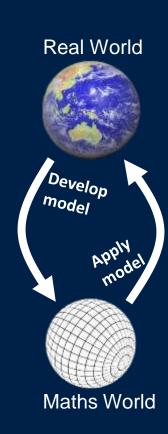
Malaria Elimination Trials and Simulations



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University of Oxford

A mathematical model...

- is a description of a system or multiple interacting systems using mathematics
- can be used to answer "what if" questions about the system
- can be simple or complex
- has many limitations:
 - Simple models have more limitations in biology but are faster to write and run with small data
 - Complex models have more detail but are slower to write and run and demand large sets of data



World Health Organization Global Technical Strategy for Malaria 2016-2030

Pillar 1		Pillar 2	Pillar 3
Ensure universal access to malaria prevention, diagnosis and treatment		Accelerate efforts towards elimination and attainment of malaria-free status	Transform malaria surveillance into a core intervention
Supporting Element 1	Harnessing Innovation & Expanding Research		
Supporting Element 2	Strengthening the Enabling Environment		

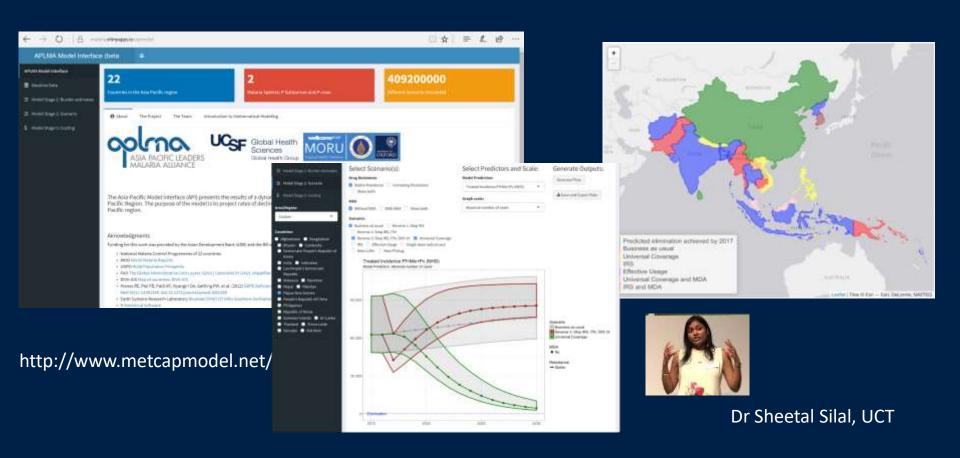
Pillar 1

Ensure universal access to malaria prevention, diagnosis and treatment

Pillar 2

Accelerate efforts towards elimination and attainment of malaria-free status

METCAP: Malaria Elimination Transmission and Costing in the AsiaPacific



Model Equations

$$\frac{dS}{dt} = \mu P(t) - \mu S - \Lambda(t)S + \omega R$$

$$\frac{dI_n}{dt} = -\mu I_n + p_{en}(1 - p_s)\Lambda(t)S - r_nI_n + r_aI_a - (1 - p_{rn})(1 - p_r)\Lambda(t)I_n - p_r\Lambda(t)I_n + p_{rn}(1 - p_r)\Lambda(t)(R + H)$$

$$\frac{dI_o}{dt} = -\mu I_o + (1 - p_{re})(1 - p_s)\Lambda(t)S + (1 - p_{ser})r_cI_c - r_oI_o + (1 - p_{re})(1 - p_r)\Lambda(t)I_n - p_r\Lambda(t)I_e + (1 - p_{re})(1 - p_r)\Lambda(t)(R + H) + ptf(1 - ptfe)r_1(T_o + T_\tau + T_h)$$

$$\frac{dI_c}{dt} = -\mu I_c + (1 - \tau)p_s\Lambda(t)S + (1 - \tau_{scv})(1 - \theta_1)r_sI_s - (1 - p_{scv})r_cI_c - p_{scv}r_cI_c + p_r(1 - \tau)\Lambda(t)(I_n + I_n + R + H) + ptf(ptfc)(1 - ptftr)r_c(T_o + T_v + T_h)$$

$$\frac{dI_s}{dt} = -\mu I_s - (1 - \tau_{scv})r_s I_s - \tau_{scv} r_Q I_s + p_{scv} r_c I_t$$

