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A Protocol for an economic evaluation of a polypill in patients with established or at high risk of cardiovascular disease in a UK NHS setting: RUPEE (NHS) study

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3 A Protocol for an economic evaluation of a polypill in patients with established or at high
4 risk of cardiovascular disease in a UK NHS setting: *RUPEE (NHS) study*
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Abstract

Introduction: The 'Use of a Multi-drug Pill in Reducing cardiovascular Events' (UMPIRE) trial was a randomised controlled clinical trial evaluating the impact of a polypill strategy on adherence to indicated medication in a population with established cardiovascular disease of or at high risk thereof. The aim of RUPEE-NHS is to estimate the potential health economic impact of a polypill strategy for CVD prevention within the NHS using UMPIRE trial and other relevant data. This paper describes the design of a modelled economic evaluation of the impact of increased adherence to the polypill versus usual care amongst the UK UMPIRE participants.

Methods and Analysis: As recommended by ISPOR-SMDM modelling guidelines a review of published CVD models was undertaken to identify the most appropriate modelling approach and structure. The review was carried out in the electronic databases, MEDLINE and EMBASE. 40 CVD models were identified from 57 studies, the majority of economic models were health state transition cohort models and individual level simulation models. The findings were discussed with clinical experts to confirm the approach and structure. An individual simulation approach was identified as the most suitable method to capture the heterogeneity in population CVD risk. RUPEE-NHS will use UMPIRE trial data on adherence to estimate the long term cost-effectiveness of the polypill strategy.

Dissemination: The evaluation findings will be presented in open access scientific and healthcare policy journals and at national and international conferences. We will also present findings to NHS policy makers and pharmaceutical companies.

Strengths and Limitations

This paper provides a clear outline of how a model for an economic evaluation is developed.

Providing an outline of the model structure which includes details on the underlying epidemiology and data inputs will add transparency to the findings of the RUPEE-NHS study

Though the model has been designed to include all major adverse and beneficial effects of treatment, the model structure will not include every potential treatment effect, for example the benefits of treatment on Alzheimer's disease will not be included.

INTRODUCTION

Adherence to recommended preventive medication regimes (1,2) in people at high risk of cardiovascular disease (CVD) is low, even in high income countries. (3) Poor adherence is associated with greater deterioration in health status and increased health care costs (4) and studies have shown that improved adherence to medication is associated with clinical benefits.(5) CVD preventive medication typically involves several drugs and adherence is inversely proportional to the number of prescriptions. Furthermore, physician inertia and patient resistance present barriers to initiating or restarting full recommended therapy. A single pill that includes several indicated drugs (a “polypill”), may improve long-term adherence by addressing these issues. If the polypill is priced lower than the price of the pills bought separately, it will also make it more affordable. (6,7) The UMPIRE (Use of a Multidrug Pill in Reducing Cardiovascular events) clinical trial was set up to evaluate the polypill in patients with or at high risk of CVD.

The UMPIRE trial randomised 2004 participants with established CVD (prior CVD event such as stroke or myocardial infarction) or at high risk of CVD (defined as a 5 year risk of >15%) based in India, England, Ireland and The Netherlands to either the polypill or usual care. The primary outcome of the trial was adherence to indicated treatments (statin, aspirin and two blood pressure lowering drugs), measured as self-reported current use of antiplatelet, statin and ≥ 2 blood pressure lowering therapies for at least 4 days in the week preceding visits (baseline and end of trial visits). Other outcomes included systolic blood pressure (SBP) and low-density lipoprotein cholesterol (LDL-C). The trial found that the use of a polypill strategy resulted in greater adherence to treatment at 15 months and significant improvements in SBP and LDL-C. Detailed results and a description of the UMPIRE trial protocol are available. (8, 9)

UMPIRE collected data on resource use and self-reported health related quality of life using the EQ-5D. In order to estimate the long term costs and health outcomes associated with the polypill strategy an economic model is required. Due to differences in the patient population, care pathway and health care costs, separate analyses are needed for the four participating countries.

The analysis of the UMPIRE English trial data, (Researching the UMPIRE Processes for Economic Evaluation in the National Health Service (RUPEE-NHS)), aims to estimate the cost-effectiveness of the polypill strategy compared to conventional multi-drug therapy for the prevention of established cardiovascular disease in English NHS patients with or at high risk of CVD. The RUPEE (NHS) study

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3 will use UMPIRE English trial data on adherence to the polypill and will develop an economic model
4 to estimate cost effectiveness.
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8 The aim of this paper is to detail the modelling plan for the RUPEE (NHS) study.
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10 11 **METHODS**

12 **Model design process**

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14 An economic model has been described as a mathematical framework that represents reality at an
15 adequate level of detail to inform clinical or policy decisions. (10) Guidelines on modelling produced
16 by the International Society for Pharmacoeconomics and Outcomes Research (ISPOR) and the
17 Society for Medical Decision Making (SMDM) joint taskforce recommend that it is best practice to
18 carry out a conceptualisation process prior to programming the economic model. This process has
19 two distinct components: specification of the study question and economic model. (11)
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22 **Specification of the study question**

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24 The first component informs choices about how to structure the economic model and parameters.
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27 The RUPEE (NHS) study aims to evaluate two different treatment strategies in a population with or
28 at high risk of CVD. The population for the economic model is defined by the inclusion criteria of
29 the UMPIRE trial. (9) The inclusion criteria are listed below:
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- 32 • Aged ≥ 18 years and
- 33 • High CVD risk defined as either established atherothrombotic CVD (history of coronary heart
34 disease (CHD), ischaemic cerebrovascular disease, or peripheral arterial disease (PAD)) or a 5
35 year risk of $\geq 15\%$ calculated using the Framingham risk equation
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44 The economic model will evaluate the polypill strategy compared to usual medication. In the
45 UMPIRE trial, participants assigned to the polypill received one of 2 versions: version 1 contained
46 aspirin 75mg, simvastatin 40mg, lisinopril 10mg and atenolol 50mg, and version 2 contained the
47 same ingredients but substituted hydrochlorothiazide 12.5mg for atenolol 50mg. Participants
48 assigned to usual care continued taking medications as prescribed by their general practitioner (GP).
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53 The RUPEE (NHS) study will follow guidelines for modelling health technologies as recommended by
54 the National Institute for Health and Care Excellence. (NICE) (12) Therefore a NHS and Personal
55 Social Services (PSS) perspective will be adopted to measure health service resource use and health
56 related quality of life will be measured by quality adjusted life years (QALYs) obtained using the EQ-
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3 5D. As per the NICE guidelines, costs and QALYS will be discounted at a rate of 3.5% per year. (12)
4 The time horizon reflected in the economic model will be lifetime to represent the chronic nature of
5 CVD.
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8 Conceptualisation of the economic model

9 The second component of the conceptualisation process involves defining the economic model.
10 There are two steps to this approach. The first step is to identify the appropriate modelling
11 approach. The modelling approach defines the analytical framework of the economic model.
12 Different types of analytical frameworks have been used to represent CVD including decision trees,
13 state transition models, compartmental models, individual simulation models and hybrid models
14 which often combine elements from different frameworks. (13-17)
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21 The second step determines the underlying structure of the analytical framework, which will
22 represent the disease and care pathway. The modelling approach needs to reflect: 1) CVD disease
23 and care pathway for this population; 2) the beneficial and adverse effect of treatment (polypill or
24 usual care); 3) the impact of increased adherence to treatment on health outcomes.
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29 The guidelines produced by ISPOR-SMDM on modelling recommend that existing models addressing
30 related problems should be reviewed as this approach can help identify both the modelling
31 approach and underlying structure.(11) To inform the RUPEE (NHS) economic model, we carried out
32 a review of published models evaluating interventions for CVD.
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38 Review of published CVD economic models

39 The purpose of the literature review was to identify the appropriate analytical framework to
40 represent the decision problem. The literature review also aimed to inform the underlying model
41 structure: disease and care pathway.
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45 *Search strategy*

46 The search strategy was conducted using the NHS Economic Evaluation Database (NHS EED), the
47 NIHR Health Technology Assessment (HTA) monograph series and the NICE guidelines website. The
48 search terms used included 'cardiovascular disease', 'coronary heart disease', 'stroke', 'myocardial
49 infarction', 'angina' and 'peripheral artery disease'. Studies were excluded from the review if they
50 did not discuss the development or review of an economic model; if no disease states for
51 cardiovascular disease were included in the model; if the focus of the study was a diagnostic test or
52 surgical intervention where the economic model used a time frame of <10 years. Studies were not
53 excluded on the basis of intervention (drug treatment or lifestyle intervention) or on the basis of
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3 date published or language. We developed a data extraction form which included fields on model
4 purpose, structure, health states and events, transparency and validation. We did not collate
5 information about the findings of the model as the objective of the review was to identify alternative
6 model frameworks and methods used to represent CVD.
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11 An initial general literature search identified a 2006 systematic review of CHD policy models by Unal
12 *et al.* which was updated in 2008 by Capewell *et al.* and expanded to include stroke models. (17,18)
13 The review by Capewell *et al.* identified seven 'notable' CHD models (of which six had been
14 identified in the previous review by Unal *et al.*), nine stroke models and several models that were
15 currently in development at the time of publication. We reviewed the notable models and models in
16 development identified by Capewell *et al.* Citation searching of both systematic reviews was carried
17 out to identify other models published since 2008.
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22 *Review findings*

23 Overall 57 studies were identified which reported on 40 CVD models. Figure 1 presents the
24 flowchart for the search strategy.
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28 The search found several studies which reported on the same model, for example the IMPACT CHD
29 model developed by Capewell *et al.* was used in analyses of CVD in other populations. (19) In some
30 cases, a model was adapted for different analyses, such as the Sheffield model which was developed
31 to evaluate statin therapy and was then adapted for use in the development of the NICE guidelines
32 for lipid modification. (1,20) The Sheffield model was also partially used in a whole population
33 modelling study by Barton *et al.* (13)
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38 Further details on the review can be found in the supplementary appendix. The appendix includes a
39 list of the reviewed models (see Table 1 supplementary appendix), an example of the data extraction
40 form and an example of an illustration and details of one of the reviewed models (see Figure 1
41 supplementary appendix). Schematic illustrations of several models were used in discussions with
42 clinical experts about the different types of modelling approaches
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48 **[Figure 1- Flowchart for search strategy for CVD models]**

49 **Modelling approach**

50 The search identified that the two most commonly used modelling approaches were health state
51 transition cohort models and individual-level simulation models. Both approaches were critically
52 assessed to determine their suitability to capture the disease and care pathway.
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3 A cohort model can be defined as any model which estimates the outcomes for a group of patients,
4 whereas with a patient level simulation, outcomes are evaluated at the individual level. Therefore,
5 one of the main differences between the two approaches is how they estimate costs and QALYs:
6 cohort models estimate expected costs and QALYs for the modelled population as a whole, whereas
7 individual level simulation models estimate cost and QALYs for each individual and the average is
8 taken across the sample.
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14 With a health state transition cohort model, the population progresses through a set of mutually
15 exclusive health states at regular intervals called cycles, determined by a predefined transition
16 matrix. Health state transition cohort models are also commonly called Markov models. However,
17 such models are only Markovian when they display the Markovian 'memoryless' property where the
18 progression of the patient through the model is only dependent on the current state in which the
19 patient resides and not on anything that happened before they entered that health state. It is also
20 possible to model at the individual level using a state transition model by sampling probabilities for
21 each individual patient to experience a particular transition in each model cycle. (21)
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29 Both model approaches can use a discrete time approach: with this approach the model cycle length
30 will be defined in advance. The cohort or individual progress through health states or events which
31 represent the disease pathway and only one event may occur within each cycle length. Costs and
32 QALYs are updated once per cycle. Alternatively, individual level simulation models are often set up
33 as discrete event simulations (DES). With a DES approach, an event can occur at any time point, for
34 example, an event could occur at three months, one year and twenty years. As an event occurs,
35 costs and QALYs are recorded and updated for each individual.
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42 A health state transition model was used to develop NICE guidance for lipid modification treatment.
43 (1) The limitation of this approach is that it may be unable to capture the underlying heterogeneity
44 in the population. Individual CVD risk can be estimated using CVD risk algorithms such as QRISK2
45 which use a range of patient characteristics such as age, sex, ethnicity, systolic blood pressure and
46 body mass index to estimate a 10 year CVD risk.(22) To capture this complexity in a health state
47 transition model would require the construction of a large number of subgroups to reflect different
48 subsets of patient characteristics and the variation in CVD risk in the population. This could become
49 impractical to model. It also has the disadvantage that accuracy could be lost by using
50 representative values for subgroups. An individual simulation model structure may be more
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3 appropriate to model the level of detail required to estimate CVD outcomes reflective of those in the
4 population.
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8 The Markovian memoryless property means that data on individual patients' history is not
9 retained as they progress through the model. Accounting for individual patient history in a
10 Markov model would require multiplying the number of health states to an infeasible level
11 where the model would become too complex and impracticable to run.
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16 To accurately identify the effectiveness of each treatment strategy in a population with or at high
17 risk of CVD, an individual simulation model was deemed the most appropriate for the RUPEE (NHS)
18 study to reflect the heterogeneity in the population which impacts on the risk of a CVD event and
19 subsequent costs and outcomes. The individual simulation model will use a discrete event approach
20 to handle time.(21)
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24 25 26 Model structure

