kinetics

required for heterochromatin opening are rate limiting steps in transcriptional bursting. Promoters with nucleosome-free regions downstream of the transcriptional start site display low gene expression noise (**Figure 4**). In contrast, promoters with higher nucleosome occupancy are associated with high noise, suggesting that chromatin remodeling during transcription initiation may contribute to bursting (**Figure 4**) (Cairns, 2009). Studies in cell culture showed that increasing acetylation at target transgenes resulted in an increase in burst frequency. These findings indicate that H3K27 acetylation around the promoter may act as a possible regulator of transcriptional burst frequency (Nicolas et al., 2018).

While H3K27 acetylation at the promoter is an activating histone mark and increases bursting, it is interesting to consider how repressive marks, such as H3K27 tri-methylation affect bursting. H3K27me3 is spread to create repressive heterochromatin by the Polycomb repressive complex (Schuettengruber et al., 2017). Despite their opposing roles, some genes contain both active RNA polymerase (CTD Serine 5, 7, and 2 phosphorylated) and repressive H3K27me3. Genes containing these conflicting markers show higher levels of gene expression noise when compared to fully active genes or fully Polycomb-repressed genes (Kar et al., 2017) (Figure 6). These genes with higher noise are often located in close proximity to

fully Polycomb-repressed genes, suggesting that heterochromatin spreading may play a role in producing noise, but the bursting parameters most affected by heterochromatin formation are still not understood.

Beyond Polycomb, little is known about the contributions of specific chromatin factors to transcriptional bursting. Recent work has examined the effect of chromatin environment on bursting by inserting transgenes into various locations within the genome of yeast and mammalian cells. Transcriptional burst size and frequency varies by insertion site, indicating that the local chromatin environment may shape bursting (Dar et al., 2012). However, the exact chromatin features that are responsible for these position effects are still unclear. Future studies could assess the effects of chromatin remodelers, pioneer transcription factors, histone modifications, and DNA marks in regulating transcriptional bursting and gene expression noise.

It is clear that enhancers and promoters play a role in regulating gene-specific bursting, while variation in DNA-polymerase interactions affect bursting genome-wide (Little et al., 2013; Fujita et al., 2016). Determining the molecular mechanisms of gene-specific and genome-wide bursting and elucidating how these two components are integrated *in vivo* is an area of intense study. With advances such as single molecule approaches, it is now possible to probe the effects of

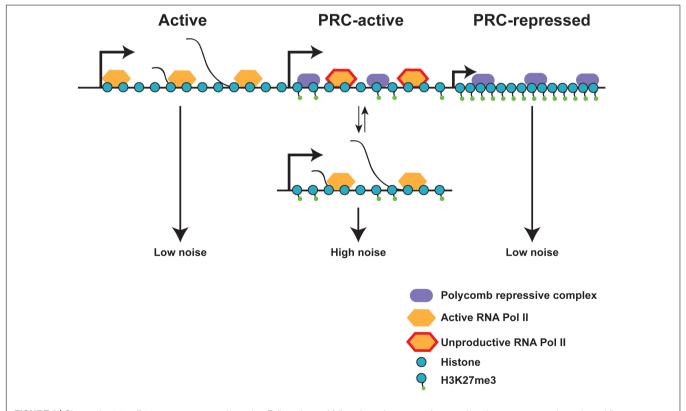


FIGURE 6 | Chromatin state effects on gene expression noise. Fully active and fully polycomb-repressed genes show low gene expression noise, while polycomb-active genes show high expression noise. Polycomb-active genes are marked by both polycomb repressive marks (H3K27Me3) and active RNA polymerase II. These polycomb-active genes are often found in close proximity to polycomb-repressed genes, and may flip between the active and inactive state, increasing gene expression noise. Local heterochromatin spreading due to proximity of polycomb-repressed genes may aid in the switch between active and repressed states. Adapted from Kar et al. (2017).

intrinsic gene expression noise on cellular heterogeneity and stochastic cell fate choices (Box 1). Next, we discuss the roles that transcriptional bursting and noise play in reproducible and stochastic specification of cell fates.

### HOW NOISE INFLUENCES CELL FATE SPECIFICATION

How does gene expression noise affect cell fate decisions during development? To ensure reproducible cell fates, developmental pathways must overcome noisy gene expression. Cells in reproducible lineages must therefore use noise buffering mechanisms to allow for proper pattern formation and reduce the effects of noise (Little et al., 2013). In contrast, isogenic populations of cells such as clonal populations of bacteria, harness gene expression noise as a mechanism to diversify cell fates (Balazsi et al., 2011). Stochastic cell fate decisions are also observed in multicellular organisms, but the mechanisms controlling these processes are still poorly understood.

### Noise Buffering in Deterministic Fate Choices

One of the most elegant examples of precise developmental patterning is the early *Drosophila* embryo. Expression patterns of the gap, segment polarity, and pair rule genes have been thoroughly mapped, but it was not until the development of smFISH and the MS2 hairpin system that the expression dynamics of these genes could be studied in greater detail (Garcia et al., 2013; Little et al., 2013; Gregor et al., 2014) (**Box 1** and **Figure 2**).

With these techniques, it has been shown that the genes controlling early embryo patterning exhibit transcriptional bursting, producing gene expression noise (Gregor et al., 2007; Pare et al., 2009; Little et al., 2013; Bothma et al., 2014). The level of transcription at developmental patterning gene loci varies by about 45% between nuclei (**Figure 7A**). This level of noise is similar to that observed for non-developmental genes in *Drosophila*, which is surprising given that development is thought to be tightly regulated (Little et al., 2013). How is this high level of noise controlled to create a reproducible body plan?

During development, the early *Drosophila* embryo is a cytoplasmic syncytium, meaning that all of the mRNA produced by high- and low-expressing nuclei is transported into the same cytoplasmic environment. One mechanism by which noise is buffered in the early embryo is by mRNA diffusion throughout the syncytium, resulting in a spatiotemporal averaging of mRNA concentration (**Figure 7A**). This averaging dampens the noise produced by transcription from individual nuclei and allows for precise development (Little et al., 2013).

