

## **A policy model of cardiovascular disease in moderate-to-advanced chronic kidney disease:**

**motivation, challenges and user interface**

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**Workshop on R for trial and model-based cost-effectiveness analysis**

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# Outline

- Motivation
- Model development
- Challenges in implementing in R
- Shiny user-friendly interface & demonstration

## References:

1) Model paper: Schlackow I et al. A policy model of cardiovascular disease in moderate-to-advanced chronic kidney disease. Heart 2017.

<https://heart.bmj.com/content/103/23/1880>

2) Model interface <http://dismod.ndph.ox.ac.uk/kidneymodel/app/>



# Motivation: research context

- Research question: What is the cost-effectiveness of lowering cholesterol in categories of CKD patients (eg, by CKD stage and CVD risk) with statin-based treatment?
- Key data: SHARP, a RCT studying the efficacy question in a range of CKD patients
  - 9,270 participants; followed for 4.9 years
  - At entry: Age 62, 63% men; 13% smokers; 23% diabetes; 15% vascular disease; CKD stage 3B (25%), 4 (29%), 5 (14%), on dialysis (33%)
  - During the study: hospital admissions data throughout; EQ-5D-3L at study end
    - 1,736 from 6,245 pre-dialysis participants progressed to dialysis
    - 1,135 received kidney transplant
    - 1,515 experienced 1+ Major Vascular Events (MVEs)
    - 2257 died: 749 died from vascular causes



# So, why R?

- Funder brief: A model fit for NICE submission (...Excel in 2010!); adaptable to other jurisdictions
- However, based on previous experience:
  - **Excel**: inflexible for the model complexity envisaged, difficult to error-proof, slow to evaluate uncertainty
  - **Stata, SAS**: programming structures limit flexibility; no user-friendly interface option
  - **R**: least restrictive technically; emerging option/s (in 2010!) for input/output from Excel (ie, Excel interface of an R model)
  - For an external use with large participant datasets the model might need to be programmed in a **general-purpose programming language** (e.g. C++) to speed execution time.



# Developing the model: premises and challenges

## Structural

### Duality between chronic kidney and cardiovascular disease

- Chronic kidney disease progresses to more advanced stages
- Kidney disease increases risk of cardiovascular events
- Cardiovascular events may accelerate kidney disease progression