$$\frac{dT_o}{dt} = -\mu T_o + \tau_o p_s \Lambda(t)S - (1 - ptf)r_l T_o + p_r \tau_o \Lambda(t)(I_n + I_a + R + H)$$

$$\frac{dT_v}{dt} = -\mu T_v + \tau_v p_s \Lambda(t)S - (1 - ptf)r_t T_v + p_r \tau_v \Lambda(t)(I_n + I_o + R + H)$$

$$\frac{dT_h}{dt} = -\mu T_h + \tau_h p_s \Lambda(t)S - (1 - ptf)r_t T_h + p_r \tau_h \Lambda(t)(I_h + I_h + R + H) + ptf(ptfc)(ptftr)r_t(T_o + T_e + T_h)$$

$$\frac{dR}{dt} = -\mu R + r_n I_n - \Lambda(t)R - \omega R + \chi I$$

$$\frac{dH}{dt} = -\mu H + (1 - ptf)r_t(T_o + T_v + T_h) + \tau_{sev}(1 - \theta_2)r_QI_s - \Lambda(t)H - \chi H$$

$$\frac{dS}{dt} = \mu P(t) - \mu S - \Lambda(t)S + \omega R + \kappa L$$

$$\begin{split} \frac{dI_n}{dt} &= -\mu I_n + p_{rn}(1-p_s)\Lambda(t)S - r_nI_n + r_aI_a - (1-p_{rn})(1-p_r)\Lambda(t)I_n - p_r\Lambda(t)I_n + p_{rn}(1-p_r)p_{rel}\nu L \\ &+ p_{rn}(1-p_r)\Lambda(t)(R+L) \end{split}$$

$$\begin{split} \frac{I_s}{dt} &= -\mu I_s + (1-p_{sn})(1-p_s)\Lambda(t)S + (1-p_{sev})r_cI_c - r_oI_s + (1-p_{rn})(1-p_r)\Lambda(t)I_n - p_r\Lambda(t)I_a + \\ &(1-p_{rn})(1-p_r)p_{rel}\nu L + (1-p_{rn})(1-p_r)\Lambda(t)(R+L) + p_tf(1-ptfc)r_t(T_o + T_o + T_h + T_{oGD} + T_{vGD} + T_{hGD}) \end{split}$$

$$\begin{split} \frac{I_c}{tt} &= -\mu I_c + (1-\tau)p_s\Lambda(t)S + (1-\tau_{sev})(1-\theta_1)r_vI_s - (1-p_{sev})r_cI_c + p_r(1-\tau)\Lambda(t)I_a - p_{sev}r_cI_c + p_r(1-\tau)\Lambda(t)I_n + p_r(1-\tau)P_{rel}\nu L + p_r(1-\tau)\Lambda(t)(R+L) + p_t(1-ptftr)(ptfc)r_t(T_o + T_v + T_h + T_{oGD} + T_{vGD} + T_{bGD}) \end{split}$$

$$\frac{dI_s}{dt} = -\mu I_s - (1 - \tau_{sev})r_s I_s - \tau_{sev} r_Q I_s + p_{sev} r_o I_c$$

$$\frac{dT_o}{dt} = -\mu T_o + (1 - p_{gd})\tau_o p_x \Lambda(t)S - (1 - p_{tf})r_t T_o + (1 - p_{gd})p_r \tau_o \Lambda(t)(I_n + I_a + R + L) + (1 - p_{gd})p_r \tau_o p_{rcl} \nu L$$

$$\frac{dT_v}{dt} = -\mu T_v + (1 - p_{gd})\tau_v p_s \Lambda(t)S - (1 - p_{tf})r_t T_v + (1 - p_{gd})p_r \tau_v \Lambda(t)(I_n + I_s + R + L) + (1 - p_{gd})p_r \tau_v p_{rel} \nu L$$

$$\frac{dT_h}{dt} = -\mu T_h + (1 - p_{gd})\tau_h p_s \Lambda(t)S - (1 - p_{ef})r_t T_h + (1 - p_{gd})p_r \tau_h \Lambda(t)(I_n + I_a + R + L) + (1 - p_{gd})p_r \tau_h p_{rel} \nu L + p_{ef}(ptftr)(ptfe)r_t(T_a + T_v + T_h)$$