Although the cytoplasmic syncytium buffers a majority of the gene expression noise produced during early development, there are other forms of noise buffering that may act directly at the gene locus during transcriptional bursting. Many of the early *Drosophila* patterning genes are under the control of multiple enhancers, including back-up, or secondary enhancers, known as shadow enhancers, to ensure their precise spatial and temporal

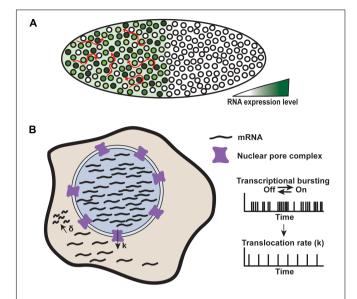


FIGURE 7 | Mechanisms of noise buffering. (A) Transcriptional noise produced by the Hunchback gene is buffered in the *Drosophila* embryo. Each nucleus expresses Hunchback at varying levels, but RNA levels are buffered by spatiotemporal averaging across the cytoplasmic syncytium, denoted by red arrows. Adapted from Stoeger et al. (2016). (Β) Nuclear retention of RNA acts to reduce cytoplasmic noise. RNA is produced through rapid bursts within the nucleus and translocated through the nuclear pour complex at a rate, k, which is slower than the RNA production rate. RNA is degraded at a rate, δ. Faster nuclear production compared to translocation constrains noise to the nucleus, while buffering cytoplasmic noise. Adapted from Battich et al. (2015).

expression (Hong et al., 2008; Levine, 2010; Bothma et al., 2014, 2015). Indeed, a weak shadow enhancer can act additively with its primary enhancer to control transcriptional bursting and reduce expression noise (Bothma et al., 2015). Other forms of shadow enhancers, such as "dark" shadow enhancers promote robust expression patterns through the combinatorial use of enhancers and negative feedback (Yan et al., 2017).

Despite our knowledge of noise buffering in *Drosophila* embryos, understanding buffering in intact mammalian systems remains a challenge. Nevertheless, studies in cultured mouse cells and liver tissue using smFISH and single cell RNA-sequencing have begun to elucidate noise buffering in mammalian cells (Bahar Halpern et al., 2015; Battich et al., 2015; Stoeger et al., 2016). These studies have uncovered another potential mechanism by which cells buffer noise. For a subset of genes, noise is high within the nucleus but reduced in the cytoplasm, suggesting that, in addition to noise buffering by mRNA degradation, regulated mRNA export from the nucleus acts as a filter for noise (**Figure 7B**).

Thus, many layers of noise buffering exist to ensure reproducible cell fates despite large variability in gene expression between nuclei. Buffering occurs directly at the gene locus through the use of distinct functional enhancer and promoter elements, or post-transcriptionally through controlled nuclear export or averaging across a syncytium. Although not discussed in detail in this review, gene regulatory mechanisms such as

negative feedback loops also reduce noise within larger regulatory systems (Raser and O'Shea, 2005; Chalancon et al., 2012).

### Noise Amplification Through Positive Feedback and Thresholding in Stochastic Cell Fate Decisions

In contrast to deterministic or reproducible cell fate specification where noise is buffered, stochastic cell fate specification utilizes transcriptional noise to generate diversity in cell fates. Some of the best-studied examples involve bet-hedging strategies. For example, within an isogenic population, a small subset of cells will take on a particular fate to diversify the population. These cells are said to be "hedging their bets" for the chance that the population can continue to survive and reproduce under stressful environmental conditions (Balazsi et al., 2011).

The common mechanism underlying these decisions often involves bistable switching between high and low expression states. A high expression state occurs when gene expression spikes above a threshold in an individual cell. Cells that never reach threshold therefore remain in the resting state (**Figure 8A**). Once the threshold is surpassed, a positive feedback loop stabilizes high expression levels to maintain the cell fate (**Figure 8A**). In these systems, noise provides the gene expression variability necessary for switching, while positive feedback stabilize the fate decisions (Maamar et al., 2007; Balazsi et al., 2011).

One of the best-understood examples of stochastic cell fate choice involves the decision between competence and dormancy in Bacillus subtilis. Competence allows individuals to take in DNA from the environment that may provide a selective advantage to overcome stresses, at the risk of not surviving dangerous conditions. Noisy expression of the transcription factor ComK regulates this decision (Figure 8). Within a clonal population, each cell exhibits transcriptional bursting of ComK, producing noise between individual organisms (Mugler et al., 2016). In most cells, fluctuations in expression remain subthreshold (Figure 8B), resulting in a dormant cell fate. However, in a subset of cells, ComK expression surpasses a threshold, thereby inducing a switch to the competent fate (Figure 8C). At high levels of ComK, the protein binds directly to its own promoter as a dimer, stimulating its expression through a positive feedback loop (Figure 8A). This thresholding and feedback mechanism results in  $\sim$ 10% of the bacterial population in the competent state at any given time (Maamar and Dubnau, 2005; Suel et al., 2006; Maamar et al., 2007).

A similar mechanism controls fate decisions for HIV-1. A key regulator of the HIV-1 life cycle is the transcription factor Trans-Activator of Transcription (Tat). If the expression of Tat reaches a certain threshold, its production will be stabilized through a positive feedback loop, thereby promoting active replication rather than proviral latency (Weinberger et al., 2005, 2008). Additionally, recent work on the mechanisms regulating Tat expression has shown that Tat is expressed in transcriptional bursts, supporting a role for an excitatory state being met through fluctuations in transcription, similar to that seen in bacteria (Weinberger et al., 2008; Hendy et al., 2017).