## Computational

- Large number of states
- Predictions based on time-updated patient covariates

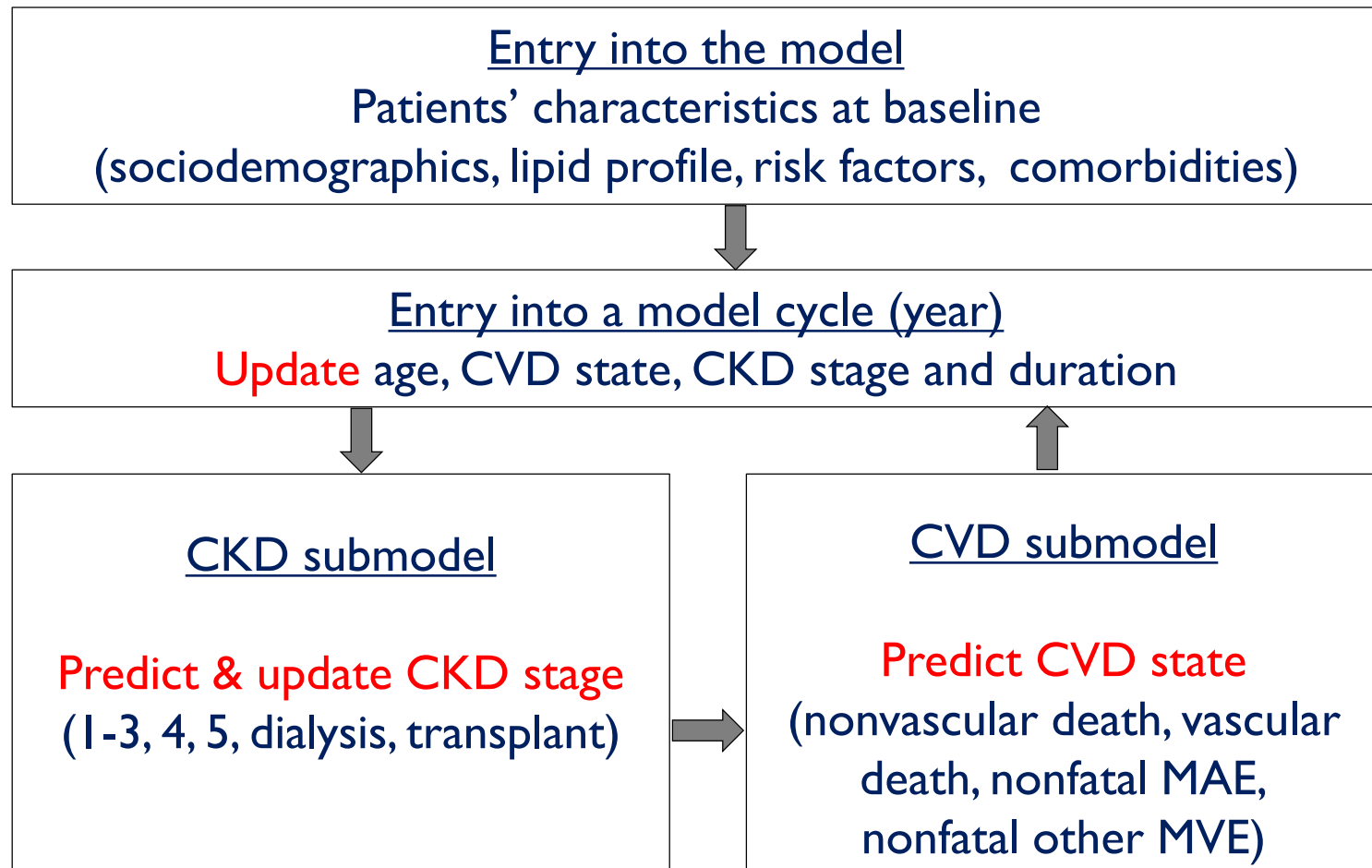


# SHARP CKD-CVD model: structure

- Markov cohort model with an annual cycle of transition
  - Sub-model 1: Chronic kidney disease progression (CKD)
  - Sub-model 2: Cardiovascular disease complications (CVD)
- Backbone: A number of Risk/Cost/QoL regression equations based on SHARP participant-level data
  - 3 CVD; 2 CKD, 1 hospital cost; 1 QoL; further data (NVD; transplant failure)
- For each patient and each year, progression of CKD and major CVD complications and deaths are simulated
- Occurrence of CKD/CVD events in the model contribute to risk of subsequent such events and mortality



# SHARP CKD-CVD model: structure



MAE= Major Atherosclerotic Event; MVE= Major Vascular Event



# CVD risk equations

		Hazard Ratio	
		Vascular death (Exponential PH)	Non-fatal MAE or vascular death (Gompertz PH)
<b>Current age</b>	x10yrs	1.6*	1.4*
<b>Current CKD stage (ref: stage 3B)</b>	stage 4	1.8*	1.3*
	stage 5	2.4*	1.8*
	transplant	1.1	0.9
	dialysis (ESRD <3y)	3.4*	2.6*
	dialysis (ESRD >3y)	5.1*	3.3*
<b>Current CV history (ref: no events)</b>	Vasc disease at baseline	1.7*	2.0*
	MAE <1y ago	3.0*	4.3*
	MAE 1-2y ago	2.6*	3.5*
	MAE >2y ago	1.5*	2.3*
	Other MVE	4.7*	4.2*
<b>Ancillary parameter in survival model</b>		-	-0.0002*

VD model adjusted for Tx allocation, sex, country/ethnicity, smoking, diabetes, DBP, haemoglobin, type of renal disease  
 NFMAE/VD model adjusted for Tx allocation, sex, country/ethnicity, diabetes, SBP, haemoglobin, total cholesterol, HDL cholesterol,  
 type of renal disease \* p<0.01