27 The findings of the review were discussed with clinical experts to confirm the health events and the
28 methods used to model the progression of persons through the disease pathway.
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31 *Model events (CVD, diabetes and adverse events)*

32 The most commonly included types of CVD events in the reviewed models were CHD (angina and
33 myocardial infarction), cerebrovascular events (transient ischaemic attack (TIA) and stroke) and
34 peripheral arterial disease (PAD). It was decided that the CVD events relevant for the current model
35 would reflect those most commonly included in prior such models. PAD will not be included as a
36 CVD event in the model as there is less likely to be a definable acute PAD event compared to other
37 CVD events such as MI and stroke. We will assume that patients can experience more than one CVD
38 event in their lifetime. The risk of CVD will also be assumed to change with age in the model.
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45 Diabetes is a risk factor for CVD with a substantial cost and impact on health related quality of life,
46 therefore diabetes will be included as a comorbidity in the model. The risk of new onset diabetes
47 will be estimated using the QDiabetes risk algorithm.(23)
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51 Adverse effects from treatment will include an increase in the risk of new onset diabetes resulting
52 from treatment with statins and antihypertensive drugs. (24-27) The risk of a persistent cough
53 resulting from treatment with angiotensin-converting-enzyme inhibitors (ACE inhibitors) will be
54 included as an event. The probability of a cough resulting from treatment will be sourced from
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3 meta-analyses of randomised controlled trial (RCT) data for ACE inhibitors. As aspirin use is
4 associated with an increased risk of gastrointestinal bleeding,(28,29) an increased risk of
5 gastrointestinal bleeding from treatment with aspirin will be included.
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9 Renal impairment will not be included in the model as an adverse effect of ACE inhibitors. Whilst
10 ACE inhibitors may cause an acute rise in serum creatinine in a few patients with renal artery
11 stenosis and more generally cause a slight short term increase in creatinine levels, the effects are
12 complex and there may be a net improvement in renal function overall in a treated population. The
13 rate of falls and fractures will be estimated not to alter, given the evidence from randomised trials of
14 blood pressure lowering agents, although this is an area of debate with regard to patients with
15 higher levels of frailty. (30, 31)
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22 Other adverse effects from statin treatment such as liver dysfunction and myopathy will not be
23 included in the model as these cases are rare and are assumed to have a minimal impact on
24 outcomes. (1) (32)
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29 Treatment with antihypertensives is associated with a reduction in heart failure, therefore this will
30 be included as an outcome in the model. (33) Other outcomes of treatment are likely but will not
31 be included – for example a reduction in cancer with aspirin use of more than 5 years. (34)
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36 *Progression of individuals through model*

37 The progression of persons through the disease pathway differs depending on the modelling
38 approach: health state transition models such as the Markov model developed for NICE guidelines
39 on lipid modification use a predefined transition matrix to determine progression through the CVD
40 health states.(35) Alternatively, simulation models can use risk algorithms to estimate the
41 probability of CVD events or new onset diabetes. The NICE guidelines for lipid modification
42 recommend the use of QRISK2, which is a risk algorithm derived to estimate primary CVD risk in UK
43 populations. (1,22) The QRISK2 risk algorithm predicts the risk of a 10 year CHD event (angina, MI)
44 or a cerebrovascular event (TIA, stroke). It does not include the risk of PAD. An alternative CVD risk
45 algorithm is the Framingham equation;(36), however, a validation study comparing QRISK2 and
46 Framingham found that QRISK2 is better calibrated to a UK population.(37) The RUPEE (NHS) model
47 will therefore use the QRISK2 risk algorithm.
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RUPEE (NHS) economic model

Figure 2 depicts the flowchart of the RUPEE (NHS) model structure. The oval shapes represent data inputs to the model, whereas the rectangular shapes represent processes.

[Figure 2 - Flowchart of RUPEE (NHS) model structure]

Model description

In the RUPEE (NHS) model costs and QALYs are recorded for each individual and an average cost and QALY for the simulated population are estimated. The RUPEE (NHS) model will be run twice, once to simulate costs and QALYs under usual care and once to simulate costs and QALYs under the polypill scenario (polypill scenario will include polypill version 1 and version 2). Individuals representing the UMPIRE trial inclusion criteria will enter the model (label 1 in Figure 2), and their baseline risk of a CVD event and onset diabetes will be estimated using the QRISK2 CVD risk algorithm and QDiabetes algorithm (label 2 in Figure 2) respectively. For each individual, whether or not they are adherent to medication will be simulated using Monte Carlo simulation based on the probability of adherence in usual care (label 3a in Figure 2). If the individual is simulated to be adherent to medication their risk of a CVD event will be modified by a treatment effect (label 4 in Figure 2). In the polypill scenario of the model, the probability of adherence will be further modified by the relative risk of adherence to medication. The relative risk of adherence to medication will be sourced from the UMPIRE trial data (label 3b in Figure 2). Individuals may experience a CVD event or onset of diabetes based on their estimated CVD and diabetes risk, which will be estimated using the QRISK2 and QDiabetes algorithms. Individuals may also experience an adverse reaction to medication (if adherent) including gastrointestinal bleeding, early onset of diabetes and a persistent cough. Costs and QALYs will be recorded for each event (including adverse events). Individuals can experience more than one event (model run for lifetime horizon) and patient characteristics such as age and history of previous events, such as a stroke or new onset diabetes, are updated during the model run, with an ensuing reflective increase in the risk of an event.

Input parameters

Each point in the flowchart is labelled and a description of the process or data requirement label is described below. Table 1 provides further details on data input parameters for the RUPEE-NHS model and potential sources of data.

1. Population Dataset

We will use the 2011 Health Survey for England (HSE) as a population dataset for the economic model. The HSE is a cross sectional survey which contains anonymised information on a representative sample of the population. The 2011 HSE dataset collected information on CVD,

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3 including individual CVD events and medication history. The dataset also contains information on
4 demographic and socio-economic characteristics and health related data such as body mass index
5 (BMI), SBP and LDL-C and history of CVD events. These data are required in order to estimate
6 individual baseline risks of CVD and diabetes in the model.
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10 11 **2. Calculation of baseline risks of events without treatment**

12 Baseline risks for CVD for each sampled individual will be calculated using published risk algorithms.
13 As per recent NICE guidance for lipid modification, we will use the recommended algorithm for CVD
14 risk, QRISK2. (1,22) The algorithm was derived using QRESEARCH, a large database derived from
15 the pseudonymised health records of over 13 million patients registered with a general practitioner
16 in the UK. If an individual has established CVD (previously experienced a CVD event), we will
17 estimate a secondary CVD risk using the REACH algorithm. (38) A baseline risk for the onset of
18 diabetes will be estimated using the QDiabetes algorithm. (23)
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25 26 **3a. Simulating adherence to treatment under usual care**

27 The RUPEE study will evaluate the effect of adherence to medication on long term costs and health
28 outcomes measured using quality adjusted life years (QALYS). The average rates of adherence in
29 clinical trials can be higher than in actual practice (4) as seen in the UMPIRE clinical trial population
30 which had an atypically high baseline adherence rate. Instead, adherence rates to medication
31 (antihypertensives, statins and aspirin) under a usual care setting will be sourced from the 2011 HSE
32 dataset. Participants in the 2011 HSE self-reported all the prescribed medications they had taken in
33 the last 7 days. This was coded in the HSE dataset using the British National Formulary (BNF)
34 classifications codes. Using this data, we are able to identify the medication patients were
35 prescribed and identify whether or not they were taking the prescribed medication in the last week.
36 This will reflect adherence to medication in a usual care population. The data will be used to
37 estimate the probability of each person being adherent or not to medication. Individual
38 characteristics will be used as predictors of adherence; the characteristics will be chosen by referring
39 to studies which have assessed predictors of adherence in persons taking treatment for CVD. (39,
40 40) A generalized linear mixed regression model will be used to estimate the probability of
41 adherence to medication for each individual. The probability of persistence with medication will not
42 be assumed to be constant, and the model will include a probability of ceasing medication over time.
43 The probability of medication cessation will be sourced from published literature on adherence.
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55 56 **3b. Estimate relative risks of adherence to medication**

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3 We will estimate the relative risks of adherence to medication, using a generalised linear mixed
4 regression model which will be applied to the UMPIRE trial dataset (UK dataset). In the polypill
5 scenario in the model, the probability of being adherent to medication will be further modified by
6 the relative risks.
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10 11 **4. Adjust risk of events for treatment**

12 We will source data on the treatment effects of statins, antihypertensives and aspirin from meta-
13 analyses of intention-to-treat RCTs. Intention-to-treat analyses account for non-adherence in their
14 findings, and therefore underestimate the impact of treatment on event risk. To overcome this, we
15 will carry out sensitivity analyses to test the impact of adjusting for adherence within the trial. The
16 risk of a CVD event will be adjusted by the relative risk of treatment with statins, antihypertensives
17 and aspirin, based on the medication(s) the person is taking and whether or not they are adherent to
18 medication.
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24 25 **5. Simulation of events**

26 Individuals in the model can experience a CVD event at a rate governed by their calculated baseline
27 risk (estimated by the QRISK2 or REACH algorithms) and adjusted for treatment effects if they have
28 been simulated as adherent to treatment. CVD events will be categorised as a TIA, stroke, MI or
29 angina. The relative incidence of each CVD event will be determined using published incidence
30 data.⁽⁴¹⁾ Similarly, the risk of new onset diabetes will be calculated using the QDiabetes algorithm.
31 We will simulate the incidence of adverse events as a result of treatment: new onset diabetes and
32 gastrointestinal bleeds. Data on the probability of an adverse event will be sourced from meta-
33 analyses of randomised controlled trials for the relevant drugs. Mortality risk will be modelled as
34 mortality from stroke and MI and other cause mortality. Data on other cause mortality will be
35 estimated using national life tables for England and Wales. ⁽⁴²⁾
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44 45 **6. Assign cost and quality of life values**

46 Costs and QALYs associated with each individual's simulated lifetime profile of CVD and related care
47 will be estimated. Costs and quality adjusted life years (QALYs) will accrue for each person to reflect
48 events, such as a stroke or new onset diabetes. Costs and utility values for health events will be
49 sourced from published studies including the NICE guidelines for lipid modification and
50 hypertension. ^(1,35,43) Costs of medication will be sourced from the NHS National Drug Tariff.⁽⁴⁴⁾
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7. Change in age, treatment, CVD status and type II diabetes status

The simulation model will run for each individual for lifetime duration (death or maximum age of 100 years) and patient characteristics will be updated after each event or every 10 years (depending on which event occurs first). A 10 year update is used as the QRISK2 algorithm returns a 10 year CVD risk.

