

FUNNEL: Automatic Mining of Spatially Coevolving Epidemics

Yasuko Matsubara, Yasushi Sakurai (Kumamoto University)

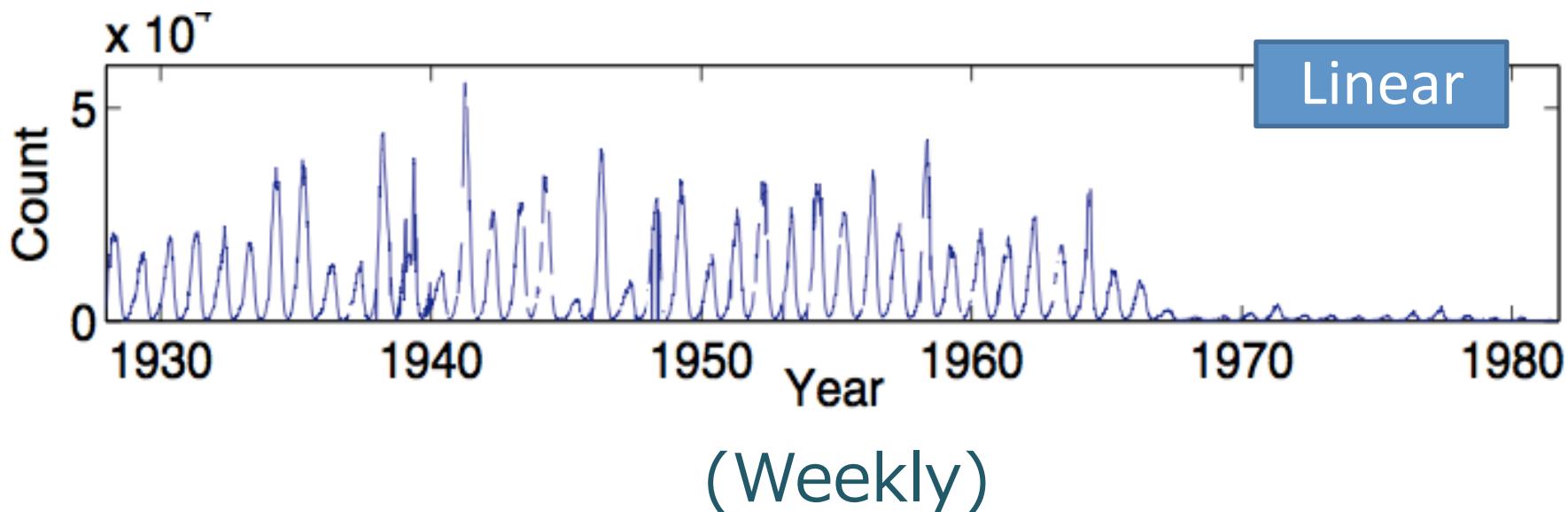
Willem G. van Panhuis (University of Pittsburgh)

Christos Faloutsos (CMU)