# Decision analytic model states

Cardiovascular states	Kidney disease states
No events (with/without prior vascular disease)	Stage 1-3
Non-fatal MAE <1y ago	Stage 4
Non-fatal MAE 1-2y ago	Stage 5
Non-fatal MAE >2y ago	Transplant
Non-fatal other MVE <1y ago	Dialysis, ESRD <3y
Non-fatal other MVE >1 ago	Dialysis, ESRD >3y
	<b>Plus the ESRD tunnel states</b>

Any combination of cardiovascular and kidney disease states is possible plus two death states- vascular death and non-vascular death.

$$6*11+2 = 68 \text{ states}$$



# SHARP CKD-CVD model: incorporating disease duality

For each person; for each year

update age and CKD duration

CKD submodel

For each possible starting state

update CVD history & CKD stage

predict possible CKD stages

update transitional probability matrix

Update distribution across possible states

CVD submodel

For each possible starting state

update CVD history & CKD stage

predict possible CVD states

update transitional probability matrix

Update distribution across possible states



# SHARP CKD-CVD model: challenges in R



# SHARP CKD-CVD model: challenges in R

- ❑ Initial run on 10,000 patients took >24 hours!
  - Even without uncertainty



# SHARP CKD-CVD model: tips for speeding up calculations

## ❑ Employ parallel programming

- Calculations run in parallel for several patients
- This reduces the time by (N of cores) times
- Even standard desktops often have 4 cores
- R packages: foreach, parallel, doParallel:
  - Start a cluster & export functions/packages
  - Replace `for (i in 1:n) {...}` with `foreach (i = 1:n) %dopar% {...}`
  - Stop the cluster
- NB: No parallel programming in Excel!



# SHARP CKD-CVD model: tips for speeding up calculations

- ❑ Pre-calculate as much as possible, eg
  - Give unique numbers for each state (1..68)
  - For each state, record
    - Values of covariates for CVD history & CKD stage (eg state 1 is “no MVE, CKD stage 1-3b”\*)
    - Possible states to transition into (eg from state 1 can go into states 1, 2, 3, 4, 8 but NOT state 23 “MAE last year, CKD stage 1-3b)
    - Information is generated once and recorded externally
    - In running through the loop, the information is only referred to, but NOT re-generated
- ❑ In each loop, only run through possible transitions
  - Remove “no MAE” row if the starting state is “MAE last year”
  - Massively reduces size of the transition matrix
    - From state 1, **seven columns instead of 68**

\*As different equations employ different covariates categories, values should be recorded separately



# SHARP CKD-CVD model: tips for speeding up calculations

- ❑ Split (and pre-save) lookup datasets
  - Separate survival tables for males & females & CKD stages
- ❑ Use fixed numbers/coordinate locations instead of names
  - Read off gender from v[2] instead of v["sex"])

```
if (vX_0[2] == 1) { # if patient male
  if (stageRand %in% c("dialysis", "transplant")) # if patient ESRD
    ratesNVD_2 <- ratesNVD_M_ESRD else # if patient not ESRD
    ratesNVD_2 <- ratesNVD_M
} else { # if patient female
  if(stageRand %in% c("dialysis", "transplant")) # if patient ESRD
    ratesNVD_2 <- ratesNVD_F_ESRD else # if patient not ESRD
    ratesNVD_2 <- ratesNVD_F
}
```



#### #### starting states

```
states0 <- states_info[unlist(states0_num)]  
states0_lab <- sapply(states0, '[', 1) # state names
```

#### #### end states

```
states1 <- states_info[unlist(states1_num)]  
states1_lab <- sapply(states1, '[', 1) # state names
```

#### #### transitional matrix

```
MCKDrownames <- states0_lab # rows  
MCKDcolnames <- states1_lab # columns  
MCKD <- matrix(nrow = length(MCKDrownames), ncol = length(MCKDcolnames),  
               dimnames = list(MCKDrownames, MCKDcolnames))  
MCKD[, ] <- 0
```

#### #### recovering information for the starting state state0

```
N_CV0 <- state0[[2]] # CV part of the state  
N_CKD0 <- state0[[8]] # CKD part of the state  
vX_state0 <- state0[[12]] # vector of covariates corresponding to the state
```