For peer review only

Table 1) Input parameters

Model inputs	Source
1. Individual dataset	
Population dataset	Initial patient characteristics (see Figure 2) for cohort of patients drawn from a representative national sample: Health Survey for England (HSE) dataset 2011. The dataset will include patients who meet the entry criteria for the UMPIRE trial.
2. Calculation of baseline risks	
Risk calculators	Risk of first CVD event and onset of type 2 diabetes estimated for individuals using QRISK2 and QDiabetes.(23, 24) <i>QRISK2: 10 year CVD risk (CVD outcomes defined as angina, MI, TIA and stroke)</i> <i>QDiabetes: risk of acquiring type 2 diabetes over 10 year time period</i> Risks for subsequent CVD events estimated for individuals using the REACH algorithm. (38) <i>CVD outcomes defined as cardiovascular death (includes fatal stroke and MI), non-fatal MI, non-fatal stroke, and cardiovascular hospitalisation (includes hospitalisation for unstable angina and TIA)</i>
Risk of heart failure	Baseline risk per age using incidence rates in Cowie et al. (1998) (45)
Relative incidence of CVD events (TIA, stroke, angina, MI)	OXVASC cohort study, Rothwell <i>et al.</i> 2005. 91,106 individuals presenting with an acute vascular event in Oxfordshire, UK in 2002-5. (41)
3. Adherence to medication	
Probability of adherence to treatment with usual care	Estimates from HSE 2011 dataset on adherence to relevant drugs (statins, antihypertensives, aspirin)
Relative risk of adherence: polypill versus usual care	Estimate the probability of adherence to ≥ 2 antihypertensives, statin or anti-platelet for at least four days in the preceding week for polypill group versus usual care by applying a binomial regression to the UMPIRE dataset.
4. Treatment effects of medication (antihypertensives, statin, anti-platelet)	
Relative risk of CVD with treatment versus no treatment	For base case analysis, conventional meta-analysis of ITT RCT data will be used from – <ul style="list-style-type: none"> ▪ Cholesterol Treatment Trialists' Collaboration ▪ Blood Pressure Lowering Treatment Trialists' Collaboration ▪ Antithrombotic Trialists' Collaboration ▪ Law and Wald (46) Sensitivity analysis: Test impact of adjusting for adherence within trials
5. Other treatment outcomes (beneficial events and adverse events) and mortality rates	
Adverse Events	
Incident type 2 diabetes	Relative risk of diabetes from statins/antihypertensives from meta-analyses of RCTs
GI bleeding	Relative risk of bleeding resulting from aspirin using estimates from meta-analyses of RCTs
Cough	Placebo-adjusted relative risk of cough resulting from ACE inhibitors using estimate from meta-analyses of RCTs

Reduction in heart failure	Relative risk reduction in heart failure from antihypertensives (33)
Mortality	
<i>Stroke case fatality (60 day)</i>	
Age <75	Estimate proportion of strokes that are fatal (with risks increasing with age). Estimate using the BHF Compendium of health statistics 2012, which has data from a record linkage study for England 2010.
Age > 75+	
<i>MI case fatality (30 day)</i>	
Age 30-54	Proportion of MI's that are fatal. Estimate using Oxford Record Linkage pill study. (47) National population based study, including all individuals admitted to hospital or who died suddenly from acute MI in 2010. Age was strongest predictive factor for 30-day case fatality.
Age 55-64	
Age 65-74	
Age 75-84	
Age 85+	
<i>Death from other causes</i>	Estimated from national life tables (Office for National Statistics, England)(42)
6. Costs (medication, monitoring costs, health events)	
<i>Drug costs (£ per year)</i>	
Statins	National Health Service (NHS) Electronic Drug Tariff (44)
AHT drugs	
Aspirin	
Polypill	Assumed to be aggregate cost of each drug in the combined pill
<i>Yearly monitoring costs while on medication</i>	
Primary care nurse (£ per hour)	Use NICE Quality Outcomes Framework to identify recommended management while on treatment (statins, antihypertensives, antiplatelet). A cost for stopping medication will also be applied (e.g. 2 GP visits, tests as recommended in NICE clinical guidelines 181) (1)
GP cost (£ per hour)	
Lipid test (£)	
Liver transaminase test	Costs sourced from Personal Social Services Research Unit Costs and NICE clinical guidelines 181
Blood tests	
<i>Costs of health states and adverse events</i>	
Stroke	Luengo-Fernandez <i>et al.</i> 2012 (48)
TIA	"
MI	NICE lipids guideline 181 (1)
Angina	"
PAD	"
Diabetes	"
GI bleeding	"
Cough (from ACE inhibitor use)	NICE Hypertension guidelines 127 (43)
7. Health Related Quality of life	
Stroke	Derived from Health Survey from England (HSE) dataset
TIA	
MI	
Angina	
PAD	
GI bleeding	
Diabetes	
Cough	

CVD: cardiovascular disease, TIA: transient ischaemic attack, MI: myocardial infarction; ITT: Intention to treat, RCT: randomised controlled trial, AHT: antihypertensives, UMPIRE: Use of a Multidrug Pill in

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3 *Reducing Cardiovascular Events, NICE: National Institute for Health and Care Excellence, GP:general*
4 *practitioner*
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Analysis

The simulation model will run for a sufficient number of iterations to provide stable results. Uncertainty in the model parameters will be examined using a probabilistic sensitivity analysis (PSA) which will reflect uncertainty over the values of the model inputs. Non-parametric bootstrapping of HSE data will be carried out to examine the uncertainty related to the sampling. For each PSA iteration, one non-parametric bootstrap sample will be drawn from the HSE dataset (by random sample with replacement of individuals in the dataset). An incremental analysis will be conducted and incremental cost-effectiveness ratios (ICERs) and net benefit statistics will be estimated. We will also carry out a number of sensitivity analyses to test the impact of varying uncertain parameters in the model. This will include an analysis testing the impact of varying the polypill cost.

Validation

The model will be internally and externally validated. A checklist produced by the RUPEE steering group based on current published guidelines for checking models will be used, to ensure the programmed model behaves as expected according to the theoretical model. (21, 49) The checklist includes tips for model developers, for example on the use of sensitivity analyses to test that the model is operating correctly, and re-programming complicated sections of code in another language. The model will also be reviewed and tested by an experienced modeller. The model results will be compared with real-world observations or the results of other models.

Dissemination of results

The findings of the economic evaluation will be presented to scientific and health care policy audiences in open access journals and at national and international conferences. We will also present findings to NHS policy makers and pharmaceutical companies.

DISCUSSION

Medication adherence is important for disease management, and benefits of increased adherence to preventative medication for CVD include improved clinical outcomes. (5) The UMPIRE clinical trial was conducted to evaluate the effect of a polypill strategy compared to usual care on adherence. It showed that the polypill strategy significantly augmented adherence and this was reflected by improvements in SBP and LDL-C. (8) Whether or not this impact remains in the long term cannot be determined from the trial data alone. The RUPEE (NHS) study is being conducted to evaluate the long term impact of a polypill strategy; in particular, the analysis will evaluate the long term impact of increased adherence on outcomes. An economic model is being developed to estimate the long term costs and QALYs associated with implementing a polypill strategy in the NHS compared to usual

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3 care. This analysis will represent the first comprehensive cost effectiveness analysis using directly
4 applicable clinical trial data.
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8 This paper outlines the process behind the design of the economic model. We carried out a review
9 of published CVD models to identify a modelling approach that would suit the health care decision:
10 use of a polypill versus usual care in a population with or at high risk of CVD. We identified an
11 individual simulation model as the most appropriate approach as it allows the heterogeneity in the
12 population to be adequately reflected. The model will use validated disease risk algorithms to
13 estimate the probability of an individual experiencing a CVD event or the onset of diabetes.
14 Individuals can also experience an increased risk of an adverse event (diabetes, cough and
15 gastrointestinal bleeding) from treatment. The risk of a CVD event will be reduced if the individual
16 is adherent to treatment. We will simulate adherence to treatment using data from the HSE 2011
17 dataset. The probability of adherence in the polypill scenario will be further modified by the relative
18 risks of adherence to medication which will be sourced from the UMPIRE trial data for the English
19 population. Costs and QALYs will be estimated for each individual and aggregated across the sample
20 population (based on the HSE 2011 dataset).
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30 The RUPEE (NHS) model will have a number of advantages over existing models constructed to
31 evaluate a CVD polypill. (50-52) One advantage is the use of an individual simulation model which
32 will allow us to capture the heterogeneity in the variation in CVD risk in the UK population unlike
33 other models which use Markov type transition state models. Another is that we will extrapolate
34 data on adherence to medication from a nationally representative population dataset (Health Survey
35 for England) which will allow us to simulate adherence per individual rather than assuming a
36 constant adherence across our population. We will also allow for adverse events from treatment
37 and treatment cessation, therefore more accurately reflecting clinical practice.
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45 It would be preferable to use per protocol treatment effectiveness data in our analysis as intention-
46 to-treat data already accounts for adherence (people switching and ceasing medication during the
47 trial period). However, per protocol data is difficult to obtain for all drugs, therefore we will use the
48 ITT treatment effect data and carry out sensitivity analyses to test the impact.
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53 The introduction of a CVD preventive polypill strategy will simplify pill taking for patients potentially
54 leading to greater adherence and better health outcomes. This analysis will provide information on
55 the cost-effectiveness of the polypill in a NHS setting.
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5 **Contributors:** CC carried out the literature review and drafted the manuscript. JL, ST, HMD, HW and
6 NP, AR and SJ contributed to the development of the protocol. JL provided input on the health
7 economics model. HMD provided statistical advice and ST, HW and NP contributed clinical advice.
8 AR and SJ peer reviewed draft manuscripts and contributed to the final version of the protocol. All
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23 Centre at Imperial College Healthcare NHS Trust and Imperial College London. SJ and AR are
24 employed by The George Institute for Global Health who own the IP for the polypill in Australia.
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29 **Ethics approval:** No ethics approval was required
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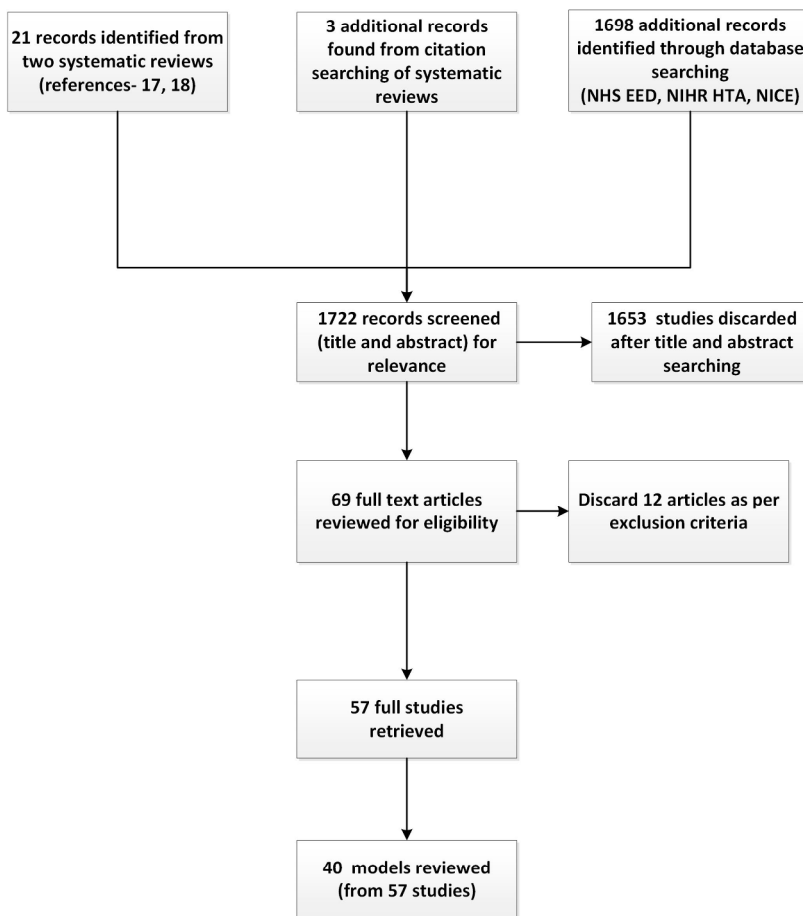
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Figure 1 Flowchart of Literature Review of cardiovascular disease models