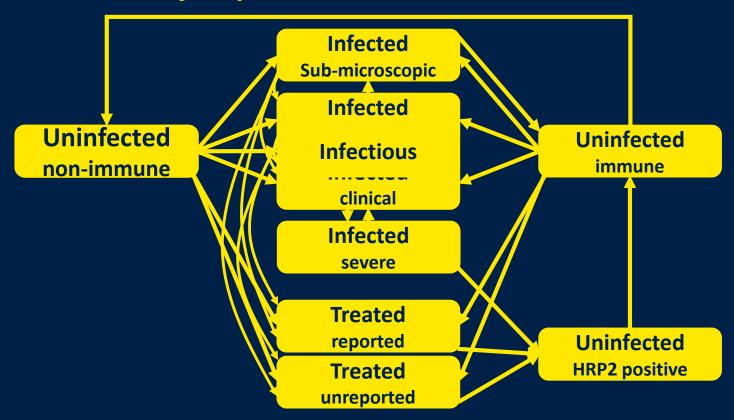
$$\frac{dT_{oGD}}{dt} = -\mu T_{oGD} + p_{gd}\tau_o p_s \Lambda(t)S - (1 - p_{1f})r_1 T_{oGD} + p_{gd}p_r \tau_o \Lambda(t)(I_n + I_a + R + L) + p_{gd}p_r \tau_o p_{rel} \nu L + p_{gd}p_r \tau_o p_{rel}$$

$$\frac{dT_{vGD}}{dt} = -\mu T_{vGD} + p_{gd}\tau_e p_s \Lambda(t)S - (1 - p_{tf})r_t T_{vGD} + p_{gd}p_r \tau_v \Lambda(t)(I_n + I_a + R + L) + p_{gd}p_r \tau_e p_{rel} \nu L$$

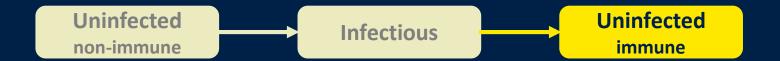
$$\begin{split} \frac{dT_{hGD}}{dt} &= -\mu T_{hGD} + p_{gd}\tau_h p_s \Lambda(t) S - (1 - p_{tf}) r_t T_{hGD} + p_{gd} p_r \tau_h \Lambda(t) (I_n + I_s + R + L) + p_{gd} p_r \tau_h p_{rel} \nu \\ &+ p_{tf} (ptftr) (ptfc) r_t (T_{eGD} + T_{eGD} + T_{bGD}) \end{split}$$

$$\begin{split} \frac{dR}{dt} &= -\mu R + (1-p_h)(1-p_{tf})r_t(T_o + T_v + T_h + T_{oGD} + T_{vGD} + T_{hGD}) + (1-p_h)\tau_{sev}(1-\theta_2)r_QI_{tf} \\ &+ r_nI_n - \Lambda(t)R - \omega R \end{split}$$

$$\frac{dL}{dt} = -\mu L + p_h (1 - p_{tf}) r_t (T_o + T_v + T_h + T_{oGD} + T_{vGD} + T_{hGD}) + p_h \tau_{sev} (1 - \theta_2) r_Q I_s - p_{rel} \nu L - \Lambda(t) L - \kappa I_{total} T_{total} T_{tota$$



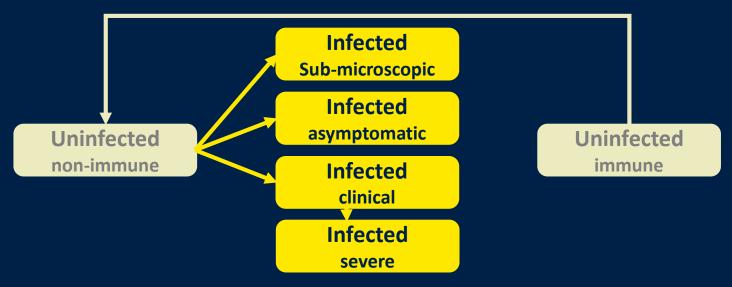
iBdividing anna plaction medideleted with malaria,