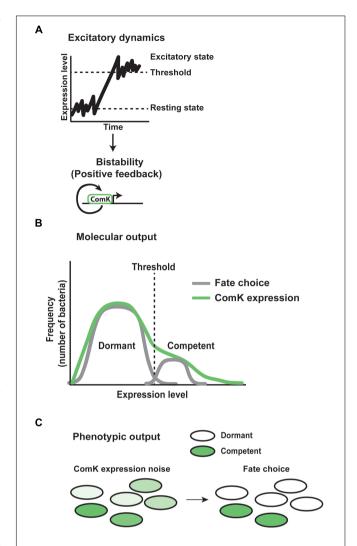


FIGURE 8 | Mechanisms of stochastic cell fate specification. Gene expression noise plays a critical role in determining stochastic cell fates in the bacteria Bacillus subtilis. (A) In Bacillus populations, gene expression fluctuates between a resting and excitatory state. Excursions above a threshold level of expression induce bistability. A positive feedback loop for the protein ComK acts to stabilize the excitatory state of expression. (B) As a result of gene expression noise within the population of bacteria, the molecular output of excitatory dynamics and positive feedback are reflected by individual cells showing varying levels of ComK expression. Cells above threshold enter the competent fate, while cells below threshold remain in the dormant state.

(C) Within the population, the phenotypic output of noise in ComK expression results in each individual expressing ComK at a different level. High expressing bacteria go on to assume the competent fate, while low expressing bacteria take on the dormant state.

Stochastic fate choices also occur in multicellular organisms, but it is largely unclear how noise controls these decisions. Some preliminary work has been performed in hematopoietic cells, whose fates are influenced by the expression levels of the cell surface receptor Sca-1 (Chang et al., 2008). Noise in the levels of Sca-1 partitions these cells into high-expressing erythroid fates and low-expressing myeloid fates. However, the underlying mechanisms that stabilize these cell fates are not yet known.

Future work may support the exciting possibility that the same general mechanisms guiding stochastic cell fates in bacteria and viruses are conserved in eukaryotic cells.

### **CONCLUSION AND PERSPECTIVES**

Noise in gene expression can be utilized or buffered to control cell fates in both unicellular and multicellular organisms. Recent technological advances are allowing more in-depth studies of gene expression noise in multicellular organisms. Single molecule FISH and the MS2 hairpin system have been applied in several organisms, and the tracking of single RNA polymerases has been established *in vitro* (Lenstra et al., 2016). Future work developing single molecule live imaging to study noise and stochastic cell fates in other multicellular systems will provide important insights into the regulation of these processes in endogenous cellular contexts.

One potential system for future investigation is the Drosophila retina. Precise signaling cues produce a crystalline pattern of ommatidia (Wolff and Ready, 1991; Kumar, 2011). Despite this highly reproducible structure, patterning of R7 photoreceptor subtypes is stochastic. R7 cells randomly express the photopigment Rhodopsin 3 (Rh3) or Rhodopsin 4 (Rh4) (Bell et al., 2007; Johnston and Desplan, 2008, 2010). The decision to express Rh3 vs. Rh4 is controlled by the transcription factor Spineless (Ss), which is stochastically expressed in  $\sim$ 70% of R7s, but the complete mechanism driving the on/off Ss decision is still unknown (Wernet et al., 2006; Johnston et al., 2011; Thanawala et al., 2013). Ss expression is partially controlled at the gene locus itself, which contains two repressing silencers and an activating enhancer (Johnston and Desplan, 2014). The transcriptional repressor Klumpfuss (Klu) also plays a critical role in determining the on/off ratio of Ss expression (Anderson et al., 2017). It is possible that the binary on/off output of Ss is

### REFERENCES

- Anderson, C., Reiss, I., Zhou, C., Cho, A., Siddiqi, H., Mormann, B., et al. (2017). Natural variation in stochastic photoreceptor specification and color preference in *Drosophila*. eLife 6:e29593. doi: 10.7554/eLife.29593
- Bahar Halpern, K., Tanami, S., Landen, S., Chapal, M., Szlak, L., Hutzler, A., et al. (2015). Bursty gene expression in the intact mammalian liver. *Mol. Cell* 58, 147–156. doi: 10.1016/j.molcel.2015.01.027
- Balazsi, G., van Oudenaarden, A., and Collins, J. J. (2011). Cellular decision making and biological noise: from microbes to mammals. Cell 144, 910–925. doi: 10.1016/j.cell.2011.01.030
- Bartman, C. R., Hsu, S. C., Hsiung, C. C., Raj, A., and Blobel, G. A. (2016). Enhancer regulation of transcriptional bursting parameters revealed by forced chromatin looping. *Mol. Cell* 62, 237–247. doi: 10.1016/j.molcel.2016.03.007
- Battich, N., Stoeger, T., and Pelkmans, L. (2015). Control of transcript variability in single mammalian cells. *Cell* 163, 1596–1610. doi: 10.1016/j.cell.2015. 11.018
- Bell, A. C., West, A. G., and Felsenfeld, G. (2001). Insulators and boundaries: versatile regulatory elements in the eukaryotic genome. *Science* 291, 447–450. doi: 10.1126/science.291.5503.447
- Bell, M. L., Earl, J. B., and Britt, S. G. (2007). Two types of *Drosophila* R7 photoreceptor cells are arranged randomly: a model for stochastic cell-fate determination. *J. Comp. Neurol.* 502, 75–85. doi: 10.1002/cne. 21298

achieved through noise in ss RNA levels during development in response to repression from Klu. For example, since transcription is inherently stochastic, and Klu binding to the ss locus will vary between cells, the level of repression in each cell will be noisy. If each cell expresses ss at a different level, this could imply that a particular threshold is necessary for producing the terminal decisions to be on or off by creating a bistable system. These observations raise the intriguing possibility that Ss is regulated similarly to ComK in Bacillus subtilis and Tat in HIV-1, suggesting a general paradigm for stochastic fate choice that is conserved through higher organisms.

