Sensitivity Analysis of Deterministic Models Introduction to Latin Hypercube Sampling

John M. Drake & Pejman Rohani

<ロト <回ト < 注ト < 注ト = 注

Key questions about uncertainty:

- Which factors are most important to determining model behavior?
- How much will my result change if the conditions (parameters, initial state) change?
- What range of outcomes are consistent with my knowledge of the parameters?
- What processes do I need more information about and how much information do I need

In simple, deterministic models with few state variables and few parameters we can often produce model visualizations to answer such questions.



Invasion boundary for a model of Ebola virus transmission

But, what do we do with more complicated models?

-∢ ⊒ →

- In 2000, \approx 30% of gay men in San Francisco were infected with HIV, \approx 50% of these were taking combination antiretroviral therapy (ART)
- ART was effective at reducing AIDS death rate, but does not completely eliminate infectivity
- It was unclear whether the net effects of increased distribution of ART would increase or decrease HIV



What is the effect of antiretroviral therapy on incidence of HIV?



Blower, S.M., et al. 2000. A tale of two futures: HIV and antiretroviral therapy in San Francisco. Science 287:650-654.

- X Susceptible
- Y Infected (R=resistant, S=sensitive; U=untreated, T=treated)
- π Rate at which gay men join the sexually active community
- μ^{-1} Average time during which new partners are acquired
 - c Average number of new partners per year
 - p Probability of a drug-resistant case transmitting drug-sensitive viruses
- q^{-1} Average time for a drug-resistant infection to revert to drug-sensitive infection
 - σ Per capita effective treatment rate
 - e Relative efficacy of ART in treating drug-resistant infections
 - r Rate of emergence of resistance due to acquired infection
 - g Proportion of cases that give up ART per year
 - ν Average rate of disease progression

• = • • = •

э

- It appears that ART could prevent \approx 15,000 cases over 20 years
- How reliable is this result?
- Model has 20 parameters but none is known exactly



Number of infections prevented as a function of the fraction of cases treated

Latin hypercube sampling



- To determine robustness of model predictions, we require a way of exploring the output of a family of parameterized models
- Realistic models will often have many parameters so that high resolution exploration of its parameter space is computationally intractable
- Latin hypercube sampling is a scheme for simulating random parameter sets that adequately cover the parameter space.

▶ < ∃ >

Latin hypercube sampling in R

> require(lhs) > x <- runif(50) > y <- runif(50) > h <- 50 > lhs<-maximinLHS(h,2) > par(mfrow=c(1,2)) > plot(x,y,type='p', main='Random Uniform', xlab='', ylab='') > plot(lhs, type='p', main='LH Sampling', xlab='', ylab='')







Latin hypercube sampling in R (3-D)

```
> require(scatterplot3d)
```

- > x <- runif(50); y <- runif(50); z <- runif(50)
- > h <- 50
- > lhs<-maximinLHS(h,3)
- > par(mfrow=c(1,2))
- > scatterplot3d(x,y,z, type='p', main='Random Uniform', xlab='', ylab='', zlab='
- > scatterplot3d(lhs, type='p', main='LH Sampling', xlab='', ylab='', zlab='')

Random Uniform

LH Sampling





- Need to re-scale the random point (a number in the interval [0,1]) to an interval from α_{min} to α_{max} where α is some parameter.
- This can be done by "stretching" the interval using the following formula

$$\alpha_0 = U(\alpha_{max} - \alpha_{min}) + \alpha_{min}$$

- Evidently, our best guesses are rather optimistic compared with the range of scenarios we believe to be plausible
- At least ART was found never to be counter-productive (an open question at the time of this study)



Range of outcomes as a function of fraction treated

- Correlation analysis can be used to investigate how model output is related to input parameters (but does not account for covariances among parameters, if there are any)
- Partial rank correlation coefficients partition effects to each input variable



Partial rank correlation coefficients of 20 parameters

< ∃ →

- A key problem is to distiguish variability that arises from intrinsic stochasticity and uncertainty that can be mitigated through the acquisition of better information
- The effect of uncertainty in model parameters can be identified through Latin Hypercube Sampling coupled with Partial Rank Correlation analysis
- Other methods (e.g. Sobol's Index, Sensitivity Heat Map) may be used to determine the effects of parameter interactions or direction of effect

For further reading: Wu et al. 2013. Sensitivity analysis of infectious disease models: methods, advances and their application. *Journal of the Royal Society Interface* 10:20121018.

э