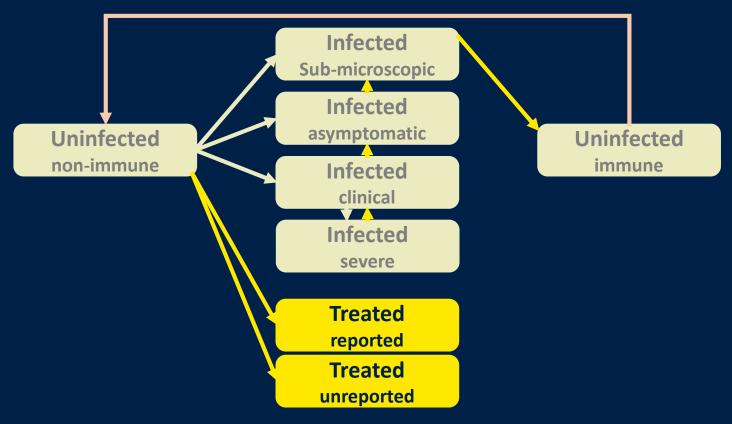
recover naturally and develop immunity



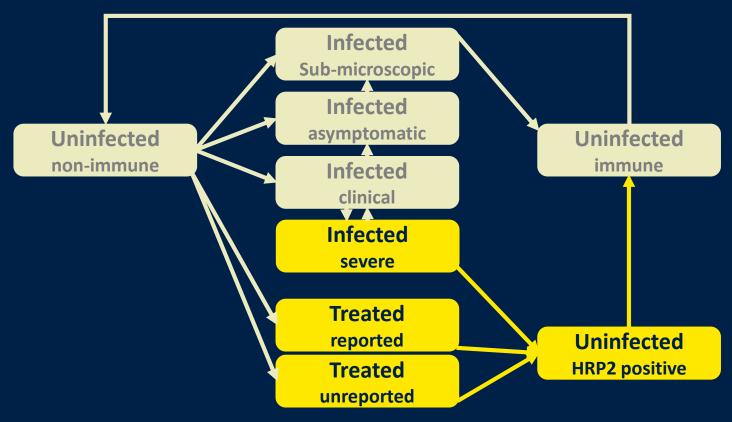
immunity may be lost over time



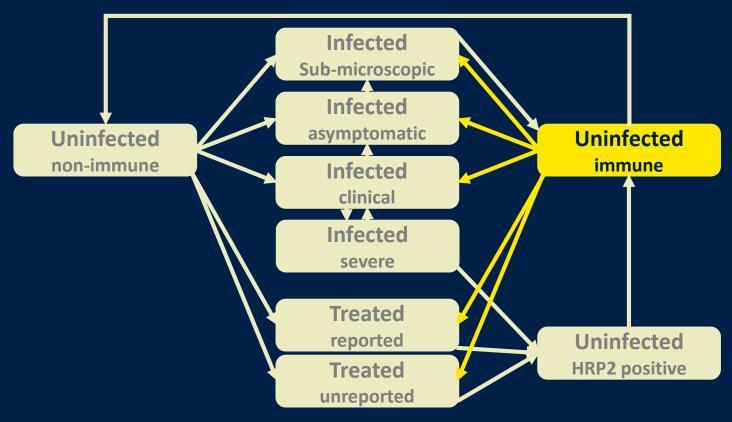
infections have different levels of severity



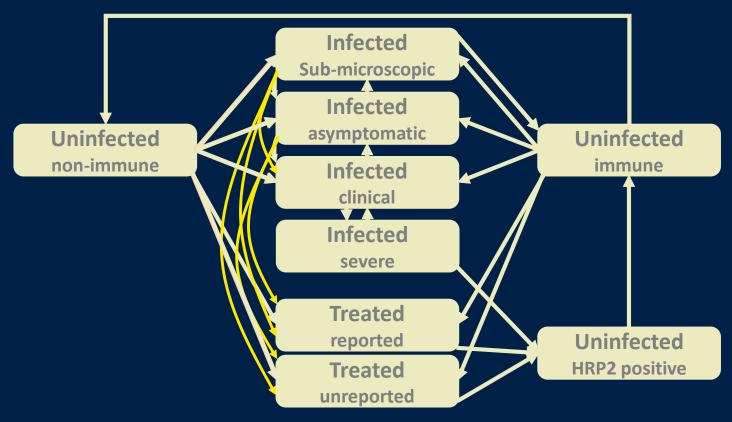
some infections are treated, others recover naturally