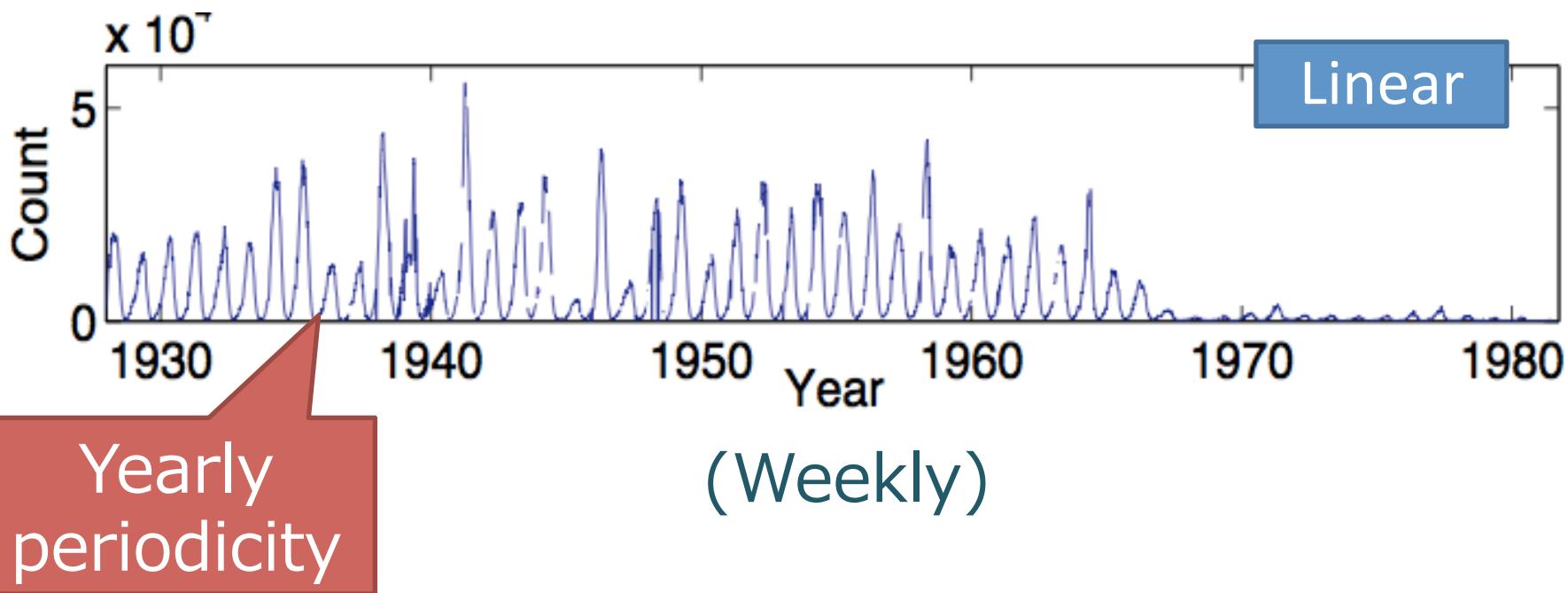
Motivation

Given: Large set of epidemiological data
e.g., Measles cases in the U.S.