# SHARP CKD-CVD model: computational challenges

- ❑ Initial run on 10,000 patients took **>24 hours!**
- ❑ Simple tinkering reduced it to **50 minutes**
- ❑ Most reduction came from optimising the algorithm
  - Parallel programming
  - Reducing size of transitional matrices
- ❑ Only basic R functionality was used
  - 9 packages: data manipulation (reshape2, plyr, data.table); parallel programming (foreach, snowfall, doSNOW); plotting and saving (xtable, scales, ggplot2)
  - No specific CEA packages



# SHARP CKD-CVD model: user-friendly interface

- The model to be useful for NICE, other analysts, clinicians...
- User-friendly interface accessible from anywhere
- No need for knowledge (or installation) of specialist software, eg R
- Adaptation to other population settings/countries (customising parameters)



# Shiny: introduction

Shiny is an R package that allows to build interactive web applications straight from R

- ❑ The application can be accessed by clicking the link
- ❑ The user only sees the front end; no knowledge/installation of R is required
- ❑ All programs/data are located on an external server
- ❑ The front end can be modified using CSS themes, htmlwidgets, and JavaScript actions
  - Possible to insert fancy fonts, links, email addresses etc
  - Possible to implement error checking on data entry
- ❑ Further information and tutorials at <https://shiny.rstudio.com/>



# SHARP CKD-CVD model: Shiny interface

← → ↻ dsmod.ndph.ox.ac.uk/kidneymodel/app/



## SHARP CKD-CVD outcomes model

Introduction

**Model overview**

Glossary

File specifications

Model parameters

Type of analysis

Patient characteristics

Treatment parameters

Annual healthcare costs

### Introduction

The SHARP CKD-CVD outcomes model simulates long-term cardiovascular event rates, kidney disease progression, (quality-of-life adjusted) survival and healthcare costs associated with individual patient profiles and treatments. It can be applied to patient populations with moderate-to-severe chronic kidney disease who are over 40 years of age, and can be used with individual patients as well as groups of patients.

The model reports long-term projections as well as cost-effectiveness results comparing against the 'no treatment' strategy. The evaluated health outcomes and costs are reported separately for each treatment arm. The user can vary parameters to assess sensitivity of the results.

To perform the analysis, specify the required parameters using the 'Model parameter' tabs and click on the 'Run analyses' button on the [Results](#) tab. Please refer to the [User guide](#) and the [published manuscript](#) for further information.

The [Glossary](#) tab contains a list of commonly used definitions.

### Citation

When referring to this program in publications, please cite the following references:



# SHARP CKD-CVD model: Shiny files structure

**ui** (compulsory in a separate file pre-shiny 0.10.2)

- Commands describing layout and structure of the user interface
- Definition of font styles (normal, headings, error messages)
  - Eg display error messages in red
- Action button, checkbox, date input, file input...
  - Eg numeric input or slider for age; select box for gender
  - Each entry is given a unique name
- Layout can be conditional
  - Eg display 95% CI only if performing uncertainty analysis



# SHARP CKD-CVD model: Shiny files structure

ui (compulsory in a separate file pre-shiny 0.10.2)

**server** (compulsory in a separate file pre-shiny 0.10.2)

- Commands describing server functions definitions
- All data entered by the user are read off and passed on to corresponding variables/parameters
  - Eg define age to be the value from the numeric input called “age”
- Actions associated with data entry are defined
  - Eg perform validity check on entering numeric values; execute master function on pressing the “run” action button.



# SHARP CKD-CVD model: Shiny files structure

ui (compulsory in a separate file pre-shiny 0.10.2)

server (compulsory in a separate file pre-shiny 0.10.2)

## master code

- master functions performing calculations and taking parameters from server
  - Eg perform the cost-effectiveness analysis based on the user-input parameters
- Output is passed on back to ui/server to be displayed in the pre-defined user-friendly format



# SHARP CKD-CVD model: Shiny files structure

ui (compulsory in a separate file pre-shiny 0.10.2)

server (compulsory in a separate file pre-shiny 0.10.2)

master code

**www folder**

- Contains files that are (or can be) used by the program and accessible externally
  - Eg user manual or example input/output files