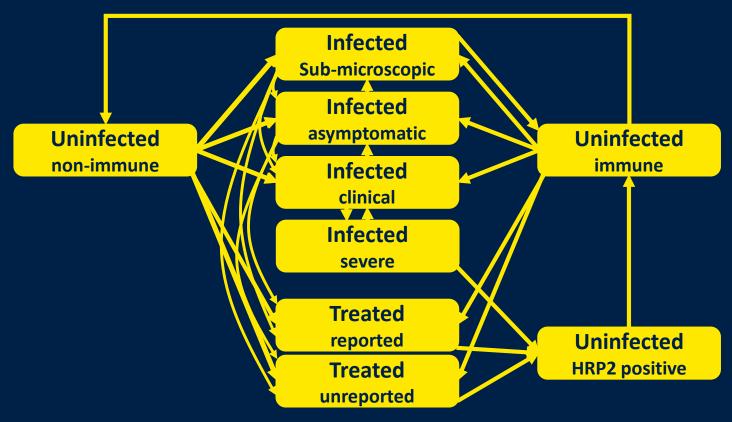
people can still test positive for after successful treatment



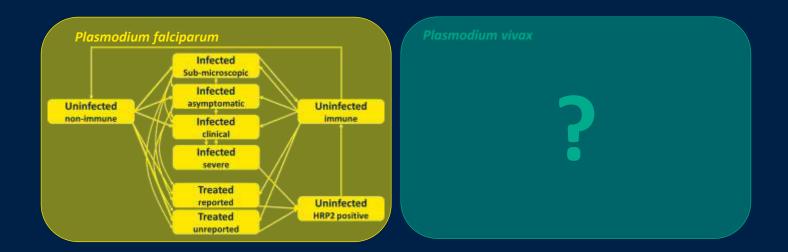
immune people get infected, but are less likely to have symptoms



more than one infection at the same time (superinfection)

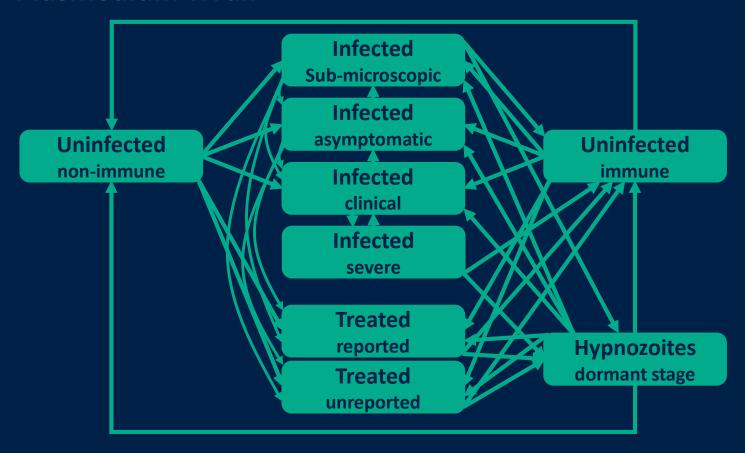


the model combines all these processes in one framework



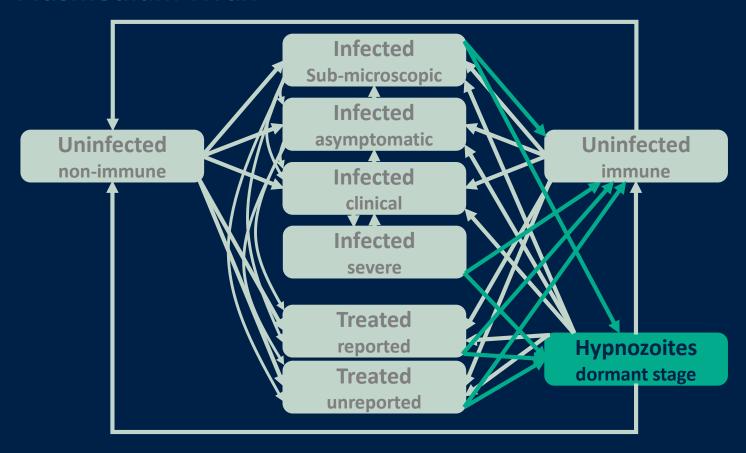
but the Asia-Pacific has two major malaria species