Stochastic fate decisions are essential for development, pattern formation, and survival. The biological roles for noise in stochastic fate specification has been well-studied in viruses and bacteria, but its functions in eukaryotes are still being elucidated. With advances in quantitative imaging techniques, work on the roles of transcriptional bursting and noise in multicellular organisms is now accessible. Disruptions of stochastic fate specification mechanisms have been linked to human diseases including vision disorders, autism, anosmia, immunodeficiencies and lymphoma (Nathans, 1999; Morrow et al., 2008; Johnston and Desplan, 2010). Elucidating the roles for gene expression noise in cell fate decisions is therefore crucial for understanding when and how these systems go awry.

### **AUTHOR CONTRIBUTIONS**

EU wrote and edited the manuscript. RJ edited the manuscript.

### **FUNDING**

This study was funded by the National Institutes of Health (Grant No. R01EY025598).

- Bertrand, E., Chartrand, P., Schaefer, M., Shenoy, S. M., Singer, R. H., and Long, R. M. (1998). Localization of ASH1 mRNA particles in living yeast. *Mol. Cell* 2, 437–445. doi: 10.1016/S1097-2765(00)80143-4
- Blake, W. J., Balazsi, G., Kohanski, M. A., Isaacs, F. J., Murphy, K. F., Kuang, Y., et al. (2006). Phenotypic consequences of promoter-mediated transcriptional noise. *Mol. Cell* 24, 853–865. doi: 10.1016/j.molcel.2006.11.003
- Bothma, J. P., Garcia, H. G., Esposito, E., Schlissel, G., Gregor, T., and Levine, M. (2014). Dynamic regulation of eve stripe 2 expression reveals transcriptional bursts in living *Drosophila* embryos. *Proc. Natl. Acad. Sci. U.S.A.* 111, 10598–10603. doi: 10.1073/pnas.1410022111
- Bothma, J. P., Garcia, H. G., Ng, S., Perry, M. W., Gregor, T., and Levine, M. (2015). Enhancer additivity and non-additivity are determined by enhancer strength in the *Drosophila* embryo. *eLife* 4:e07956. doi: 10.7554/eLife.07956
- Buecker, C., and Wysocka, J. (2012). Enhancers as information integration hubs in development: lessons from genomics. *Trends Genet.* 28, 276–284. doi: 10.1016/ j.tig.2012.02.008
- Bulger, M., and Groudine, M. (2011). Functional and mechanistic diversity of distal transcription enhancers. Cell 144, 327–339. doi: 10.1016/j.cell.2011.01.024
- Bury-Mone, S., and Sclavi, B. (2017). Stochasticity of gene expression as a motor of epigenetics in bacteria: from individual to collective behaviors. *Res. Microbiol.* 168, 503–514. doi: 10.1016/j.resmic.2017.03.009
- Cai, L., Friedman, N., and Xie, X. S. (2006). Stochastic protein expression in individual cells at the single molecule level. *Nature* 440, 358–362. doi: 10.1038/ nature04599

- Cairns, B. R. (2009). The logic of chromatin architecture and remodelling at promoters. *Nature* 461, 193–198. doi: 10.1038/nature08450
- Chalancon, G., Ravarani, C. N., Balaji, S., Martinez-Arias, A., Aravind, L., Jothi, R., et al. (2012). Interplay between gene expression noise and regulatory network architecture. *Trends Genet.* 28, 221–232. doi: 10.1016/j.tig.2012.01.006
- Chang, H. H., Hemberg, M., Barahona, M., Ingber, D. E., and Huang, S. (2008). Transcriptome-wide noise controls lineage choice in mammalian progenitor cells. *Nature* 453, 544–547. doi: 10.1038/nature06965
- Chang, S., Johnston, R. J. Jr., Frokjaer-Jensen, C., Lockery, S., and Hobert, O. (2004). MicroRNAs act sequentially and asymmetrically to control chemosensory laterality in the nematode. *Nature* 430, 785–789. doi:10.1038/nature02752
- Chong, S., Chen, C., Ge, H., and Xie, X. S. (2014). Mechanism of transcriptional bursting in bacteria. Cell 158, 314–326. doi: 10.1016/j.cell.2014.05.038
- Cochella, L., Tursun, B., Hsieh, Y. W., Galindo, S., Johnston, R. J., Chuang, C. F., et al. (2014). Two distinct types of neuronal asymmetries are controlled by the *Caenorhabditis elegans* zinc finger transcription factor die-1. *Genes Dev.* 28, 34–43. doi: 10.1101/gad.233643.113
- Dar, R. D., Razooky, B. S., Singh, A., Trimeloni, T. V., McCollum, J. M., Cox, C. D., et al. (2012). Transcriptional burst frequency and burst size are equally modulated across the human genome. *Proc. Natl. Acad. Sci. U.S.A.* 109, 17454–17459. doi: 10.1073/pnas.1213530109
- Dixon, J. R., Gorkin, D. U., and Ren, B. (2016). Chromatin domains: the unit of chromosome organization. Mol. Cell 62, 668–680. doi: 10.1016/j.molcel.2016. 05.018
- Dixon, J. R., Selvaraj, S., Yue, F., Kim, A., Li, Y., Shen, Y., et al. (2012). Topological domains in mammalian genomes identified by analysis of chromatin interactions. *Nature* 485, 376–380. doi: 10.1038/nature11082
- Dorsett, D. (1993). Distance-independent inactivation of an enhancer by the suppressor of Hairy-wing DNA-binding protein of Drosophila. *Genetics* 134, 1135–1144.
- Dowen, J. M., Fan, Z. P., Hnisz, D., Ren, G., Abraham, B. J., Zhang, L. N., et al. (2014). Control of cell identity genes occurs in insulated neighborhoods in mammalian chromosomes. *Cell* 159, 374–387. doi: 10.1016/j.cell.2014.09.030
- Elowitz, M. B., Levine, A. J., Siggia, E. D., and Swain, P. S. (2002). Stochastic gene expression in a single cell. *Science* 297, 1183–1186. doi: 10.1126/science.1070919
- Enver, T., Pera, M., Peterson, C., and Andrews, P. W. (2009). Stem cell states, fates, and the rules of attraction. *Cell Stem Cell* 4, 387–397. doi: 10.1016/j.stem.2009. 04.011
- Fudenberg, G., Imakaev, M., Lu, C., Goloborodko, A., Abdennur, N., and Mirny, L. A. (2016). Formation of chromosomal domains by loop extrusion. *Cell Rep.* 15, 2038–2049. doi: 10.1016/j.celrep.2016.04.085
- Fujita, K., Iwaki, M., and Yanagida, T. (2016). Transcriptional bursting is intrinsically caused by interplay between RNA polymerases on DNA. Nat. Commun. 7:13788. doi: 10.1038/ncomms13788
- Fukaya, T., Lim, B., and Levine, M. (2016). Enhancer control of transcriptional bursting. Cell 166, 358–368. doi: 10.1016/j.cell.2016.05.025
- Garcia, H. G., Tikhonov, M., Lin, A., and Gregor, T. (2013). Quantitative imaging of transcription in living *Drosophila* embryos links polymerase activity to patterning. *Curr. Biol.* 23, 2140–2145. doi: 10.1016/j.cub.2013.08.054
- Geyer, P. K., and Corces, V. G. (1992). DNA position-specific repression of transcription by a Drosophila zinc finger protein. *Genes Dev.* 6, 1865–1873. doi:10.1101/gad.6.10.1865
- Golding, I., Paulsson, J., Zawilski, S. M., and Cox, E. C. (2005). Real-time kinetics of gene activity in individual bacteria. *Cell* 123, 1025–1036. doi: 10.1016/j.cell. 2005.09.031
- Gregor, T., Garcia, H. G., and Little, S. C. (2014). The embryo as a laboratory: quantifying transcription in *Drosophila*. *Trends Genet.* 30, 364–375. doi: 10. 1016/j.tig.2014.06.002
- Gregor, T., Tank, D. W., Wieschaus, E. F., and Bialek, W. (2007). Probing the limits to positional information. Cell 130, 153–164. doi: 10.1016/j.cell.2007.05.025
- Handoko, L., Xu, H., Li, G., Ngan, C. Y., Chew, E., Schnapp, M., et al. (2011). CTCF-mediated functional chromatin interactome in pluripotent cells. *Nat. Genet.* 43, 630–638. doi: 10.1038/ng.857
- Hardy, C. D., and Cozzarelli, N. R. (2005). A genetic selection for supercoiling mutants of *Escherichia coli* reveals proteins implicated in chromosome structure. *Mol. Microbiol.* 57, 1636–1652. doi: 10.1111/j.1365-2958.2005. 04799.x