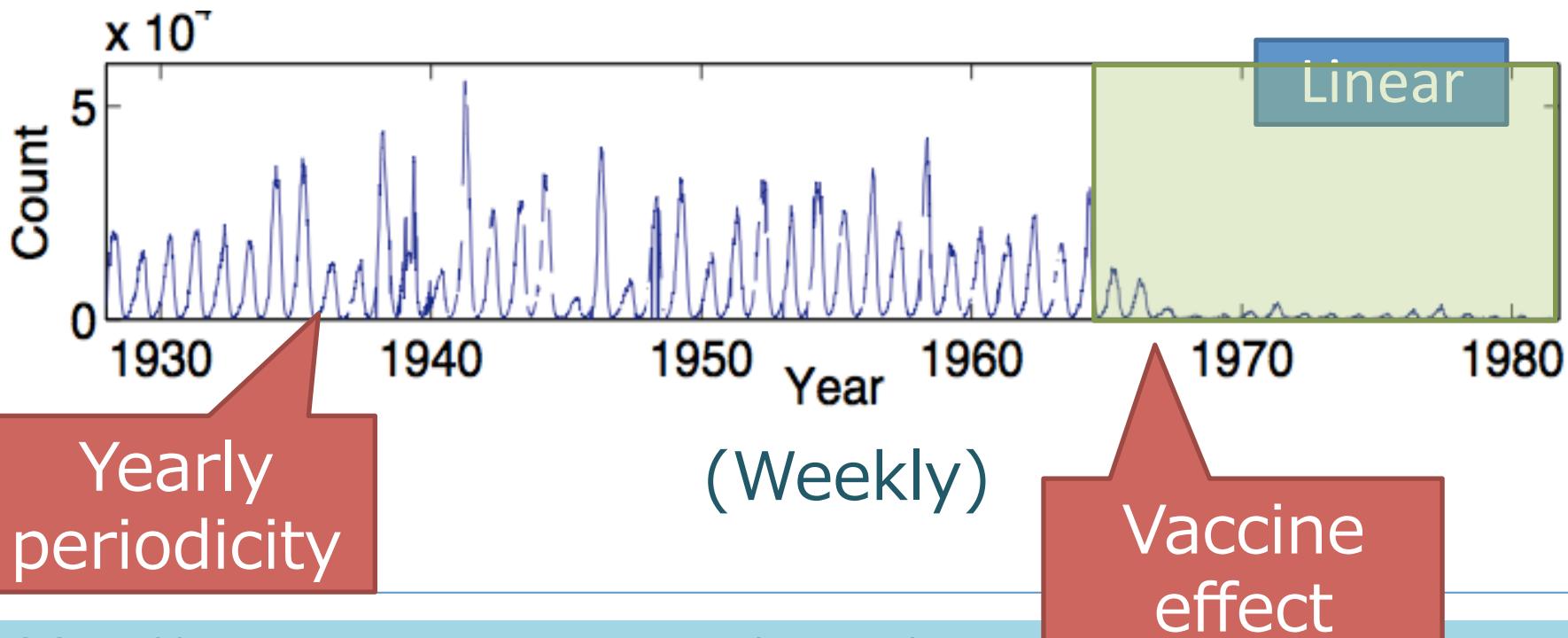
Motivation

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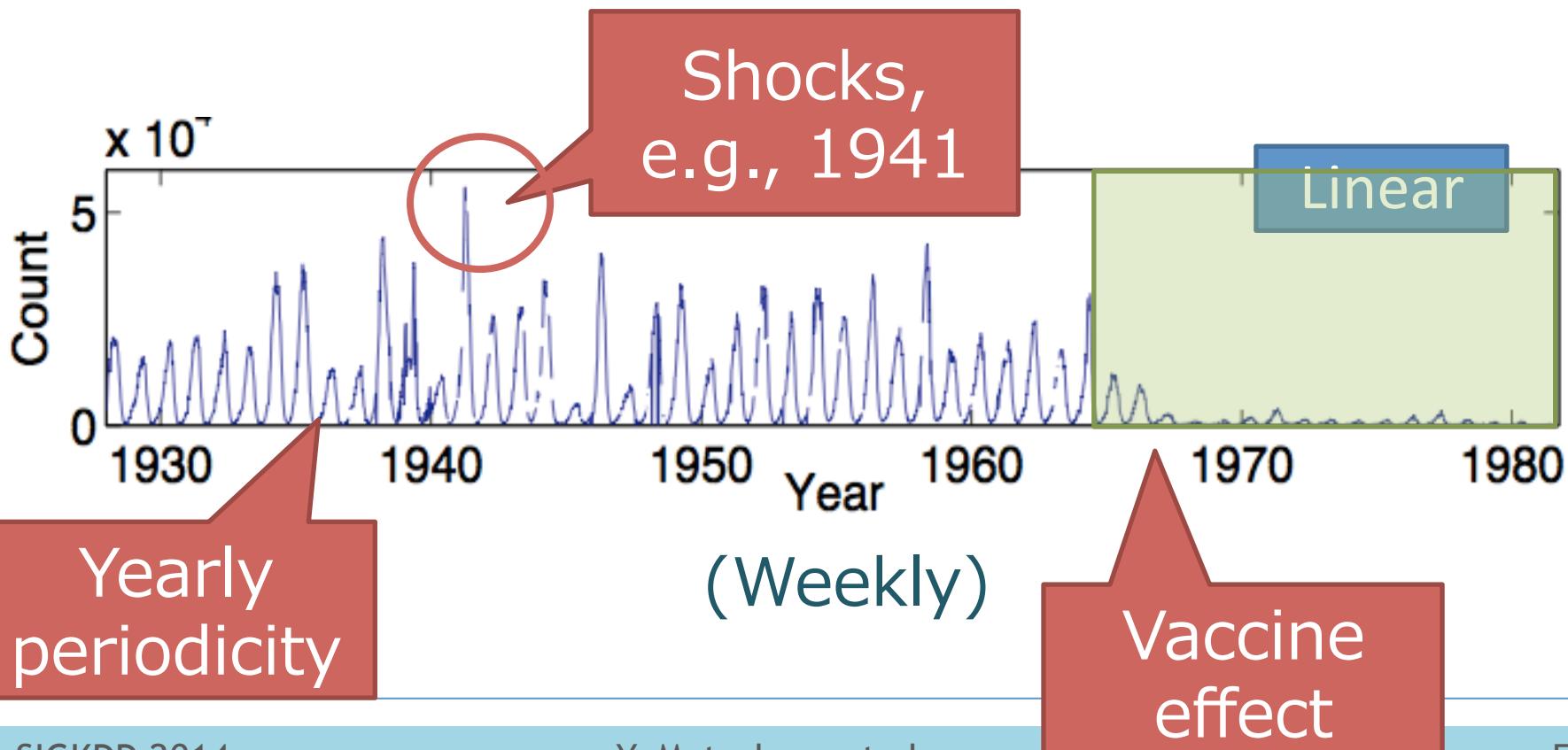
Motivation

Given: Large set of epidemiological data
e.g., Measles cases in the U.S.



Motivation

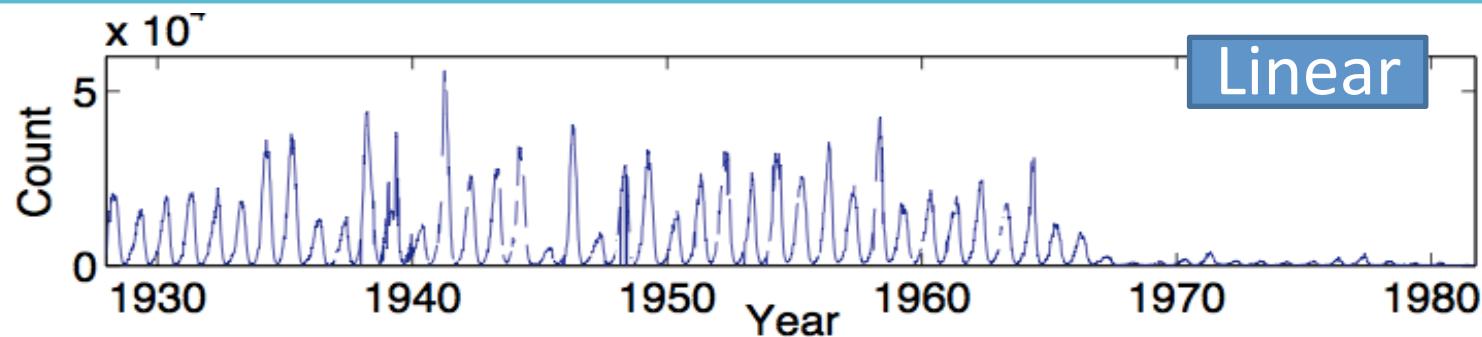
Given: Large set of epidemiological data
e.g., Measles cases in the U.S.