Plasmodium vivax



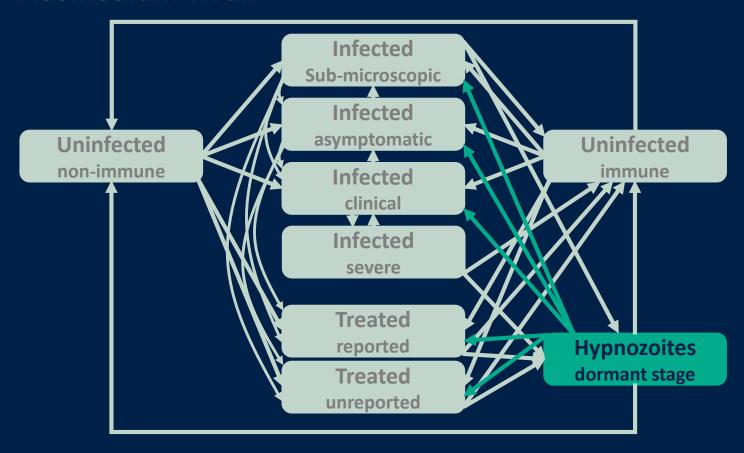
The model for P. vivax is different from the model for P. falciparum

Plasmodium vivax

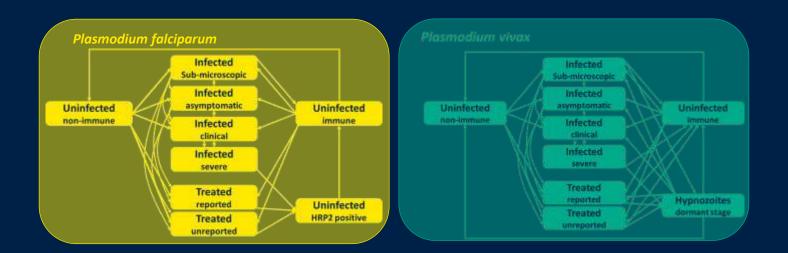


recovery may result in the dormant stage (hypnozoites)

Plasmodium vivax



hypnozoites are a source of relapse months after the initial infection

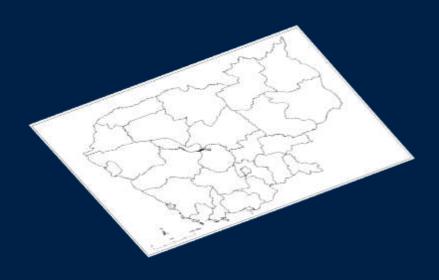


the transmission and control of both species is simultaneous

Plasmodium falciparum

Plasmodium vivax

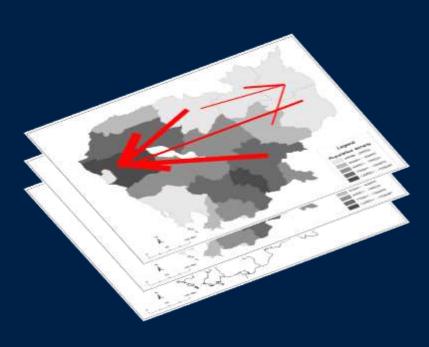
the species interact



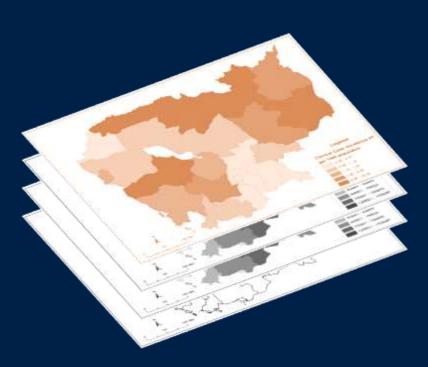
 Start with the geographic scope and define the subpopulation or "patch"



 Include data on population distribution by patches



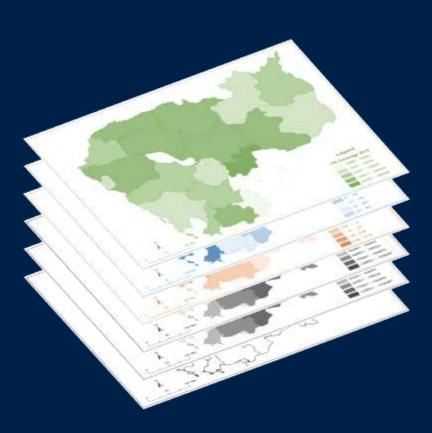
Include some
 assumptions or data
 on connectivity
 between patches