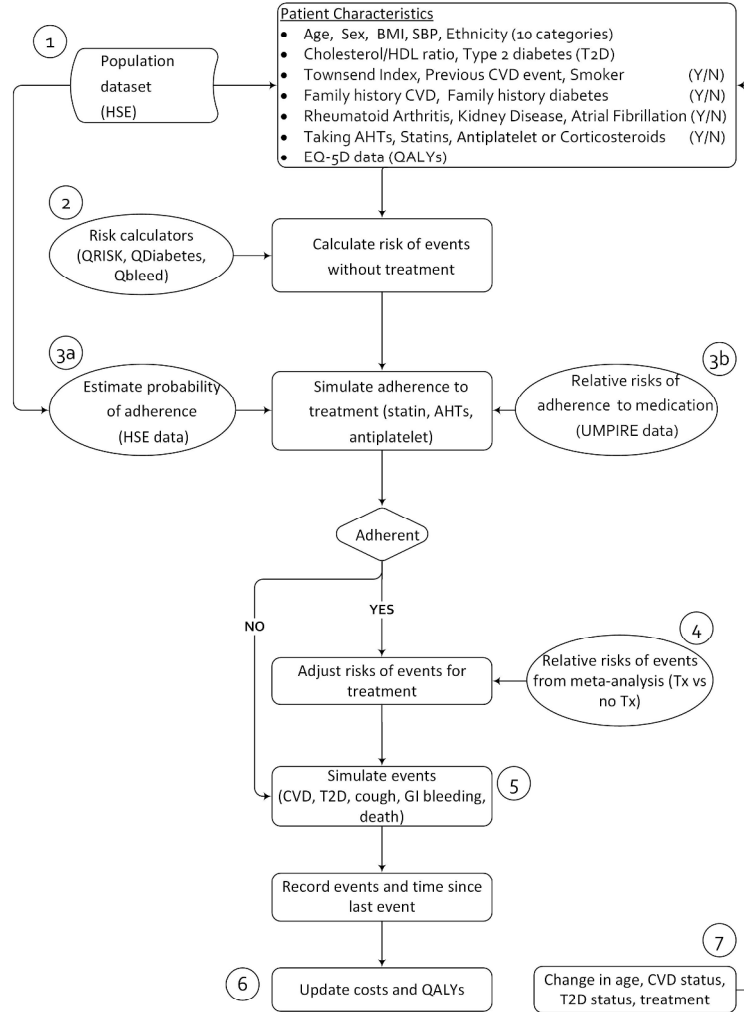
NHS EED: National Health Service Economic Evaluation Database,
 NIHR HTA: National Institute for Health Research Health Technology Assessment,
 NICE: National Institute for Health and Care Excellence

Figure 1 Flowchart of literature review search for cardiovascular disease models

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Figure 2 RUPEE (NHS) simulation model flowchart



HSE: Health Survey for England, T2D: Type 2 Diabetes, CVD: cardiovascular disease, AHTs: antihypertensives, tx: treatment, GI: gastrointestinal, QALYs: quality adjusted life years

Figure 2 RUPEE (NHS) Simulation model flowchart

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3 A Protocol for a modelled economic evaluation to evaluate aof a polypill in patients with
4 established or at high risk of cardiovascular disease in a UK NHS setting: *Researching the*
5 *UMPIRE Processes for Economic Evaluation in the UK National Health Service—RUPEE*
6 *(NHS) study*
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54 **Keywords:** Cardiovascular disease, polypill, adherence, cost-effectiveness, economic
55 evaluation
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58 **Word Count** 4,951784
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Figures 2 Tables 1

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Abstract

Introduction: The 'Use of a Multi-drug Pill in Reducing cardiovascular Events' (UMPIRE) trial was a randomised controlled clinical trial evaluating the impact of a polypill strategy on adherence to indicated medication in a population with established cardiovascular disease of or at high risk thereof. The aim of RUPEE-NHS is to estimate the potential health economic impact of a polypill strategy for CVD prevention within the NHS using UMPIRE trial and other relevant data. This paper describes the design of a modelled economic evaluation of the impact of increased adherence to the polypill versus usual care amongst the UK UMPIRE participants.

Methods and Analysis: As recommended by ISPOR-SMDM modelling guidelines a review of published CVD models was undertaken to identify the most appropriate modelling approach and structure. The review was carried out in the electronic databases, MEDLINE and EMBASE. 40 CVD models were identified from 57 studies, the majority of economic models were health state transition cohort models and individual level simulation models. The findings were discussed with clinical experts to confirm the approach and structure. An individual simulation approach was identified as the most suitable method to capture the heterogeneity in population CVD risk. RUPEE-NHS will use UMPIRE trial data on adherence ~~to medication if receiving the polypill versus usual care~~ to estimate the long term cost-effectiveness of the polypill strategy.

Dissemination: The evaluation findings will be presented in open access scientific and healthcare policy journals and at national and international conferences. We will also present findings to NHS policy makers and pharmaceutical companies.

Strengths and Limitations

This paper provides a clear outline of how a model for an economic evaluation is developed.

Providing an outline of the model structure which includes details on the underlying epidemiology and data inputs will add transparency to the findings of the RUPEE-NHS study

Though the model has been designed to include all major adverse and beneficial effects of treatment, the model structure will not include every potential treatment effect, for example the benefits of treatment on Alzheimer's disease will not be included.

INTRODUCTION

Adherence to recommended preventive medication regimes (1,2) in people at high risk of cardiovascular disease (CVD) is low, even in high income countries. (3) Poor adherence is associated with greater deterioration in health status and increased health care costs (4) and studies have shown that improved adherence to medication is associated with clinical benefits.(5) CVD preventive medication typically involves several drugs and adherence is inversely proportional to the number of prescriptions. Furthermore, physician inertia and patient resistance present barriers to initiating or restarting full recommended therapy. A single pill that includes several indicated drugs (a “polypill”), may improve long-term adherence by addressing these issues. If the polypill is priced lower than the price of the pills bought separately, it will also make it more affordable. (6,7) The UMPIRE (Use of a Multidrug Pill in Reducing Cardiovascular events) clinical trial was set up to evaluate the polypill in patients with or at high risk of CVD.

The UMPIRE trial randomised 2004 participants with established CVD (prior CVD event such as stroke or myocardial infarction) or at high risk of CVD (defined as a 5 year risk of >15%) based in India, England, Ireland and The Netherlands to either the polypill or usual care. The primary outcome of the trial was adherence to indicated treatments (statin, aspirin and two blood pressure lowering drugs), **measured as self-reported current use of antiplatelet, statin and ≥ 2 blood pressure lowering therapies for at least 4 days in the week preceding visits (baseline and end of trial visits)**. Other outcomes included systolic blood pressure (SBP) and low-density lipoprotein cholesterol (LDL-C). The trial found that the use of a polypill strategy resulted in greater adherence to treatment at 15 months and significant improvements in SBP and LDL-C. Detailed results and a description of the UMPIRE trial protocol are available. (8, 9)

UMPIRE collected data on resource use and self-reported health related quality of life using the EQ-5D. In order to estimate the long term costs and health outcomes associated with the polypill strategy an economic model is required. Due to differences in the patient population, care pathway and health care costs, separate analyses are needed for the four participating countries.

The analysis of the UMPIRE English trial data, (Researching the UMPIRE Processes for Economic Evaluation in the National Health Service (RUPEE-NHS)), aims to estimate the cost-effectiveness of the polypill strategy compared to conventional multi-drug therapy for the prevention of established cardiovascular disease in English NHS patients with or at high risk of CVD. The RUPEE (NHS) study

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3 will use UMPIRE English trial data on adherence to the polypill and will develop an economic model
4 to estimate cost effectiveness.
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8 The aim of this paper is to detail the modelling plan for the RUPEE (NHS) study.
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10 11 **METHODS**

12 13 **Model design process**

14 An economic model has been described as a mathematical framework that represents reality at an
15 adequate level of detail to inform clinical or policy decisions. (10) Guidelines on modelling produced
16 by the International Society for Pharmacoeconomics and Outcomes Research (ISPOR) and the
17 Society for Medical Decision Making (SMDM) joint taskforce recommend that it is best practice to
18 carry out a conceptualisation process prior to programming the economic model. This process has
19 two distinct components: specification of the study question and economic model. (11)
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25 26 **Specification of the study question**

27 The first component informs choices about how to structure the economic model and parameters.

28 The RUPEE (NHS) study aims to evaluate two different treatment strategies in a population with or
29 at high risk of CVD. The population for the economic model is defined by the inclusion criteria of
30 the UMPIRE trial. (9) The inclusion criteria are listed below:
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- 33 • Aged ≥ 18 years and
- 34 • High CVD risk defined as either established atherothrombotic CVD (history of coronary heart
35 disease (CHD), ischaemic cerebrovascular disease, or peripheral arterial disease (PAD)) or a 5
36 year risk of $\geq 15\%$ calculated using the Framingham risk equation
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43 The economic model will evaluate the polypill strategy compared to usual medication. In the
44 UMPIRE trial, participants assigned to the polypill received one of 2 versions: version 1 contained
45 aspirin 75mg, simvastatin 40mg, lisinopril 10mg and atenolol 50mg, and version 2 contained the
46 same ingredients but substituted hydrochlorothiazide 12.5mg for atenolol 50mg. Participants
47 assigned to usual care continued taking medications as prescribed by their general practitioner (GP).
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52 The RUPEE (NHS) study will follow guidelines for modelling health technologies as recommended by
53 the National Institute for Health and Care Excellence. (NICE) (11,12) Therefore a NHS and Personal
54 Social Services (PSS) perspective will be adopted to measure health service resource use and health
55 related quality of life will be measured by quality adjusted life years (QALYs) obtained using the EQ-
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3 5D. As per the NICE guidelines, costs and QALYS will be discounted at a rate of 3.5% per year. (12)
4 The time horizon reflected in the economic model will be lifetime to represent the chronic nature of
5 CVD.
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8 Conceptualisation of the economic model

9 The second component of the conceptualisation process involves defining the economic model.
10 There are two steps to this approach. The first step is to identify the appropriate modelling
11 approach. The modelling approach defines the analytical framework of the economic model.
12 Different types of analytical frameworks have been used to represent CVD including decision trees,
13 state transition models, compartmental models, individual simulation models and hybrid models
14 which often combine elements from different frameworks. (13-17)
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21 The second step determines the underlying structure of the analytical framework, which will
22 represent the disease and care pathway. The modelling approach needs to reflect: 1) CVD disease
23 and care pathway for this population; 2) the beneficial and adverse effect of treatment (polypill or
24 usual care); 3) the impact of increased adherence to treatment on health outcomes.
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30 The gGuidelines produced by ISPOR-SMDM on modelling recommend that existing models
31 addressing related problems should be reviewed as this approach can help identify both the
32 modelling approach and underlying structure.(11) To inform the RUPEE (NHS) economic model, we
33 carried out a review of published models evaluating interventions for CVD.
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38 Review of published CVD economic models

39 The purpose of the literature review was to identify the appropriate analytical framework to
40 represent the decision problem. The literature review also aimed to inform the underlying model
41 structure: disease and care pathway.
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45 Search strategy