Motivation

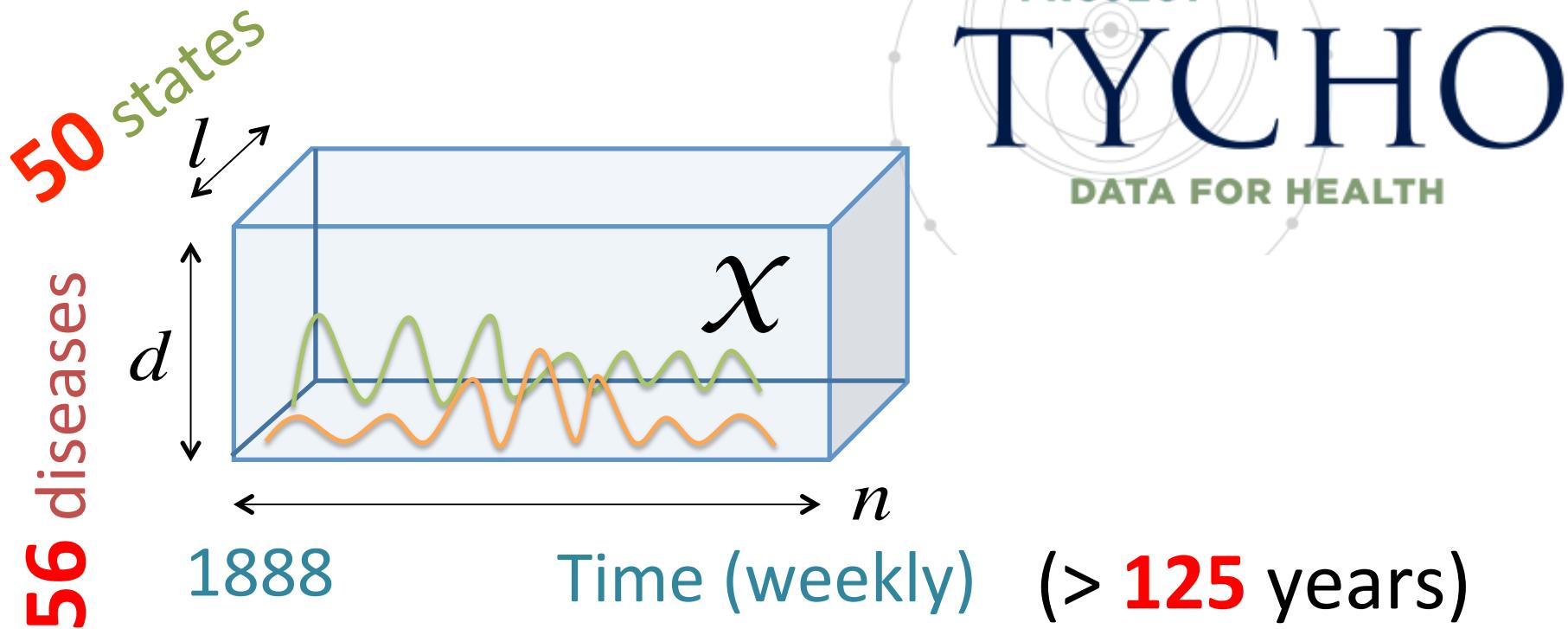
Given: Large set of epidemiological data
e.g., Measles cases in the U.S.

Goal: summarize all the epidemic time-series, “fully-automatically”