# SHARP CKD-CVD model: Shiny files structure

▼	code	Today at 14:42	--	Folder
	master code.R	Today at 00:43	77 KB	R Source File
	shiny functions.R	8 Aug 2016 at 12:12	16 KB	R Source File
▼	data	Today at 00:13	--	Folder
	coeffs_default.Rdata	24 Jan 2017 at 15:04	3 KB	R Data File
	coeffs_PSA.Rdata	24 Jan 2017 at 15:04	1.7 MB	R Data File
	Default_coefficients.Rdata	24 Jan 2017 at 14:51	3 KB	R Data File
	default_patient.Rdata	24 Jan 2017 at 14:55	318 bytes	R Data File
	pat0.Rdata	24 Jan 2017 at 14:56	417 bytes	R Data File
	patT.Rdata	24 Jan 2017 at 14:55	154 bytes	R Data File
	PSA_coefficientsNew.Rdata	24 Jan 2017 at 14:51	1.9 MB	R Data File
	ratesNVD.Rdata	24 Jan 2017 at 14:52	1 KB	R Data File
	states_and_endpoints.Rdata	24 Jan 2017 at 15:00	7 KB	R Data File
	server.R	Today at 00:46	101 KB	R Source File
	ui.R	Today at 00:43	30 KB	R Source File
▼	www	Today at 00:13	--	Folder
	2014_UK_CKD_NVD.csv	28 Jun 2016 at 15:01	1 KB	comma...values
	default_patient.csv	24 Jan 2017 at 14:55	320 bytes	comma...values
	example_input.csv	10 Apr 2017 at 12:47	429 bytes	comma...values
	example_output_CE_ind.csv	10 Apr 2017 at 12:58	3 KB	comma...values
	example_output_CE_PSA_ind.csv	10 Apr 2017 at 16:47	10 KB	comma...values
	example_output_CE_PSA_sum.csv	10 Apr 2017 at 16:47	4 KB	comma...values
	example_output_CE_sum.csv	10 Apr 2017 at 12:57	1 KB	comma...values
	example_output_LP_ind.csv	10 Apr 2017 at 12:59	2 KB	comma...values
	example_output_LP_PSA_ind.csv	10 Apr 2017 at 15:33	3 KB	comma...values
	example_output_LP_PSA_sum.csv	10 Apr 2017 at 15:33	1 KB	comma...values
	example_output_LP_sum.csv	10 Apr 2017 at 12:58	440 bytes	comma...values
	glossary.pdf	21 Jul 2016 at 12:19	10 KB	PDF document
	userguide.pdf	10 Apr 2017 at 17:02	2.4 MB	PDF document



# SHARP CKD-CVD model: Shiny interface



<http://one-elevenbooks.com/shiny-or-the-truth/>

<http://dismod.ndph.ox.ac.uk/kidneymodel/app/>



Discount cost-effectiveness results

Long-term projections in the control group (cumulative probabilities per 1,000 participants)