46 The search strategy was conducted using the NHS Economic Evaluation Database (NHS EED), the
47 NIHR Health Technology Assessment (HTA) monograph series and the NICE guidelines website. The
48 search terms used included 'cardiovascular disease', 'coronary heart disease', 'stroke', 'myocardial
49 infarction', 'angina' and 'peripheral artery disease'. Studies were excluded from the review if they
50 did not discuss the development or review of an economic model; if no disease states for
51 cardiovascular disease were included in the model; -if the focus of the study was a diagnostic test or
52 surgical intervention where the economic model used a time frame of <10 years. Studies were not
53 excluded on the basis of intervention (drug treatment or lifestyle intervention) or on the basis of
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3 date published or language. We developed a data extraction form which included fields on model
4 purpose, structure, health states and events, transparency and validation. We did not collate
5 information about the findings of the model as the objective of the review was to identify alternative
6 model frameworks and methods used to represent CVD.
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11 An initial general literature search identified a 2006 systematic review of CHD policy models by Unal
12 *et al.* which was updated in 2008 by Capewell *et al.* and expanded to include stroke models. (17,18)
13 The review by Capewell *et al.* identified seven 'notable' CHD models (of which six had been
14 identified in the previous review by Unal *et al.*), nine stroke models and several models that were
15 currently in development at the time of publication. We reviewed the notable models and models in
16 development identified by Capewell *et al.* Citation searching of both systematic reviews was carried
17 out to identify other models published since 2008.
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22 23 *Review findings*

24 Overall [22-57](#) studies were identified which reported on 40 CVD models. Figure 1 presents the
25 flowchart for the search strategy.
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28 The searches found several studies which reported on the same model, for example the IMPACT
29 CHD model developed by Capewell *et al.* was used in analyses of CVD in other populations. (19) In
30 some cases, a model was adapted for different analyses, such as the Sheffield model which was
31 developed to evaluate statin therapy and was then adapted for use in the development of the NICE
32 guidelines for lipid modification. (1,20) The Sheffield model was also partially used in a whole
33 population modelling study by Barton *et al.* (13)
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39 Further details on the review can be found in the supplementary appendix. The appendix includes a
40 list of the reviewed models (see Table 1 supplementary appendix), an example of the data extraction
41 form and an example of a ~~schematic~~ schematic illustration and details of one of the reviewed models (see
42 Figure 1 supplementary appendix). Schematic illustrations of several models were used in
43 discussions with clinical experts about the different types of modelling approaches.
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49 **[Figure 1- Flowchart for ~~search strategy for CVD model~~ search strategy]**

50 Modelling approach

51 The search identified that the two most commonly used modelling approaches were health state
52 transition cohort models and individual-level simulation models. Both approaches were critically
53 assessed to determine their suitability to capture the disease and care pathway.
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3 A cohort model can be defined as any model which estimates the outcomes for a group of patients,
4 whereas with a patient level simulation, outcomes are evaluated at the individual level. Therefore,
5 one of the main differences between the two approaches is how they estimate costs and QALYs:
6 cohort models estimate expected costs and QALYs for the modelled population as a whole, whereas
7 individual level simulation models estimate cost and QALYs for each individual and the average is
8 taken across the sample.
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14 With a health state transition cohort model, the population progresses through a set of mutually
15 exclusive health states at regular intervals called cycles, determined by a predefined transition
16 matrix. Health state transition cohort models are also commonly called Markov models. However,
17 such models are only Markovian when they display the Markovian 'memoryless' property where the
18 progression of the patient through the model is only dependent on the current state in which the
19 patient resides and not on anything that happened before they entered that health state. It is also
20 possible to model at the individual level using a state transition model by sampling probabilities for
21 each individual patient to experience a particular transition in each model cycle. (21)
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29 Both model approaches can use a discrete time approach: with this approach the model cycle length
30 will be defined in advance. The cohort or individual progress through health states or events which
31 represent the disease pathway and only one event may occur within each cycle length. Costs and
32 QALYs are updated once per cycle. Alternatively, individual level simulation models are often set up
33 as discrete event simulations (DES). With a DES approach, an event can occur at any time point, for
34 example, an event could occur at three months, one year and twenty years. As an event occurs,
35 costs and QALYs are recorded and updated for each individual.
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42 A health state transition model was used to develop NICE guidance for lipid modification treatment.
43 (1) The limitation of this approach is that it may be unable to capture the underlying heterogeneity
44 in the population. Individual CVD risk can be estimated using CVD risk algorithms such as QRISK2
45 which use a range of patient characteristics such as age, sex, ethnicity, systolic blood pressure and
46 body mass index to estimate a 10 year CVD risk.(22) To capture this complexity in a health state
47 transition model would require the construction of a large number of subgroups to reflect different
48 subsets of patient characteristics and the variation in CVD risk in the population. This could become
49 impractical to model. It also has the disadvantage that accuracy could be lost by using
50 representative values for subgroups. An individual simulation model structure may be more
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3 appropriate to model the level of detail required to estimate CVD outcomes reflective of those in the
4 population.
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8 The Markovian memoryless property means that data on individual patients' history is not
9 retained as they progress through the model. Accounting for individual patient history in a
10 Markov model would require multiplying the number of health states to an infeasible level
11 where the model would become too complex and impracticable to run.
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16 To accurately identify the effectiveness of each treatment strategy in a population with or at high
17 risk of CVD, an individual simulation model was deemed the most appropriate for the RUPEE (NHS)
18 study to reflect the heterogeneity in the population which impacts on the risk of a CVD event and
19 subsequent costs and outcomes. The individual simulation model will use a discrete event approach
20 to handle time.(21)
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25 Model structure

26 The findings of the review were discussed with clinical experts to confirm the health events and the
27 methods used to model the progression of persons through the disease pathway.
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30 *Model events (CVD, diabetes and adverse events)*

31 The most commonly included types of CVD events in the reviewed models were CHD (angina and
32 myocardial infarction), cerebrovascular events (transient ischaemic attack (TIA) and stroke) and
33 Peripheral arterial disease (PAD). It was decided that the CVD events relevant for the current
34 model would reflect those most commonly included in prior such models. PAD will not be included
35 as a CVD event in the model as there is less likely to be a definable acute PAD event compared to
36 other CVD events such as MI and stroke. We will assume that patients can experience more than
37 one CVD event in their lifetime. The risk of CVD will also be assumed to change with age in the
38 model.
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46 Diabetes is a risk factor for CVD with a substantial cost and impact on health related quality of life,
47 therefore diabetes will be included as a comorbidity in the model. The risk of new onset diabetes
48 will be estimated using the QDiabetes risk algorithm.(23)
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53 Adverse effects from treatment will include an increase in the risk of new onset diabetes resulting
54 from treatment with statins and antihypertensive drugs. (24-27) The risk of a persistent cough
55 resulting from treatment with angiotensin-converting-enzyme inhibitors (ACE inhibitors) will be
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3 included as an event. The probability of a cough resulting from treatment will be sourced from
4 meta-analyses of randomised controlled trial (RCT) data for ACE inhibitors. As aspirin use is
5 associated with an increased risk of gastrointestinal bleeding,(28,29) an increased risk of
6 gastrointestinal bleeding from treatment with aspirin will be included.
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11 Renal impairment will not be included in the model as an adverse effect of ACE inhibitors. Whilst
12 ACE inhibitors may cause an acute rise in serum creatinine in a few patients with renal artery
13 stenosis and more generally cause a slight short term increase in creatinine levels, the effects are
14 complex and there may be a net improvement in renal function overall in a treated population. The
15 rate of falls and fractures will be estimated not to alter, given the evidence from randomised trials of
16 blood pressure lowering agents, although this is an area of debate with regard to patients with
17 higher levels of frailty. (30, 31)
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24 Other adverse effects from statin treatment such as liver dysfunction and myopathy will not be
25 included in the model as these cases are rare and are assumed to have a minimal impact on
26 outcomes. (1) (32)
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30 Treatment with antihypertensives is associated with a reduction in heart failure, therefore this will
31 be included as an outcome in the model. (33) Other outcomes of treatment are likely but will not
32 be included – for example a reduction in cancer with aspirin use of more than 5 years. (34)
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37 *Progression of individuals through model*

38 The progression of persons through the disease pathway differs depending on the modelling
39 approach: health state transition models such as the Markov model developed for NICE guidelines
40 on lipid modification use a predefined transition matrix to determine progression through the CVD
41 health states.(35) Alternatively, simulation models can use risk algorithms to estimate the
42 probability of CVD events or new onset diabetes. The NICE guidelines for lipid modification
43 recommend the use of QRISK2, which is a risk algorithm derived to estimate primary CVD risk in UK
44 populations. (1,22) The QRISK2 risk algorithm predicts the risk of a 10 year CHD event (angina, MI)
45 or a cerebrovascular event (TIA, stroke). It does not include the risk of PAD. An alternative CVD risk
46 algorithm is the Framingham equation;(36), however, a validation study comparing QRISK2 and
47 Framingham found that QRISK2 is better calibrated to a UK population.(37) The RUPEE (NHS) model
48 will therefore use the QRISK2 risk algorithm.
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RUPEE (NHS) economic model

Figure 2 depicts the flowchart of the RUPEE (NHS) model structure. The oval shapes represent data inputs to the model, whereas the rectangular shapes represent processes.

[Figure 2 - Flowchart of RUPEE (NHS) model structure]

Model description

In the RUPEE (NHS) model costs and QALYs are recorded for each individual and an average cost and QALY for the simulated population are estimated. The RUPEE (NHS) model will be run twice, once to simulate costs and QALYs under usual care and once to simulate costs and QALYs under the polypill strategy-scenario (polypill scenario will include polypill version 1 and version 2). Individuals representing the UMPIRE trial inclusion criteria will enter the model (label 1 in Figure 2), and their baseline risk of a CVD event and onset diabetes will be estimated using the QRISK2 CVD risk algorithm and or the QDiabetes algorithm (label 2 in Figure 2) respectively. For each individual, whether or not they are adherent to medication will be simulated using Monte Carlo simulation based on the probability of adherence in usual care (label 3a in Figure 2). If the individual is simulated to be adherent to medication their risk of a CVD event will be modified by a treatment effect (label 4 in Figure 2). In the polypill scenario of the model, the probability of adherence will be further modified by the relative risk of adherence to medication. ~~For simulated adherent individuals taking the polypill, the risk of an event will be further modified by the relative risk of adherence for polypill versus usual care.~~ The relative risk of adherence for polypill to medication versus usual care will be sourced from the UMPIRE trial data (label 3b in Figure 2). Individuals may experience a CVD event or onset of diabetes based on their estimated CVD and diabetes risk, which will be estimated using the QRISK2 and QDiabetes algorithms. Individuals may also experience an adverse reaction to medication (if adherent) including gastrointestinal bleeding, early onset of diabetes and a persistent cough. Costs and QALYs will be recorded for each event (including adverse events). Individuals can experience more than one event (model run for lifetime horizon) and patient characteristics such as age and history of previous events, such as a stroke or new onset diabetes, are updated during the model run, with an ensuing reflective increase in the risk of an event.

Input parameters

Each point in the flowchart is labelled and a description of the process or data requirement label is described below. Table 1 provides further details on data input parameters for the RUPEE-NHS model and potential sources of data.