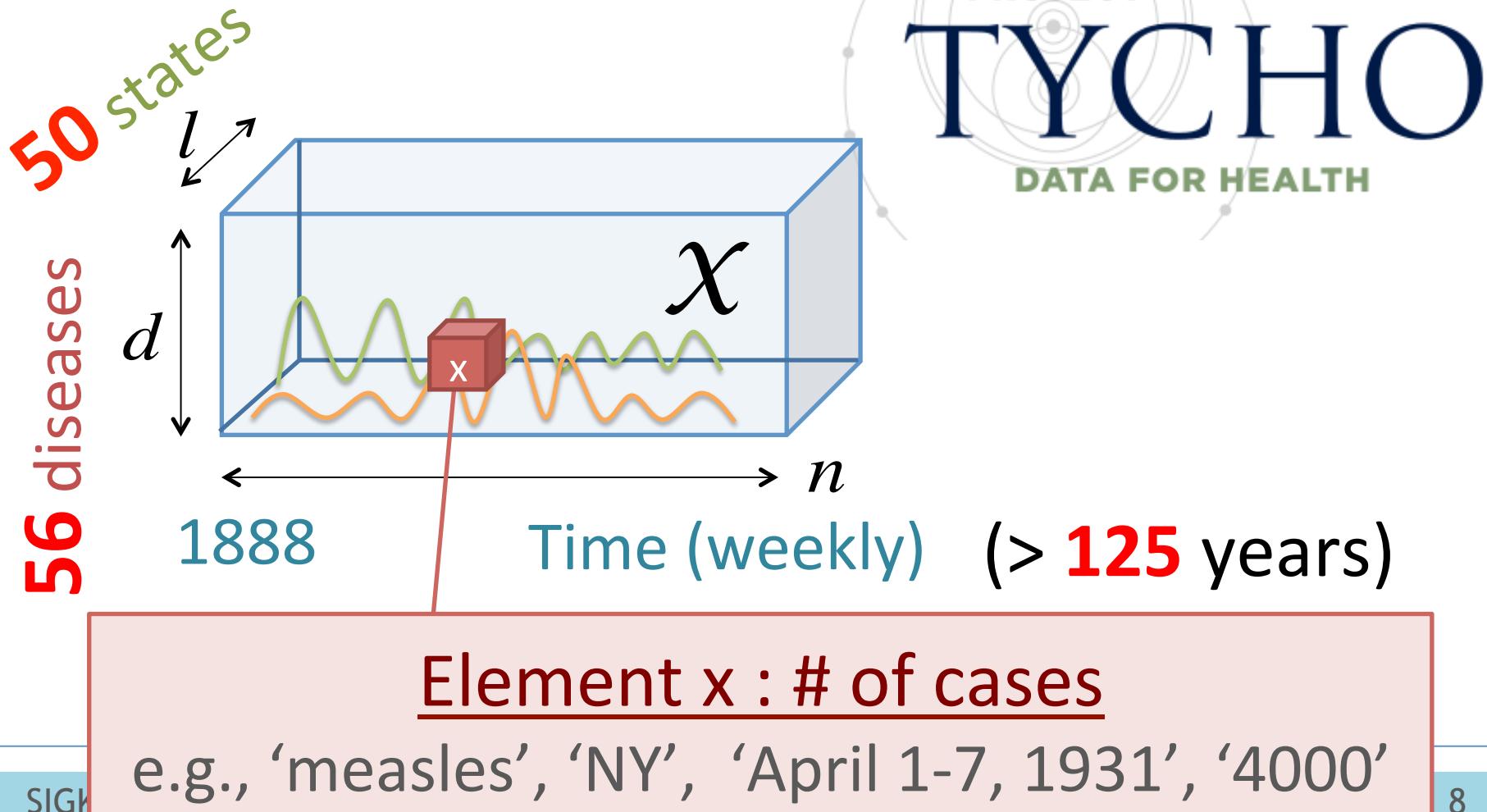
Data description

Project Tycho: infectious diseases in the U.S.



Data description

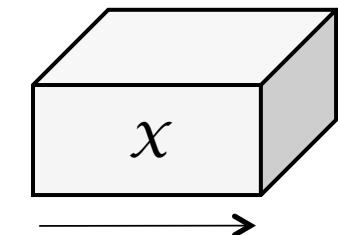
Project Tycho: infectious diseases in the U.S.



Problem definition

Given:

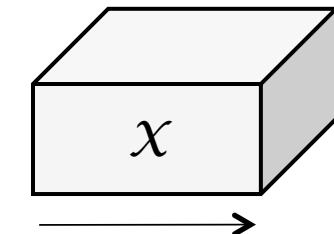
Tensor \mathcal{X} (**disease** x **state** x **time**)



Problem definition

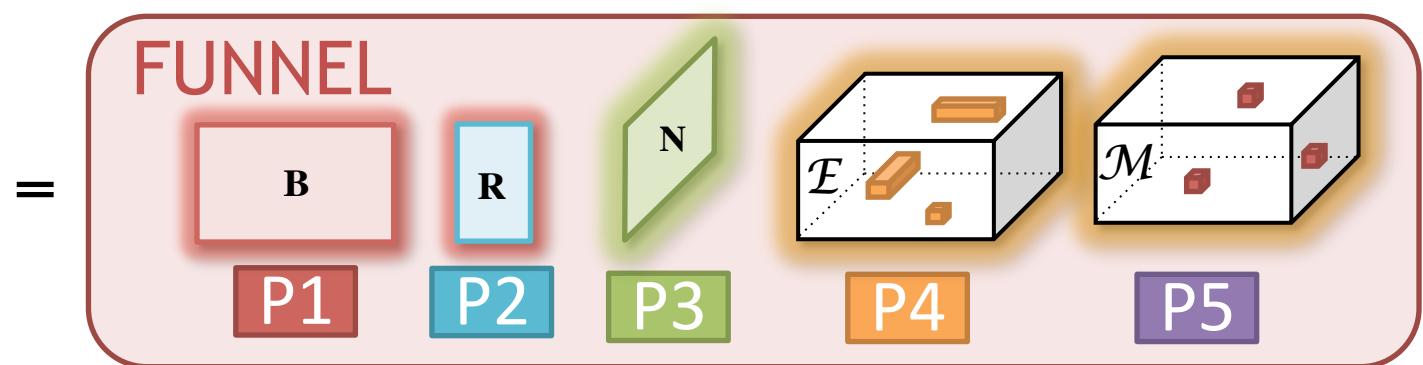
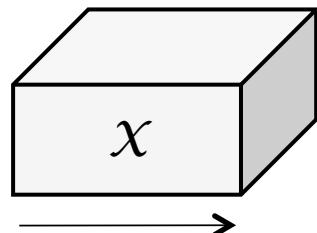
Given:

Tensor \mathcal{X} (**disease** x **state** x **time**)



Find:

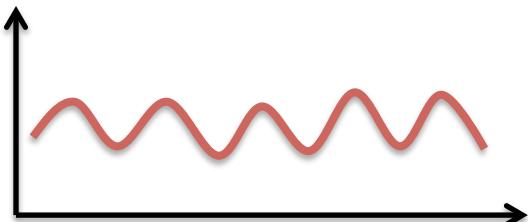
Compact description of \mathcal{X} , “*automatically*”



Problem definition

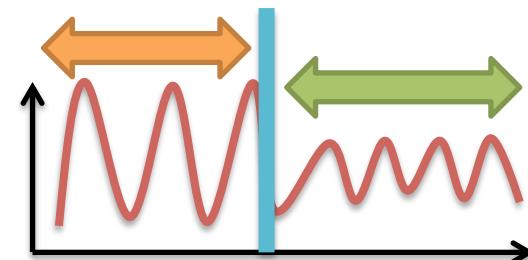
Gr
Te
Fi

Seasonality



state

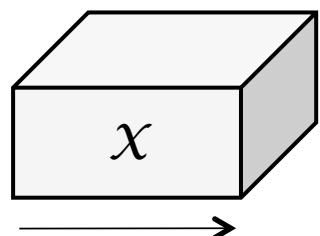
Discontinuities



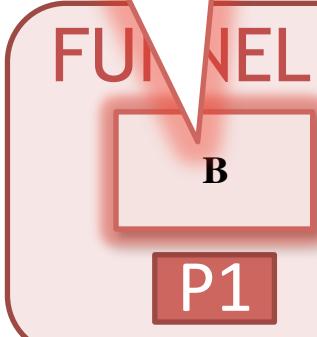
Compact desc

ption of χ ,

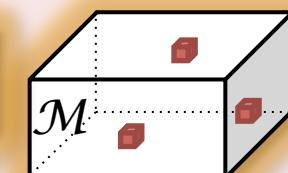
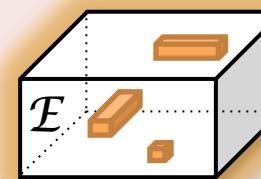
“automatically”



=



N



P4

P5

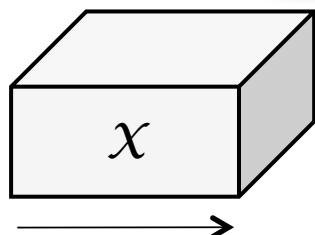
Problem definition

Given:

Tensor

Find:

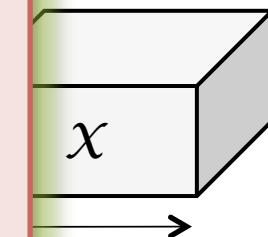
Compact



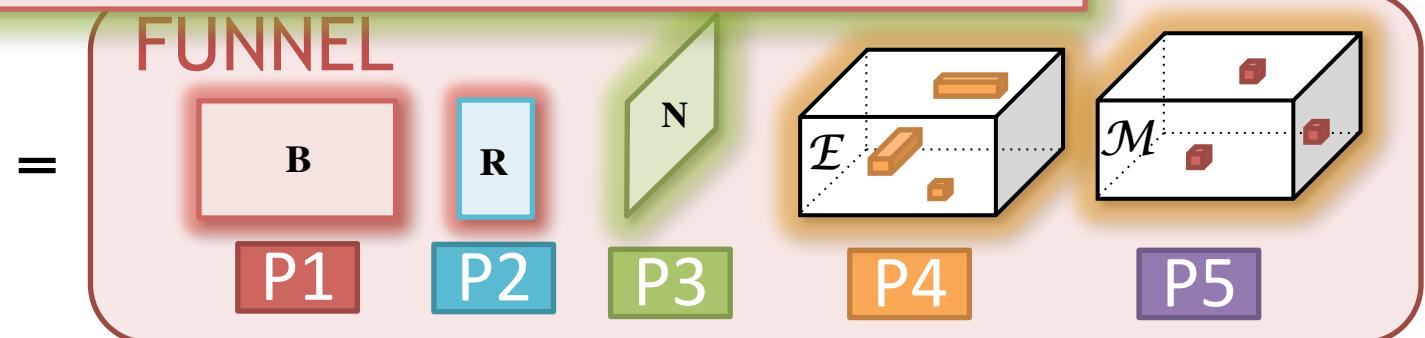
NO magic numbers !



Parameter-free!