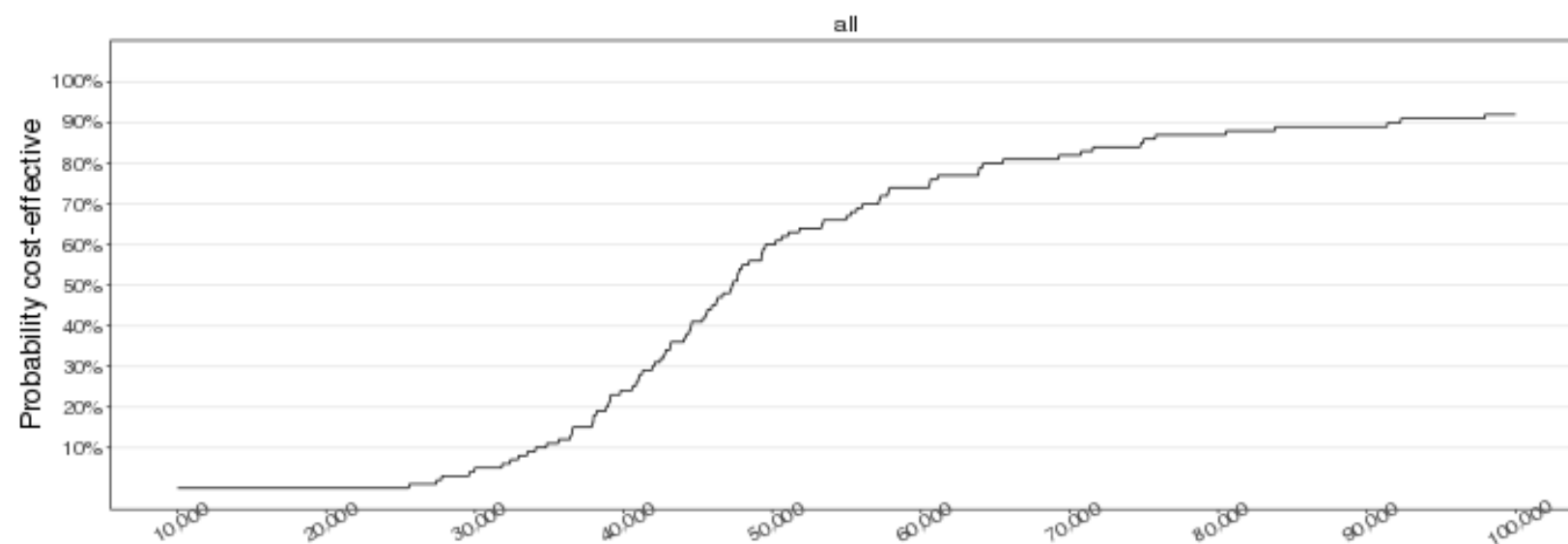
	MVE or VD	RRT	Vascular deaths	All deaths
At 5 years	184 (159, 213)	409 (357, 444)	57 (44, 76)	205 (194, 222)
At 10 years	281 (244, 319)	643 (594, 683)	118 (92, 155)	415 (398, 438)
Over simulation duration	419 (358, 501)	884 (826, 935)	292 (225, 379)	907 (897, 918)

Long-term projections in the treatment group (cumulative probabilities per 1,000 participants)

	MVE or VD	RRT	Vascular deaths	All deaths
At 5 years	189 (138, 193)	407 (355, 439)	51 (37, 70)	200 (189, 216)
At 10 years	283 (214, 299)	638 (593, 675)	106 (81, 140)	407 (389, 429)
Over simulation duration	397 (328, 477)	877 (813, 927)	271 (207, 371)	905 (896, 915)

Incremental cost-effectiveness over the simulation duration (results per 1,000 participants)

LYs gained	QALYs gained	Incremental hospital costs	Treatment costs	Cost per LY gained	Cost per QALY gained
135 (-4, 279)	107 (22, 227)	698,152 (-416,384, 1,306,000)	5,074,512 (4,904,776, 5,201,336)	42,646 (20,617, 304,068)	54,085 (27,412, 179,555)



# SHARP CKD-CVD model: conclusions and more challenges

## Conclusions:

- SHARP CKD-CVD model is a novel resource for evaluating health outcomes and cost-effectiveness of interventions in CKD
- Efficient coding is important and much can be achieved using basic functionality only
- The model has a user-friendly web-based and freely available interface
- The user can adapt model with enter their own parameter values and perform calculations in different settings
- The model is already being used externally by others!



# SHARP CKD-CVD model: conclusions and more challenges

## Further/remaining challenges:

- To further increase speed of execution
  - Particular problem when running uncertainty
  - Re-write in another language, eg C?
  - Re-design the model structure? Eg fewer states; fewer covariates; different endpoints; move away from cohort simulation?
- Day-to-day support
  - Replying to queries, bug fixing
  - R/package updates may break everything!



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# SHARP CKD-CVD model



essentially,  
all models are wrong,  
but some are useful

George E. P. Box

freshspectrum.com

<http://dismod.ndph.ox.ac.uk/kidneymodel/app/>

[kidneymodel@ndph.ox.ac.uk](mailto:kidneymodel@ndph.ox.ac.uk)