1. Population Dataset

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3 We will use the 2011 Health Survey for England (HSE) as a population dataset for the economic
4 model. The HSE is a cross sectional survey which contains anonymised information on a
5 representative sample of the population. The 2011 HSE dataset collected information on CVD,
6 including individual CVD events and medication history. The dataset also contains information on
7 demographic and socio-economic characteristics and health related data such as body mass index
8 (BMI), SBP and LDL-C and history of CVD events. These data are required in order to estimate
9 individual baseline risks of CVD and diabetes in the model.
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14 15 16 **2. Calculation of baseline risks of events without treatment**

17 Baseline risks for CVD for each sampled individual will be calculated using published risk algorithms.
18 As per recent NICE guidance for lipid modification, we will use the recommended algorithm for CVD
19 risk, QRISK2. (1,221) The algorithm was derived using QRESEARCH, a large database derived from
20 the pseudonymised health records of over 13 million patients registered with a general practitioner
21 in the UK. If an individual has established CVD (previously experienced a CVD event), we will
22 estimate a secondary CVD risk using the REACH algorithm. (38) A baseline risk for the onset of
23 diabetes will be estimated using the QDiabetes algorithm. (23)
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32 **3a. Simulating adherence to treatment under usual care**

33 The RUPEE study will evaluate the effect of adherence to medication on long term costs and health
34 outcomes measured using quality adjusted life years (QALYS). The average rates of adherence in
35 clinical trials can be higher than in actual practice (4) as seen in the UMPIRE clinical trial population
36 which had an atypically high baseline adherence rate. Instead, adherence rates to medication
37 (antihypertensives-(AHT), statins and aspirin) under a usual care setting will be sourced from the
38 2011 HSE dataset. Participants in the 2011 HSE self-reported all the prescribed medications they
39 had taken in the last 7 days. This was coded in the HSE dataset using the British National Formulary
40 (BNF) classifications codes. Using this data, we are able to identify the medication patients were
41 prescribed and identify whether or not they were taking the prescribed medication in the last week.
42 This will reflect adherence to medication in a usual care population. The data will be used to
43 estimate the probability of each person being adherent or not to medication. Individual
44 characteristics will be used as predictors of adherence: the characteristics will be chosen by referring
45 to studies which have assessed predictors of adherence in persons taking treatment for CVD.-(39,
46 40) A generalized linear mixed regression model will be used to estimate the probability of
47 adherence to medication for each individual. The probability of persistence with medication will not
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3 be assumed to be constant, and the model will include a probability of ceasing medication over time.
4 The probability of medication cessation will be sourced from published literature on adherence.
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8 **3b. Estimate relative risks of adherence to medication for polypill versus usual care**

9 We will estimate the relative risks of adherence to medication for the polypill strategy versus usual
10 care, using a generalised linear mixed regression model which will be applied to the UMPIRE trial
11 dataset (UK dataset).—A generalised linear mixed regression model will be applied to the UMPIRE
12 trial dataset, with adherence to medication indicated as taking ≥ 2 antihypertensive drugs, a statin
13 and aspirin for at least four days in the week prior to a recorded visit.—The This definition of
14 adherence reflects that used in UMPIRE. (8). In the polypill scenario in the model, the probability of
15 being adherent to medication will be further modified by the relative risks.
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22 **4. Adjust risk of events for treatment**

23 We will source data on the treatment effects of statins, antihypertensives and aspirin from meta-
24 analyses of intention-to-treat RCTs. Intention-to-treat analyses account for non-adherence in their
25 findings, and therefore underestimate the impact of treatment on event risk. To overcome this, we
26 will carry out sensitivity analyses to test the impact of adjusting for adherence within the trial. The
27 risk of a CVD event will be adjusted by the relative risk of ~~treatment~~ treatment with statins,
28 antihypertensives and aspirin, based on the ~~treatment~~ medication(s) the person is taking and
29 whether or not they are adherent to medication.
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37 **5. Simulation of events**

38 Individuals in the model can experience a CVD event at a rate governed by their calculated baseline
39 risk (estimated by the QRISK2 or REACH algorithms) and adjusted for treatment effects if they have
40 been simulated as adherent to treatment. CVD events will be categorised as a TIA, stroke, MI or
41 angina. The relative incidence of each CVD event will be determined using published incidence
42 data.⁽³⁹⁴¹⁾ Similarly, the risk of new onset diabetes will be calculated using the QDiabetes
43 algorithm. We will simulate the incidence of adverse events as a result of treatment: new onset
44 diabetes and gastrointestinal bleeds. Data on the probability of an adverse event will be sourced
45 from meta-analyses of randomised controlled trials for the relevant drugs. Mortality risk will be
46 modelled as mortality from stroke and MI and other cause mortality. Data on other cause mortality
47 will be estimated using national life tables for England and Wales. ⁽⁴⁰⁴²⁾
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55 **6. Assign cost and quality of life values**

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3 Costs and QALYs associated with each individual's simulated lifetime profile of CVD and related care
4 will be estimated. Costs and quality adjusted life years (QALYs) will accrue for each person to reflect
5 events, such as a stroke or new onset diabetes. Costs and utility values for health events will be
6 sourced from published studies including the NICE guidelines for lipid modification and
7 hypertension. (1,35,4143) Costs of medication will be sourced from the NHS National Drug
8 Tariff.(4244)
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20 **7. Change in age, treatment, CVD status and type II diabetes status**

21 The simulation model will run for each individual for lifetime duration (death or maximum age of
22 100 years) and patient characteristics will be updated after each event or every 10 years (depending
23 on which event occurs first). A 10 year update is used as the QRISK2 algorithm returns a 10 year CVD
24 risk.
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Table 1) Input parameters

Model inputs	Source
1. Individual dataset	
Population dataset	Initial patient characteristics (see Figure 2) for cohort of patients drawn from a representative national sample: Health Survey for England (HSE) dataset 2011. The dataset will include patients who meet the entry criteria for the UMPIRE trial.
2. Calculation of baseline risks	
Risk calculators	Risk of first CVD event and onset of type 2 diabetes estimated for individuals using QRISK2 and QDiabetes.(23, 24) <i>QRISK2: 10 year CVD risk (CVD outcomes defined as angina, MI, TIA and stroke)</i> <i>QDiabetes: risk of acquiring type 2 diabetes over 10 year time period</i> Risks for subsequent CVD events estimated for individuals using the REACH algorithm. (38) <i>CVD outcomes defined as cardiovascular death (includes fatal stroke and MI), non-fatal MI, non-fatal stroke, and cardiovascular hospitalisation (includes hospitalisation for unstable angina and TIA)</i>
Risk of heart failure	Baseline risk per age using incidence rates in Cowie et al. (1998) (4345)
Relative incidence of CVD events (TIA, stroke, angina, MI)	OXVASC cohort study, Rothwell <i>et al.</i> 2005. 91,106 individuals presenting with an acute vascular event in Oxfordshire, UK in 2002-5. (3441)
3. Adherence to medication	
Probability of adherence to treatment with usual care	Estimates from HSE 2011 dataset on adherence to relevant drugs (statins, antihypertensivesAHT, aspirin)
Relative risk of adherence: polypill versus usual care	Estimate the probability of adherence to ≥ 2 antihypertensivesAHT, statin or anti-platelet for at least four days in the preceding week -for polypill group versus usual care by applying a binomial regression to the UMPIRE dataset.
4. Treatment effects of medication (antihypertensivesAHT, statin, anti-platelet)	
Relative risk of CVD with treatment versus no treatment	For base case analysis, conventional meta-analysis of ITT RCT data will be used from – <ul style="list-style-type: none"> ▪ Cholesterol Treatment Trialists' Collaboration ▪ Blood Pressure Lowering Treatment Trialists' Collaboration ▪ Antithrombotic Trialists' Collaboration ▪ Law and Wald (4446) Sensitivity analysis: Test impact of adjusting for adherence within trials
5. Other treatment outcomes (beneficial events and adverse events) and mortality rates	
Adverse Events	
Incident type 2 diabetes	Relative risk of diabetes from statins/antihypertensives from meta-analyses of RCTs
GI bleeding	Relative risk of bleeding resulting from aspirin using estimates from meta-analyses of RCTs

Cough	Placebo-adjusted relative risk of cough resulting from ACE inhibitors using estimate from meta-analyses of RCTs
Reduction in heart failure	Relative risk reduction in heart failure from antihypertensives (33)
Mortality	
<i>Stroke case fatality (60 day)</i>	
Age <75	Estimate proportion of strokes that are fatal (with risks increasing with age). Estimate using the BHF Compendium of health statistics 2012, which has data from a record linkage study for England 2010.
Age > 75+	
<i>MI case fatality (30 day)</i>	
Age 30-54	Proportion of MI's that are fatal. Estimate using Oxford Record Linkage pill study (study. (4547) National population based study, including all individuals admitted to hospital or who died suddenly from acute MI in 2010. Age was strongest predictive factor for 30-day case fatality.
Age 55-64	
Age 65-74	
Age 75-84	
Age 85+	
<i>Death from other causes</i>	Estimated from national life tables (Office for National Statistics, England)(4042)
6. Costs (medication, monitoring costs, health events)	
<i>Drug costs (£ per year)</i>	
Statins	National Health Service (NHS) Electronic Drug Tariff (44)
AHT drugs	
Aspirin	
Polypill	
<i>Yearly monitoring costs while on medication</i>	
Primary care nurse (£ per hour)	Use NICE Quality Outcomes Framework to identify recommended management while on treatment (statins, antihypertensivesAHT, antiplatelet). A cost for stopping medication will also be applied (e.g. 2 GP visits, tests as recommended in NICE clinical guidelines 181) (1)
GP cost (£ per hour)	
Lipid test (£)	
Liver transaminase test	Costs sourced from Personal Social Services Research Unit Costs and NICE clinical guidelines 181
Blood tests	
<i>Costs of health states and adverse events</i>	
Stroke	Luengo-Fernandez <i>et al.</i> 2012 (4648)
TIA	"
MI	NICE lipids guideline 181 (1)
Angina	"
PAD	"
Diabetes	"
GI bleeding	"
Cough (from ACE inhibitor use)	NICE lipids-Hypertension guidelines 127 (43)
7. Health Related Quality of life	
Stroke	Derived from Health Survey from England (HSE) dataset
TIA	
MI	
Angina	
PAD	
GI bleeding	
Diabetes	
Cough	

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3 CVD: cardiovascular disease, TIA: transient ischaemic attack, MI: myocardial infarction; ITT: Intention
4 to treat, RCT: ~~r~~andomised controlled trial, AHT: antihypertensives, UMPIRE: Use of a Multidrug Pill in
5 Reducing Cardiovascular Events, NICE: National Institute for Health and Care Excellence, GP: general
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For peer review only

Analysis

The simulation model will run for a sufficient number of iterations to provide stable results. Uncertainty in the model parameters will be examined using a probabilistic sensitivity analysis (PSA) which will reflect uncertainty over the values of the model inputs. Non-parametric bootstrapping of HSE data will be carried out to examine the uncertainty related to the sampling. For each PSA iteration, one non-parametric bootstrap sample will be drawn from the HSE dataset (by random sample with replacement of individuals in the dataset). An incremental analysis will be conducted and incremental cost-effectiveness ratios (ICERs) and net benefit statistics will be estimated. We will also carry out a number of sensitivity analyses to test the impact of varying uncertain parameters in the model. This will include an analysis testing the impact of varying the polypill cost.

Validation

The model will be internally and externally validated. A checklist produced by the RUPEE steering group based on current published guidelines for checking models will be used, to ensure the programmed model behaves as expected according to the theoretical model. (21, 4497) The checklist includes tips for model developers, for example on the use of sensitivity analyses to test that the model is operating correctly, and re-programming complicated sections of code in another language. The model will also be reviewed and tested by an experienced modeller. The model results will be compared with real-world observations or the results of other models.

Dissemination of results

The findings of the economic evaluation will be presented to scientific and health care policy audiences in open access journals and at national and international conferences. We will also present findings to NHS policy makers and pharmaceutical companies.