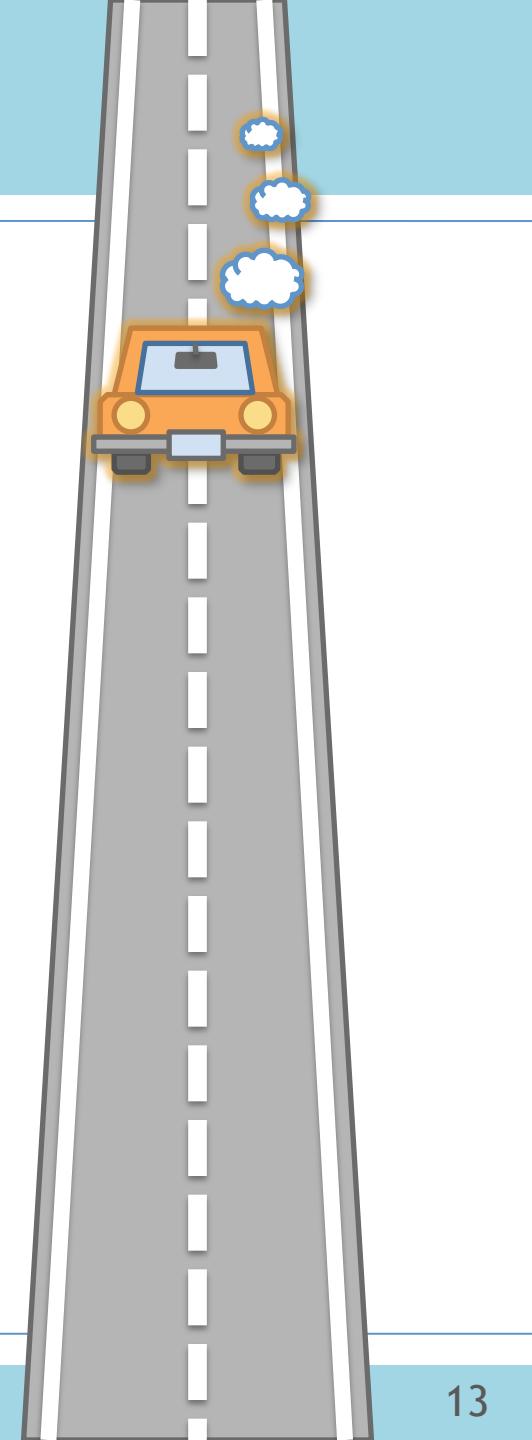


“*rically*”



Roadmap

- ✓ Motivation
- Modeling power of FUNNEL
- Overview - main ideas
- Proposed model - idea #1
- Algorithm - idea #2
- Experiments
- Discussion
- Conclusions



Modeling power of FUNNEL

Questions about epidemics

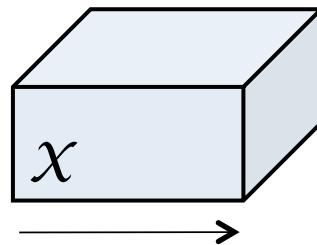
Q1

Q2

Q3

Q4

Q5



Questions

Q1

Q2

Q3

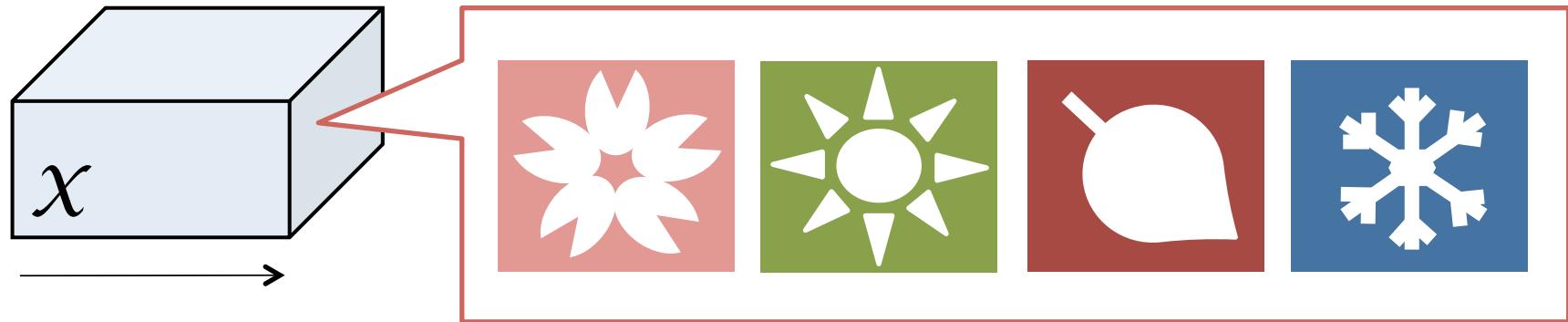
Q4

Q5

Q1

Are there any periodicities?

If yes, when is the peak season?