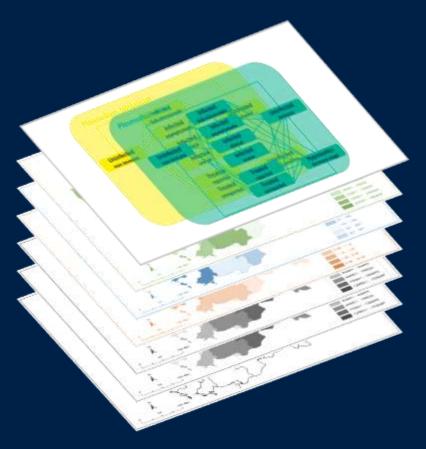
 Include surveillance data on spatial heterogeneity in transmission



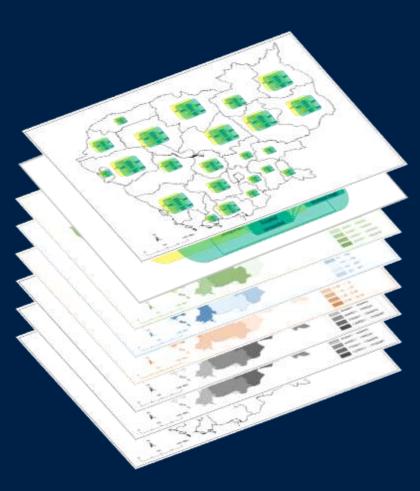
 Include data on historical and planned clinical case management



 Include data on historical and planned vector control



Create a model

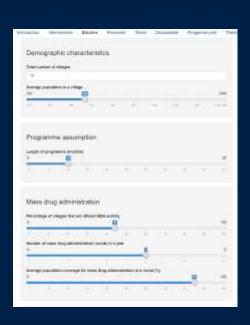


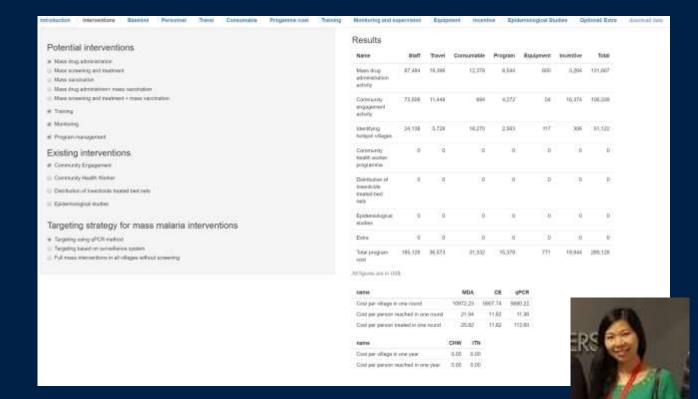
- Replicate the compartmental model for each patch
- link the submodels using connectivity assumption or data
- Include the historical intervention data
- Fit the model to surveillance data
- Include costs

~1000 lines of code

```
108 # 4=Ic: infected clinical
109 # 5=Is; infected severe
110 # 5=To; treated unrecorded
111 # 7wTv: treated VMW
112 # 8-Th; treated HIS
113 # 9-R: uninfected immune
114 # 10-H: uninfected and HRP2 positive
115 #
116 | 1shumpop<-1:10
117
121 varind -- matrix (0, nrow-8, ncol-N)
122 traind<-matrix(0,nrow=A,ncol=N)
123 - for (n in 1:N)
124+ for (b in 1:8) (
      varind[b,n]<-(n-1)*8+b
126
127 -
    for (a in 1:A) (
      traind[a,n]<-(n-1)"A-a
129
130
131
135 # first transition is given without index
136 transitions =ssa.maketrans(V.rbind(varind[4,1], +1)) # birth mos patch 1
137+ for (n in 1:N) (ma)
190 #Alternate formulation of transitions matrix (used in epimodel function)
191 transitions2<-NULL
192 - for (1 in 1: length(transitions))(
193 transitions2~rbind(transitions2,cbind(as.integer(names(transitions[[1]]))[1],as.integer(names(transitions[[1]]))[2], transitions[[1]][1], transitions[[1]][2]))
194
195 row.names(transitions2) <- MULL
196 transitionsiul -- transitions2[,1]
197 transitionsiu2 -transitions2[,2]
198 transitionsivi - transitions2[.3]
199 transitionsiv2 transitions2[,4]
200
201
203 # Set the parameters
205 pars = 1ist(
    # Nosquittees
     Peax-40000 seq_len(N)/seq_len(N), # the maximum populations of mosquitoes in each patch
     delta_m=365.25/14, # death rate of mosquitoes
```