DISCUSSION

Medication adherence is important for disease management, and benefits of increased adherence to preventative medication for CVD include improved clinical outcomes. (5) The UMPIRE clinical trial was conducted to evaluate the effect of a polypill strategy compared to usual care on adherence. It showed that the polypill strategy significantly augmented adherence and this was reflected by improvements in SBP and LDL-C. (8) Whether or not this impact remains in the long term cannot be determined from the trial data alone. The RUPEE (NHS) study is being conducted to evaluate the long term impact of a polypill strategy; in particular, the analysis will evaluate the long term impact of increased adherence on outcomes. An economic model is being developed to estimate the long term costs and QALYs associated with implementing a polypill strategy in the NHS compared to usual

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3 care. This analysis will represent the first comprehensive cost effectiveness analysis using directly
4 applicable clinical trial data.
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8 This paper outlines the process behind the design of the economic model. We carried out a review
9 of published CVD models to identify a modelling approach that would suit the health care decision:
10 use of a polypill versus usual care in a population with or at high risk of CVD. We identified an
11 individual simulation model as the most appropriate approach as it allows the heterogeneity in the
12 population to be adequately reflected. The model will use validated disease risk algorithms to
13 estimate the probability of an individual experiencing a CVD event or the onset of diabetes.
14 Individuals can also experience an increased risk of an adverse event (diabetes, cough and
15 gastrointestinal bleeding) from treatment. The risk of a CVD event will be reduced if the individual
16 is adherent to treatment. We will simulate adherence to treatment using data from the HSE 2011
17 dataset. ~~The probability of adherence in the polypill scenario will be further modified by the -with-an~~
18 ~~increased modified relative risk reduction probability of adherence in the polypill scenario if the~~
19 ~~person is adherent to the polypill. The relative risks~~ of adherence to ~~the polypill versus usual~~
20 ~~care medication~~ which will be sourced from the UMPIRE trial data for the English population. Costs
21 and QALYs will be estimated for each individual and aggregated across the sample population (based
22 on the HSE 2011 dataset).
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34 The RUPEE (NHS) model will have a number of advantages over existing models constructed to
35 evaluate a CVD polypill. (4850-5052) One advantage is the use of an individual simulation model
36 which will allow us to capture the heterogeneity in the variation in CVD risk in the UK population
37 unlike other models which use Markov type transition state models. Another is that we will
38 extrapolate data on adherence to medication from a nationally representative population dataset
39 (Health Survey for England) which will allow us to simulate adherence per individual rather than
40 assuming a constant adherence across our population. We will also allow for adverse events from
41 treatment and treatment cessation, therefore more accurately reflecting clinical practice.
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49 It would be preferable to use per protocol treatment effectiveness data in our analysis as intention-
50 to-treat data already accounts for adherence (people switching and ceasing medication during the
51 trial period). However, per protocol data is difficult to obtain for all drugs, therefore we will use the
52 ITT treatment effect data and carry out sensitivity analyses to test the impact.
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3 The introduction of a CVD preventive polypill strategy will simplify pill taking for patients potentially
4 leading to greater adherence and better health outcomes. This analysis will provide information on
5 the cost-effectiveness of the polypill in a NHS setting.
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10 **Contributors:** CC carried out the literature review and drafted the manuscript. JL, ST, HMD, HW and
11 NP, AR and SJ contributed to the development of the protocol. JL provided input on the health
12 economics model. HMD provided statistical advice and ST, HW and NP contributed clinical advice.
13 AR and SJ peer reviewed draft manuscripts and contributed to the final version of the protocol. All
14 authors approved the final version of the manuscript submitted for publication.
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23 ~~1112-29080). The views expressed are those of the author(s) and not necessarily those of the NHS,~~
24 ~~the NIHR or the Department of Health.~~
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31 Centre at Imperial College Healthcare NHS Trust and Imperial College London. SJ and AR are
32 employed by The George Institute for Global Health who own the IP for the polypill in Australia.
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37 **Ethics approval:** No ethics approval was required
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41 ~~(NIHR) under its Research for Patient Benefit (RfPB) Programme (Grant Reference Number PB-PG-~~
42 ~~1112-29080). The views expressed are those of the author(s) and not necessarily those of the NHS,~~
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Supplementary appendix

Section 1- Literature review

The purpose of the literature review was to identify the appropriate analytical framework to represent the decision problem. The literature review also aimed to inform the underlying model structure: disease and care pathway.

A general search of the literature identified a known review of coronary heart disease policy models by Unal et al. (2006).⁽¹⁾ This review was updated and expanded in 2008 by Capewell et al. to include stroke models. (2) A search carried out in Medline found no further systematic reviews of coronary heart disease or cardiovascular disease models published since 2008. The review by Capewell et al. (2008) identified seven notable CHD models (six of these had been identified in the previous review by Unal et al), nine stroke models and several models that were currently in development at the time of publication. The notable models and models in development were reviewed. Additionally, citation searching of both reviews was carried out to potentially identify any further models published since 2008.

Development of search strategy

The purpose of the review was not to identify every single model for cardiovascular disease but rather to identify potential model structures that could be adapted or used to help construct the RUPEE NHS model. Initially, it was planned that an updated search using the search strategy devised by Unal et al. (2006) and updated by Capewell et al. (2008) would be used. However, the purpose of both reviews had been to identify notable policy models at the population level. It was felt that redoing this review could potentially fail to return other models which could be used such as those developed for NICE guidelines. The choice of databases was discussed with a systematic reviewer based at HERG. The search strategy was carried out using the following databases:

- NHS economic evaluation Database (NHS EED): this database contains economic evaluations of healthcare interventions and is produced by the NIHR Centre for Reviews and Dissemination (CRD) at the University of York, UK.
- National Institute for Health Research (NIHR) Health Technology Assessment (HTA) monograph series: This series publishes research about the effectiveness, costs and broader impact of healthcare treatments and tests (within a UK National Health Service (NHS) setting).
- National Institute for health and care excellence (NICE) website: this database publishes evidence based guidance on preventative, diagnostic and treatment interventions for disease and ill health.

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3 NHS EED was identified as an appropriate database as this database reviews and produces critical
4 commentaries economic evaluations of 'key' relevance to the UK NHS. The critical commentaries
5 provide a summary of the overall reliability and generalisability of the study. The NICE HTA monograph
6 series publishes research including cost-effectiveness analyses of healthcare treatment and tests; the
7 series was searched to identify published HTA's which have developed or used a cardiovascular
8 disease model. The NICE guidelines website was searched to identify guidelines related to
9 cardiovascular disease (for example guidelines for lipid modification).

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12 The search terms used in the search included 'cardiovascular disease', 'coronary heart disease',
13 'stroke', 'myocardial infarction', 'angina' and 'peripheral artery disease'. Appendix 1 contains further
14 details of the searches carried out in each database.

21 22 **Exclusion criteria**

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24 Studies were excluded if they did not discuss the development or review of an economic model; if no
25 disease states for cardiovascular disease were included in the model; if the focus of the study was a
26 diagnostic test or surgical intervention where the focus of the evaluation was a short term follow up,
27 (<10 years). Studies were not excluded on the basis of intervention (treatment or lifestyle
28 intervention) or on the basis of date published or language.

31 32 **Data extraction form**

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34 The use of a standard checklist such as the Drummond economic evaluation checklist was considered
35 to review each study but was found to be unsuitable for reviewing the models as the design of the
36 checklist leads the reviewer to evaluate the cost-effectiveness analysis inputs and outcomes with only
37 two questions referring to the model structure: regarding the choice and details of the model. (3)

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39 Therefore, a data extraction form was designed to extract data that was required to meet the purpose
40 of the review. An initial data extraction form was developed which extracted data on the following
41 items:

- 42 • Paper (Author, Year)
- 43 • Purpose of the Model
- 44 • Setting and Population
- 45 • Interventions
- 46 • Type of model (Simulation, Markov Model, other)
- 47 • Brief description of Model
- 48 • Cardiovascular disease risk algorithms

- Risk factors included to calculate cardiovascular disease risk
- Disease stages (Health states) included in model
- Source of data inputs used in model (Population data, mortality rate, treatment uptake and effectiveness, other)
- Probabilistic Distributions and Parameters

The form was refined further to only extract data which was relevant for this review. As the purpose of the review was to inform the model structure and design the extraction of data inputs and probabilistic distributions and parameters were removed from the data extraction form. The initial data extraction form also extracted data on the quality of each model. An assessment of quality criteria for models has been suggested in guidelines from the International Society for Pharmacoeconomics and Outcomes Research (ISPOR). (4,5) The systematic review by Unal et al. used the guidelines suggested by ISPOR to create a grading system for model papers based on the sensitivity, validity and transparency of a model. As the purpose of the review is not to evaluate inputs, the form was further refined and information on sensitivity analyses were not extracted. However, the data extraction forms did extract information on whether the model had been validated (including details of validation). The refined data extraction form also included a section on whether the model had been adapted for further studies. An example of a completed data extraction form can be found in Section 2.

Categorisation of Models

Each model was categorised (modelling approach) based on the taxonomy of model structures as developed by Brennan et al. (6)

Findings

The majority of models identified for review used a state transition approach (13 models) with five models adopting a hybrid state transition, in all cases a hybrid Markov-simulation model. (7-24) Only one decision tree model was identified, whereas 10 individual simulation model were identified. (25-35) Another popular approach was to use a systems dynamics modelling approach (5 models).(36-40) Other modelling approaches identified included an age period cohort (APC) approach (1 model); a tabular cell based model used by the World Health Organisation to estimate the global burden of disease; two life table approaches; a mathematical stroke epidemiological model and the Archimedes model which uses a method based on Fourier expansions using standard mathematical techniques to simulate individuals (proprietary model). (41-46)

The identified models categorised according to modelling approach can be found in Table 1.

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5 The purpose of the review was to identify the best modelling approach for the RUPEE study. To this
6 end, we reviewed the models to identify advantages and disadvantages of each approach. Details of
7 the advantages and disadvantages of the two main modelling approaches used (Markov models and
8 Simulation models) can be found in the paper associated with this supplementary appendix. Table 2
9 presents a summary of six models which used a different modelling approach (decision tree, state
10 transition model, simulation model, systems dynamic and hybrid model).
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17 Additionally, schematic illustrations of several models were used to aid discussions about the different
18 types of modelling approaches with clinical experts. Figure 1 in Section 2 is an example of the
19 schematic illustration for the model developed for the NICE clinical guidelines 181 which evaluated
20 statin treatment in primary and secondary care. (47)
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Table 1- Models identified in Literature Review Search

Life Table/Cell base/Tabular model	Decision Tree	State Transition	Hybrid model	Simulation	Systems Dynamic/ Compartmental model
WHO Global Burden of Disease (42)	Whitfield et al. (UK) (25)	Grover et al. CVD Life Expectancy model (Canada)- Markov Model (7)	Rotterdam Ischemic disease and stroke (RISC) model Markov model structure with individual simulation (20)	Southampton CHD Policy Analysis Model 'Treatment' – individual simulation (26)	Weinstein et al. CHD heart disease policy model (USA) (36)
Schau et al. Stroke Model (Denmark)(45)		Stroke Treatment Economic Model (STEM)- USA (8)	Duke Stroke Policy and Prevention Model USA (SPPM) Semi-Markov/simulation model (21)	CHD Policy Analysis Model 'Prevention'- individual simulation (27)	IMPACT model (including adaptations of model) (37)
Tobias et al. APC Model (41)		RIVM Chronic Disease- Markov Model (9)	A Dynamic modelling tool for generic health impact assessments (Dynamo-HIA) Markov/partial simulation model (22)	Prevent – Macro simulation model using aggregated data (policy tool) (28)	Sundberg et al.- Compartmental model (38)
Struijs et al. Dynamic multi-state life table (43)		Ward et al . (ScHAAR statins model) and adaptations- Markov Model (10) (47)	Korean Individual Microsimulation Model for Cardiovascular Health Interventions Hybrid Markov/ individual simulation model (23)	Foresight Obesity Model UK – stochastic cohort simulation approach (29)	Model of Resource Utilization, Costs and Outcomes for Stroke, (MORUCOS, Australia)- Compartmental model (39)
Archimedes (USA) (46)		Smith-Spangler et al- Markov Model (11)	Sorensen et al. Simulation model Markov model/individual simulation (24)	POHEM- Canada, Microsimulation (30)	PopMod: a longitudinal population model with two interacting disease

					states- Compartmental model (40)
Malik et al. Life Table (44)		Newman et al. Combination polypharmacy, Markov model (12)		EUROASPIRE III health economics project- Individual simulation (31)	
		Grosso et al, Markov Model (13)		OECD and WHO microsimulation chronic disease prevention simulation model- microsimulation (32)	
		Gillespie et al. SPHERE Markov Model (14)		Ara et al. Obesity model- Cohort simulation (33)	
		Wisloff et al. NorCaD Markov Model (15)		Department of Health Vascular Checks Model- Simulation (34)	
		Nash et al. Markov Model (16)		Green et al. Chronic Disease Policy Model- Discrete Event Simulation (35)	
		Lovibond et al. Markov Model (17)			
		Greving et al. Markov Model (18)			
		NICE Clinical Guidelines CG127 (19)			