- Harper, C. V., Finkenstadt, B., Woodcock, D. J., Friedrichsen, S., Semprini, S., Ashall, L., et al. (2011). Dynamic analysis of stochastic transcription cycles. *PLoS Biol.* 9:e1000607. doi: 10.1371/journal.pbio.1000607
- Hendy, O., Campbell, J. Jr., Weissman, J. D., Larson, D. R., and Singer, D. S. (2017).
  Differential context-specific impact of individual core promoter elements on transcriptional dynamics. *Mol. Biol. Cell* 28, 3360–3370. doi: 10.1091/mbc.E17-06-0408
- Hnisz, D., Shrinivas, K., Young, R. A., Chakraborty, A. K., and Sharp, P. A. (2017). A phase separation model for transcriptional control. *Cell* 169, 13–23. doi: 10.1016/j.cell.2017.02.007
- Hobert, O., Johnston, R. J. Jr., and Chang, S. (2002). Left-right asymmetry in the nervous system: the *Caenorhabditis elegans* model. *Nat. Rev. Neurosci.* 3, 629–640. doi: 10.1038/nrn897
- Hong, J. W., Hendrix, D. A., and Levine, M. S. (2008). Shadow enhancers as a source of evolutionary novelty. Science 321:1314. doi: 10.1126/science.1160631
- Hornung, G., Bar-Ziv, R., Rosin, D., Tokuriki, N., Tawfik, D. S., Oren, M., et al. (2012). Noise-mean relationship in mutated promoters. *Genome Res.* 22, 2409–2417. doi: 10.1101/gr.139378.112
- Ing-Simmons, E., Seitan, V. C., Faure, A. J., Flicek, P., Carroll, T., Dekker, J., et al. (2015). Spatial enhancer clustering and regulation of enhancer-proximal genes by cohesin. *Genome Res.* 25, 504–513. doi: 10.1101/gr.184986.114
- Jackson, D. A., and Pombo, A. (1998). Replicon clusters are stable units of chromosome structure: evidence that nuclear organization contributes to the efficient activation and propagation of S phase in human cells. *J. Cell Biol.* 140, 1285–1295. doi: 10.1083/jcb.140.6.1285
- Johnston, R. J. Jr., Chang, S., Etchberger, J. F., Ortiz, C. O., and Hobert, O. (2005). MicroRNAs acting in a double-negative feedback loop to control a neuronal cell fate decision. *Proc. Natl. Acad. Sci. U.S.A.* 102, 12449–12454. doi: 10.1073/pnas. 0505530102
- Johnston, R. J. Jr., Copeland, J. W., Fasnacht, M., Etchberger, J. F., Liu, J., Honig, B., et al. (2006). An unusual Zn-finger/FH2 domain protein controls a left/right asymmetric neuronal fate decision in C. elegans. Development 133, 3317–3328. doi: 10.1242/dev.02494
- Johnston, R. J. Jr., and Desplan, C. (2008). Stochastic neuronal cell fate choices. Curr. Opin. Neurobiol. 18, 20–27. doi: 10.1016/j.conb.2008.04.004
- Johnston, R. J. Jr., and Desplan, C. (2010). Stochastic mechanisms of cell fate specification that yield random or robust outcomes. Annu. Rev. Cell Dev. Biol. 26, 689–719. doi: 10.1146/annurev-cellbio-100109-104113
- Johnston, R. J. Jr., and Desplan, C. (2014). Interchromosomal communication coordinates intrinsically stochastic expression between alleles. *Science* 343, 661–665. doi: 10.1126/science.1243039
- Johnston, R. J. Jr., and Hobert, O. (2005). A novel C. elegans zinc finger transcription factor, lsy-2, required for the cell type-specific expression of the lsy-6 microRNA. Development 132, 5451–5460. doi: 10.1242/dev. 02163
- Johnston, R. J. Jr., Otake, Y., Sood, P., Vogt, N., Behnia, R., Vasiliauskas, D., et al. (2011). Interlocked feedforward loops control cell-type-specific Rhodopsin expression in the *Drosophila* eye. Cell 145, 956–968. doi: 10.1016/j.cell.2011. 05.003
- Johnston, R. J., and Hobert, O. (2003). A microRNA controlling left/right neuronal asymmetry in Caenorhabditis elegans. Nature 426, 845–849. doi: 10.1038/ nature02255
- Kar, G., Kim, J. K., Kolodziejczyk, A. A., Natarajan, K. N., Triglia, E. T., Mifsud, B., et al. (2017). Flipping between Polycomb repressed and active transcriptional states introduces noise in gene expression. *Nat. Commun.* 8:36. doi: 10.1038/s41467-017-00052-2
- Kumar, J. P. (2011). My what big eyes you have: how the Drosophila retina grows. Dev. Neurobiol. 71, 1133–1152. doi: 10.1002/dneu. 20921
- Kumar, J. P. (2012). Building an ommatidium one cell at a time. Dev. Dyn. 241, 136–149. doi: 10.1002/dvdy.23707
- Kumar, N., Singh, A., and Kulkarni, R. V. (2015). Transcriptional bursting in gene expression: analytical results for general stochastic models. *PLoS Comput. Biol.* 11:e1004292. doi: 10.1371/journal.pcbi.1004292
- Larson, D. R., Fritzsch, C., Sun, L., Meng, X., Lawrence, D. S., and Singer,
   R. H. (2013). Direct observation of frequency modulated transcription
   in single cells using light activation. *eLife* 2:e00750. doi: 10.7554/eLife.