Estimate the cost of future interventions





World Health Organization Global Technical Strategy for Malaria 2016-2030

Pillar 1		Pillar 2	Pillar 3
Ensure universal access to malaria prevention, diagnosis and treatment		Accelerate efforts towards elimination and attainment of malaria-free status	Transform malaria surveillance into a core intervention
Supporting Element 1	Harnessing Innovation & Expanding Research		
Supporting Element 2	Strengthening the Enabling Environment		

Pillar 3

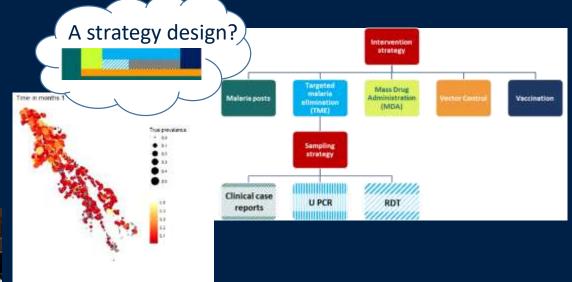
Transform malaria surveillance into a core intervention

Smartphones for Community Health in Rural Cambodia





Pengby Ngor, CNM



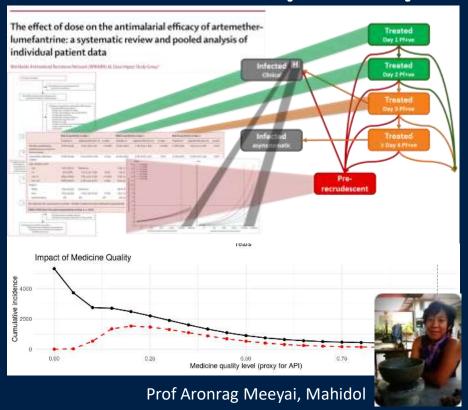


World Health Organization Global Technical Strategy for Malaria 2016-2030

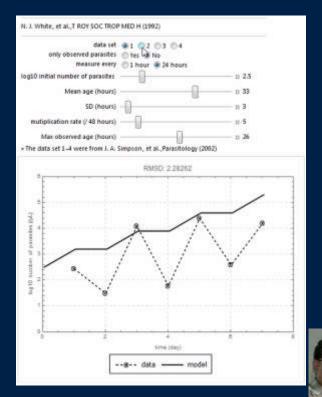
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Harnessing Innovation & Expanding Research

Medicine quality



Within-host models



Dr Sompob Saralamba, MORU

World Health Organization Global Technical Strategy for Malaria 2016-2030

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Supporting Element 2

Strengthening the Enabling Environment

Which is faster?

1 modeller learns 5 cultures and health systems?
 OR

• 5 people learn modelling?

Individual and institutional capacity building in modelling



Conclusions

- Research
 - Understanding the context
 - Developing the best methods
 - Economic features
- Communication
 - Stakeholders, collaborators, policymakers, funders
 - Participatory modelling
 - User-friendly tools and interfaces
- Training
 - Modellers leading groups in endemic settings
 - Modellers providing training to stakeholders



Acknowledgements

- Mahidol Oxford Tropical Medicine Research Unit
 - Mathematical and Economic Modelling group
 - Arjen M Dondorp
 - Nicholas J White
 - Nicholas P J Day
 - Paul Newton
- Shoklo Malaria Research Unit
 - Francois Nosten
 - Rose McReady
 - METF team
- National malaria control programs
 - Cambodia, Myanmar, Bangladesh and Lao PDR
- Wildecarde Productions Ltd
 - Ben Burgess



- Modelling And Simulation Hub, Africa (MASHA), UCT
 - Sheetal Silal
- University of California San Francisco
 - Rima Shretta
- Work funded by
 - BMGF, Wellcome, APLMA, ADB

Any questions?