WHO- World Health Organisation, APC- Age Period Cohort, CHD- Coronary heart disease, OECD- Organisation for Economic Co-operation and Development, SchAAR- School of Health and Related Research, NICE- National Institute for Health and Care Excellence

Table 2- Summary of CVD models

Model name (Author)	Model Type	Risk factors	Health States/Events	Transparency & Validation	Limitations
Stroke Model (Whitfield et al.) (25)	Decision Tree	BMI, Type II diabetes, smoking, total and HDL cholesterol, SBP	Acute episode included: Acute CVD, Elective CVD, Heart Failure, Renal replacement procedures, Stroke, Diabetes (hypoglycaemia)	Internal validation: predicted number of CVD related admissions based on risk factor data compared to actual data (from five UK primary care trusts)- found results to be accurate No external validation conducted	The model uses an aggregate approach despite having individual data Short time frame also used, suitable for decision tree but potentially if a longer time frame was used this would not be a suitable model
NICE lipid modification guidelines economic model (CG181) (47)	Markov Model	Not explicitly stated	Death from cardiovascular cause and non CVD death, stable angina, unstable angina, myocardial infarction, transient ischaemic attack, heart failure, peripheral artery disease and post event states for each non-fatal event	Yes, the model structure, assumptions and inputs are clearly reported Validation has not been stated, this is an update of a previously widely used model (SchAAR statins model NICE TA94) (10)	The model is limited by the Markovian assumption of memoryless though it does have tunnel (post event) states. The cohort can experience each event only once The model structure is not suitable to simulate a heterogeneous population
RISC state transition hybrid model (20)	Hybrid Markov Model	sex, age, smoking status, SBP & DBP, BMI, waist to hip ratio, ankle-brachial index, levels of plasma glucose, total cholesterol,	Well, Stroke, CHD, CHD & Stroke, Other Death, CVD death	Internal validation: cumulative incidences simulated by RISC model compared to Rotterdam study incidences- similar.	Allow for individual heterogeneity to be modelled, but limited by Markovian state transition model (progression

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		HLD, creatinine, family history CVD, hypertension, taking antihypertensives or BP over 160/90, presence diabetes II, intermittent claudication, angina, AF, TIA or prevalent CVD		External validity tested- used NORFOLK EPIC dataset and simulation incidences using model- incidences similar	between states and handling of time). Could potentially be slow computationally to run (uses six transition probabilities equations per individual) if more health states or risk factors are required
IMPACT (Capewell et al) (37)	Compartmental/ systems dynamics model	Cigarette smoking, total cholesterol, systolic blood pressure, BMI, diabetes, physical activity and fruit and vegetable consumption	Deaths prevented or postponed from reductions in risk were the main CHD outcome <u>Nine patient groups were evaluated:</u> Patients treated in hospital for acute myocardial infarction (MI) Patients admitted to hospital with unstable angina Community dwelling patients who have survived a MI >1yr Patients who had undergone a previous	A technical appendix was provided a recent paper which used the IMPACT model and this provided detailed information on the equations used to estimate deaths prevented or postponed from a treatment intervention or a reduction in CVD risk factors and provided all data sources that were used in the modelling	Cost and QALYs were not Considered The model did not look at the reduction in CVD events, it was limited to avoided mortality from CHD A recent expansion of the model (IMPACT 2) is available, however though online this model is a black box and a technical appendix was not available IMPACT2 is a DES model

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			<p>revascularisation procedure</p> <p>Community dwelling patients with coronary artery disease</p> <p>Patients admitted to hospital with heart failure</p> <p>Community dwelling patients with heart failure</p> <p>Hypercholesterolaemic patients without CHD</p> <p>Hypertensive patients without CHD</p>		
<p>CHD Policy Analysis Model – Prevention component (Babad et al). (27)</p>	<p>Discrete Event Simulation</p>	<p>Age, sex, SBP, total cholesterol and smoking</p>	<p>Onset of stable angina, unstable angina, myocardial infarction, sudden cardiac death, stroke death, other cardiovascular disease, cancer death and death from other or unknown cause (potential to include HDL cholesterol)</p>	<p>The model structure could be replicated – however no data inputs are given regarding treatment effectiveness</p>	<p>Use of Framingham study to estimate baseline risk-recent studies have shown that QRISK is more suited to a UK population</p> <p>Computational requirements: Model was run in special software (POST, DELPHI framework). This type of model would be computationally intensive to run in widely available</p>

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					packages such as Microsoft Excel
Department of Health Vascular checks economic model (34)	Individual simulation	Age, gender, townsend score, BMI, SBP, Smoking status, Total cholesterol/HDL ratio, record family history of CHD	Not explicitly stated. Costs and health benefits applied in the model were based on published NICE guidance (PH1002, CG43, CG34 and TA94)	The model inputs and data inputs are clear	Cost and QALYs relating to interventions were not directly estimated: rather they were sourced from existing guidance and linked to the simulation outputs Requirement for a suitable large dataset to simulate can be expensive. The Department of Health used the proprietary GP database QRESEARCH (approximate cost of dataset £15-20,000).

BMI-body mass index, HDL- high-density lipoprotein, SBP- systolic blood pressure, DBP- diastolic blood pressure, CVD-cardiovascular disease, CHD-coronary heart disease, BP- blood pressure, AF- atrial fibrillation, TIA- transient ischemic attack, NICE-National Institute for Health and Care Excellence, DES-discrete event simulation, MI-myocardial infarction, SchAAR-School of Health and Related Research, RISC- Rotterdam Ischemic disease and stroke model

Section 2- Example of completed data extraction form

Model name: Southampton Disease Model (CHD Policy Analysis Model) 'Treatment Model'

Paper (Author, Year): The development of a simulation model of the treatment of coronary heart disease (Keith Cooper and Ruth Davies, 2002)

Journal: Health Care Management Science 5, 259-267

Model Details

Model Structure: Discrete Event Simulation

Model software: Patient orientated simulation technique (POST) software with a Delphi interface.

Study Population & Setting: Individuals with stable angina, unstable angina or myocardial infarction (till age 85 or death).

Purpose of Model: The model is used to evaluate revascularisation at a hospital level rather than population based. Looks at progress of patients after a coronary event.

Patient characteristics: Given attributes of age, gender, vessel disease, time before cardiac death and time to age 85.

Model Description: New patients enter the model with SA, UA or MI (proportion randomly determined using incidence rate of disease). The following assumptions are employed:

- Risk of non-cardiac death
- Risk SA or UA leads to risk of MI
- SA leads to risk of UA
- Sampled time to event (MI, death, UA) depend on age and vessel disease
- Risks of UA, MI, & death increase with age, severe vessel disease and with a history of previous myocardial infarctions.
- Risks are independent of each other and are multiplied by baseline risks to change the projections of MI and death.
- Time updated Gompertz distribution (hazard function) used to estimate time to event (includes relative risks from vessel disease, prior history and interventions)

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4 **Progression to health events:** Stable angina- Start in GP state receiving treatment (medical), assuming
5 some have symptoms controlled and some not. Some are transferred to outpatient's investigations
6 (now or in x years). After outpatients, some join a waiting list for an angiogram (queue) and those
7 who do not go to a medical treatment stage. Vessel disease extent will determine next step after
8 angiogram (can change this rule/input in simulation). Patients can bypass graft, angioplasty.
9 Incidence data from Health Survey for England and GP Morbidity data.

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14 **Validation of model:** validated cardiac deaths against mortality data from Office for National Statistics,
15 based on death certificates. Model did underestimate deaths in females. Authors surmised this was
16 due to poor reporting of causes of death on certificates.

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20 **Limitations** Study (2002) does not mention the application of costs or QALYs and it looks at CHD events
21 only. The authors noted that the model will be developed further to link the outputs to costs and to
22 include secondary prevention such as aspirin or anti-cholesterol agents and to link the treatment with
23 the prevention model (Prevent model developed by Babad et al.)

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28 **Summary-** Discrete event simulation model for progress of patients after a coronary event. Individuals
29 have angina and can progress to unstable angina or myocardial infarction. Changes in risks in one part
30 can affect other parts of model. This model did allow for resource constraints such as availability of
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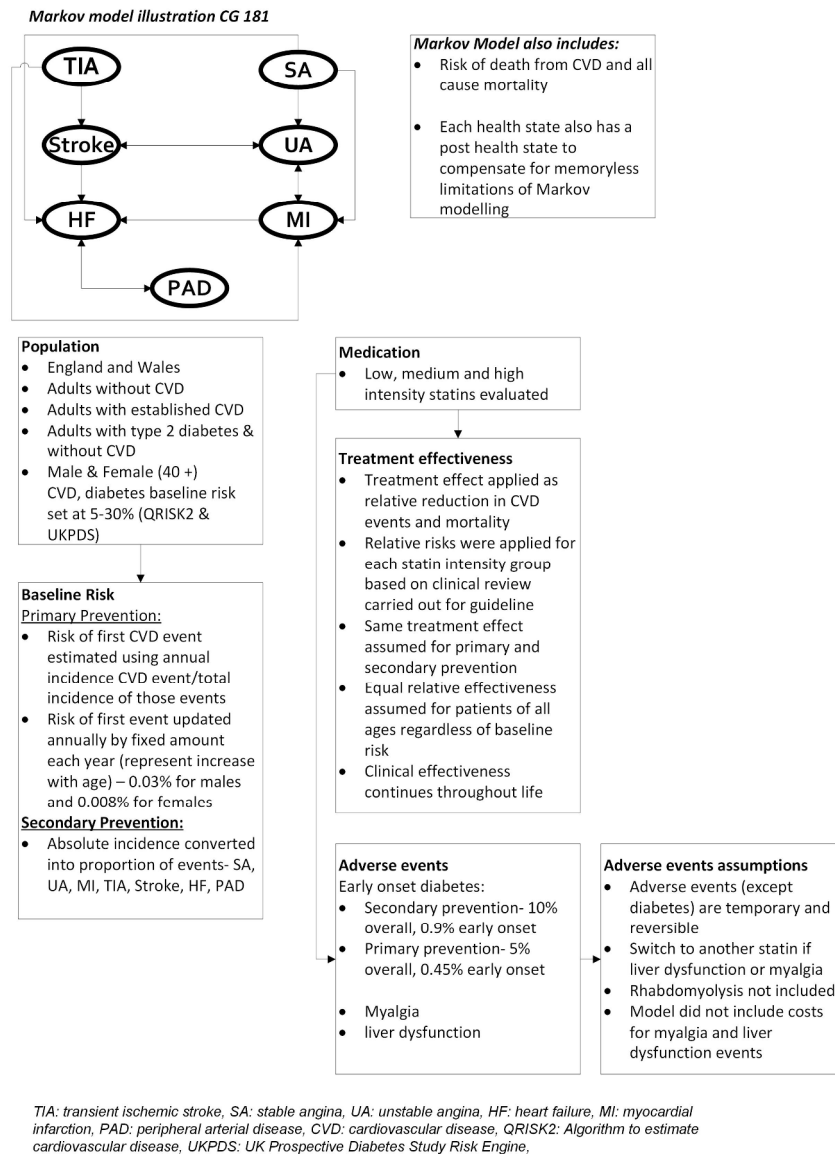
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Supplementary Appendix Figure 1 Illustration and details of NICE Clinical Guidelines 181 Markov Model



Supplementary Appendix Figure 1 Illustration and details of Markov model developed for NICE clinical guidelines 181

209x293mm (300 x 300 DPI)