- Lenstra, T. L., Rodriguez, J., Chen, H., and Larson, D. R. (2016). Transcription dynamics in living cells. Annu. Rev. Biophys. 45, 25–47. doi: 10.1146/annurevbiophys-062215-010838
- Lettice, L. A., Heaney, S. J., Purdie, L. A., Li, L., de Beer, P., Oostra, B. A., et al. (2003). A long-range Shh enhancer regulates expression in the developing limb and fin and is associated with preaxial polydactyly. *Hum. Mol. Genet.* 12, 1725–1735. doi: 10.1093/hmg/ddg180
- Levine, M. (2010). Transcriptional enhancers in animal development and evolution. Curr. Biol. 20, R754–R763. doi: 10.1016/j.cub.2010.06.070
- Lewis, E. B. (1978). A gene complex controlling segmentation in *Drosophila. Nature* 276, 565–570. doi: 10.1038/276565a0
- Little, S. C., Tikhonov, M., and Gregor, T. (2013). Precise developmental gene expression arises from globally stochastic transcriptional activity. *Cell* 154, 789–800. doi: 10.1016/j.cell.2013.07.025
- Losick, R., and Desplan, C. (2008). Stochasticity and cell fate. Science 320, 65–68. doi: 10.1126/science.1147888
- Ma, H., Samarabandu, J., Devdhar, R. S., Acharya, R., Cheng, P. C., Meng, C., et al. (1998). Spatial and temporal dynamics of DNA replication sites in mammalian cells. *J. Cell Biol.* 143, 1415–1425. doi: 10.1083/jcb.143.6.1415
- Maamar, H., and Dubnau, D. (2005). Bistability in the *Bacillus subtilis* K-state (competence) system requires a positive feedback loop. *Mol. Microbiol.* 56, 615–624. doi: 10.1111/j.1365-2958.2005.04592.x
- Maamar, H., Raj, A., and Dubnau, D. (2007). Noise in gene expression determines cell fate in *Bacillus subtilis*. Science 317, 526–529. doi: 10.1126/science.1140818
- Maduro, M. F. (2010). Cell fate specification in the *C. elegans* embryo. *Dev. Dyn.* 239, 1315–1329. doi: 10.1002/dvdy.22233
- McAdams, H. H., and Arkin, A. (1997). Stochastic mechanisms in gene expression. Proc. Natl. Acad. Sci. U.S.A. 94, 814–819. doi: 10.1073/pnas.94.3.814
- Morrow, E. M., Yoo, S. Y., Flavell, S. W., Kim, T. K., Lin, Y., Hill, R. S., et al. (2008). Identifying autism loci and genes by tracing recent shared ancestry. *Science* 321, 218–223. doi: 10.1126/science.1157657
- Mugler, A., Kittisopikul, M., Hayden, L., Liu, J., Wiggins, C. H., Suel, G. M., et al. (2016). Noise expands the response range of the *Bacillus subtilis* competence circuit. *PLoS Comput. Biol.* 12:e1004793. doi: 10.1371/journal.pcbi.100 4793
- Nasmyth, K. (2011). Cohesin: a catenase with separate entry and exit gates? *Nat. Cell Biol.* 13, 1170–1177. doi: 10.1038/ncb2349
- Nathans, J. (1999). The evolution and physiology of human color vision: insights from molecular genetic studies of visual pigments. *Neuron* 24, 299–312. doi:10.1016/S0896-6273(00)80845-4
- Newlands, S., Levitt, L. K., Robinson, C. S., Karpf, A. B., Hodgson, V. R., Wade, R. P., et al. (1998). Transcription occurs in pulses in muscle fibers. *Genes Dev.* 12, 2748–2758. doi: 10.1101/gad.12.17.2748
- Nicolas, D., Zoller, B., Suter, D. M., and Naef, F. (2018). Modulation of transcriptional burst frequency by histone acetylation. *Proc. Natl. Acad. Sci.* U.S.A. 115, 7153–7158. doi: 10.1073/pnas.1722330115
- Nora, E. P., Lajoie, B. R., Schulz, E. G., Giorgetti, L., Okamoto, I., Servant, N., et al. (2012). Spatial partitioning of the regulatory landscape of the X-inactivation centre. *Nature* 485, 381–385. doi: 10.1038/nature11049
- Ozbudak, E. M., Thattai, M., Kurtser, I., Grossman, A. D., and van Oudenaarden, A. (2002). Regulation of noise in the expression of a single gene. *Nat. Genet.* 31, 69–73. doi: 10.1038/ng869
- Pare, A., Lemons, D., Kosman, D., Beaver, W., Freund, Y., and McGinnis, W. (2009). Visualization of individual Scr mRNAs during *Drosophila* embryogenesis yields evidence for transcriptional bursting. *Curr. Biol.* 19, 2037–2042. doi: 10.1016/j.cub.2009.10.028
- Peccoud, J., and Ycart, B. (1995). Markovian modeling of gene-product synthesis. *Theor. Popul. Biol.* 48, 222–234. doi: 10.1006/tpbi.1995.1027
- Poole, R. J., and Hobert, O. (2006). Early embryonic programming of neuronal left/right asymmetry in *C. elegans. Curr. Biol.* 16, 2279–2292. doi: 10.1016/j.cub. 2006.09.041
- Postow, L., Hardy, C. D., Arsuaga, J., and Cozzarelli, N. R. (2004). Topological domain structure of the *Escherichia coli* chromosome. *Genes Dev.* 18, 1766–1779. doi: 10.1101/gad.1207504
- Racko, D., Benedetti, F., Dorier, J., and Stasiak, A. (2018). Transcription-induced supercoiling as the driving force of chromatin loop extrusion during formation of TADs in interphase chromosomes. *Nucleic Acids Res.* 46, 1648–1660. doi: 10.1093/nar/gkx1123

- Raj, A., Peskin, C. S., Tranchina, D., Vargas, D. Y., and Tyagi, S. (2006). Stochastic mRNA synthesis in mammalian cells. *PLoS Biol.* 4:e309. doi: 10.1371/journal. pbio.0040309
- Raj, A., and van Oudenaarden, A. (2008). Nature, nurture, or chance: stochastic gene expression and its consequences. *Cell* 135, 216–226. doi: 10.1016/j.cell. 2008.09.050
- Raser, J. M., and O'Shea, E. K. (2004). Control of stochasticity in eukaryotic gene expression. Science 304, 1811–1814. doi: 10.1126/science.1098641
- Raser, J. M., and O'Shea, E. K. (2005). Noise in gene expression: origins, consequences, and control. *Science* 309, 2010–2013. doi: 10.1126/science. 1105891
- Ren, G., Jin, W. F., Cui, K. R., Rodrigez, J., Hu, G. Q., Zhang, Z. Y., et al. (2017). CTCF-mediated enhancer-promoter interaction is a critical regulator of cell-to-cell variation of gene expression. *Mol. Cell* 67, 1049–1058.e6. doi: 10.1016/j. molcel.2017.08.026
- Ross, I. L., Browne, C. M., and Hume, D. A. (1994). Transcription of individual genes in eukaryotic cells occurs randomly and infrequently. *Immunol. Cell Biol.* 72, 177–185. doi: 10.1038/icb.1994.26
- Sanborn, A. L., Rao, S. S., Huang, S. C., Durand, N. C., Huntley, M. H., Jewett, A. I., et al. (2015). Chromatin extrusion explains key features of loop and domain formation in wild-type and engineered genomes. *Proc. Natl. Acad. Sci. U.S.A.* 112, E6456–E6465. doi: 10.1073/pnas.151855 2112
- Sarin, S., O'Meara, M. M., Flowers, E. B., Antonio, C., Poole, R. J., Didiano, D., et al. (2007). Genetic screens for *Caenorhabditis elegans* mutants defective in left/right asymmetric neuronal fate specification. *Genetics* 176, 2109–2130. doi: 10.1534/genetics.107.075648
- Schuettengruber, B., Bourbon, H. M., Di Croce, L., and Cavalli, G. (2017). Genome regulation by polycomb and trithorax: 70 years and counting. *Cell* 171, 34–57. doi: 10.1016/j.cell.2017.08.002
- Seitan, V. C., Faure, A. J., Zhan, Y., McCord, R. P., Lajoie, B. R., Ing-Simmons, E., et al. (2013). Cohesin-based chromatin interactions enable regulated gene expression within preexisting architectural compartments. *Genome Res.* 23, 2066–2077. doi: 10.1101/gr.161620.113
- Senecal, A., Munsky, B., Proux, F., Ly, N., Braye, F. E., Zimmer, C., et al. (2014). Transcription factors modulate c-Fos transcriptional bursts. *Cell Rep.* 8, 75–83. doi: 10.1016/j.celrep.2014.05.053
- Sexton, T., Yaffe, E., Kenigsberg, E., Bantignies, F., Leblanc, B., Hoichman, M., et al. (2012). Three-dimensional folding and functional organization principles of the *Drosophila* genome. *Cell* 148, 458–472. doi: 10.1016/j.cell.2012. 01.010
- Singh, A. (2011). Negative feedback through mRNA provides the best control of gene-expression noise. *IEEE Trans. Nanobioscience* 10, 194–200. doi: 10.1109/ TNB.2011.2168826
- Skupsky, R., Burnett, J. C., Foley, J. E., Schaffer, D. V., and Arkin, A. P. (2010). HIV promoter integration site primarily modulates transcriptional burst size rather than frequency. *PLoS Comput. Biol.* 6:e1000952. doi: 10.1371/journal. pcbi 1000952
- Sofueva, S., Yaffe, E., Chan, W. C., Georgopoulou, D., Vietri Rudan, M., Mira-Bontenbal, H., et al. (2013). Cohesin-mediated interactions organize chromosomal domain architecture. *EMBO J.* 32, 3119–3129. doi: 10.1038/ emboj.2013.237
- Spitz, F., and Furlong, E. E. (2012). Transcription factors: from enhancer binding to developmental control. *Nat. Rev. Genet.* 13, 613–626. doi: 10.1038/nrg 3207
- Stigler, J., Camdere, G. O., Koshland, D. E., and Greene, E. C. (2016).
  Single-molecule imaging reveals a collapsed conformational state for DNA-bound cohesin. *Cell Rep.* 15, 988–998. doi: 10.1016/j.celrep.2016. 04.003
- Stoeger, T., Battich, N., and Pelkmans, L. (2016). Passive noise filtering by cellular compartmentalization. Cell 164, 1151–1161. doi: 10.1016/j.cell.2016. 02.005
- Suel, G. M., Garcia-Ojalvo, J., Liberman, L. M., and Elowitz, M. B. (2006). An excitable gene regulatory circuit induces transient cellular differentiation. *Nature* 440, 545–550. doi: 10.1038/nature04588
- Sulston, J. E., Schierenberg, E., White, J. G., and Thomson, J. N. (1983). The embryonic cell lineage of the nematode *Caenorhabditis elegans*. *Dev. Biol.* 100, 64–119. doi: 10.1016/0012-1606(83)90201-4

- Suter, D. M., Molina, N., Gatfield, D., Schneider, K., Schibler, U., and Naef, F. (2011). Mammalian genes are transcribed with widely different bursting kinetics. Science 332, 472–474. doi: 10.1126/science.1198817
- Swain, P. S. (2004). Efficient attenuation of stochasticity in gene expression through post-transcriptional control. J. Mol. Biol. 344, 965–976. doi: 10.1016/j.jmb.2004. 09 073
- Swain, P. S., Elowitz, M. B., and Siggia, E. D. (2002). Intrinsic and extrinsic contributions to stochasticity in gene expression. *Proc. Natl. Acad. Sci. U.S.A.* 99, 12795–12800. doi: 10.1073/pnas.162041399
- Takasuka, N., White, M. R., Wood, C. D., Robertson, W. R., and Davis, J. R. (1998). Dynamic changes in prolactin promoter activation in individual living lactotrophic cells. *Endocrinology* 139, 1361–1368. doi: 10.1210/endo.139.3.5826
- Tantale, K., Mueller, F., Kozulic-Pirher, A., Lesne, A., Victor, J. M., Robert, M. C., et al. (2016). A single-molecule view of transcription reveals convoys of RNA polymerases and multi-scale bursting. *Nat. Commun.* 7:12248. doi: 10.1038/ncomms12248
- Thanawala, S. U., Rister, J., Goldberg, G. W., Zuskov, A., Olesnicky, E. C., Flowers, J. M., et al. (2013). Regional modulation of a stochastically expressed factor determines photoreceptor subtypes in the *Drosophila* retina. *Dev. Cell* 25, 93–105. doi: 10.1016/j.devcel.2013.02.016
- Waddington, C. H. (1957). The Strategy of the Genes: A Discussion of Some Aspects of Theoretical Biology. London: Allen and Unwin.
- Weinberger, L. S., Burnett, J. C., Toettcher, J. E., Arkin, A. P., and Schaffer, D. V. (2005). Stochastic gene expression in a lentiviral positive-feedback loop: HIV-1 Tat fluctuations drive phenotypic diversity. *Cell* 122, 169–182. doi: 10.1016/j. cell.2005.06.006
- Weinberger, L. S., Dar, R. D., and Simpson, M. L. (2008). Transient-mediated fate determination in a transcriptional circuit of HIV. *Nat. Genet.* 40, 466–470. doi: 10.1038/ng.116
- Wernet, M. F., Mazzoni, E. O., Celik, A., Duncan, D. M., Duncan, I., and Desplan, C. (2006). Stochastic spineless expression creates the retinal mosaic for colour vision. *Nature* 440, 174–180. doi: 10.1038/nature04615

- Wolf, L., Silander, O. K., and van Nimwegen, E. (2015). Expression noise facilitates the evolution of gene regulation. eLife 4:e05856. doi: 10.7554/eLife. 05856
- Wolff, T., and Ready, D. F. (1991). The beginning of pattern formation in the Drosophila compound eye: the morphogenetic furrow and the second mitotic wave. *Development* 113, 841–850.
- Yan, J., Anderson, C., Viets, K., Tran, S., Goldberg, G., Small, S., et al. (2017). Regulatory logic driving stable levels of defective proventriculus expression during terminal photoreceptor specification in flies. *Development* 144, 844–855. doi: 10.1242/dev.144030
- Zabidi, M. A., Arnold, C. D., Schernhuber, K., Pagani, M., Rath, M., Frank, O., et al. (2015). Enhancer-core-promoter specificity separates developmental and housekeeping gene regulation. *Nature* 518, 556–559. doi: 10.1038/nature 13994
- Zoller, B., Little, S. C., and Gregor, T. (2018). Diverse spatial expression patterns emerge from unified kinetics of transcriptional bursting. *Cell* 175, 835–847.e5. doi: 10.1016/j.cell.2018.09.056
- Zuin, J., Dixon, J. R., van der Reijden, M. I., Ye, Z., Kolovos, P., Brouwer, R. W., et al. (2014). Cohesin and CTCF differentially affect chromatin architecture and gene expression in human cells. *Proc. Natl. Acad. Sci. U.S.A.* 111, 996–1001. doi: 10.1073/pnas.1317788111

**Conflict of Interest Statement:** The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

Copyright © 2018 Urban and Johnston. This is an open-access article distributed under the terms of the Creative Commons Attribution License (CC BY). The use, distribution or reproduction in other forums is permitted, provided the original author(s) and the copyright owner(s) are credited and that the original publication in this journal is cited, in accordance with accepted academic practice. No use, distribution or reproduction is permitted which does not comply with these terms.