Answers

Q1

Q2

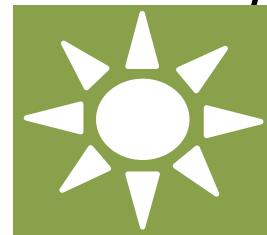
Q3

Q4

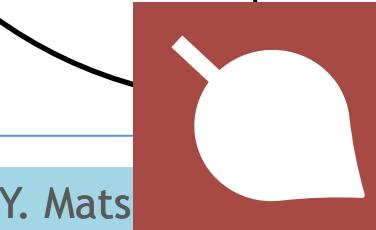
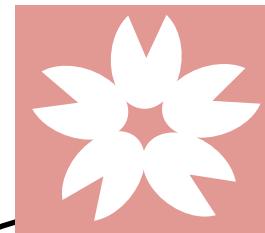
Q5

P1 Seasonality

FUNNEL:
Polar plot



Angle:
peak season



Y. Mats

Y. Mats

Radius:
seasonality
strength

Y. Mats

Answers

Q1

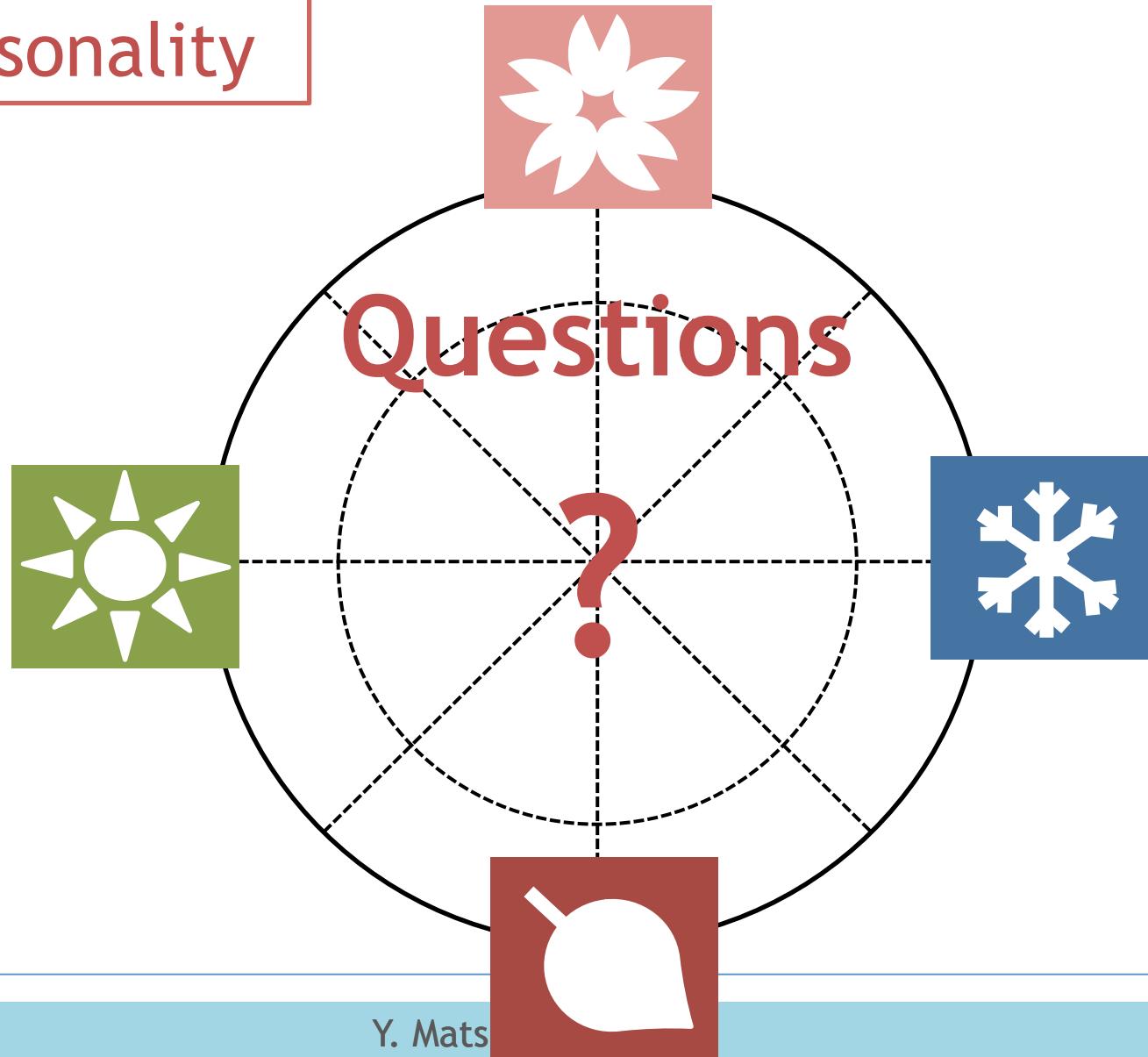
Q2

Q3

Q4

Q5

P1 Seasonality



Answers

Q1

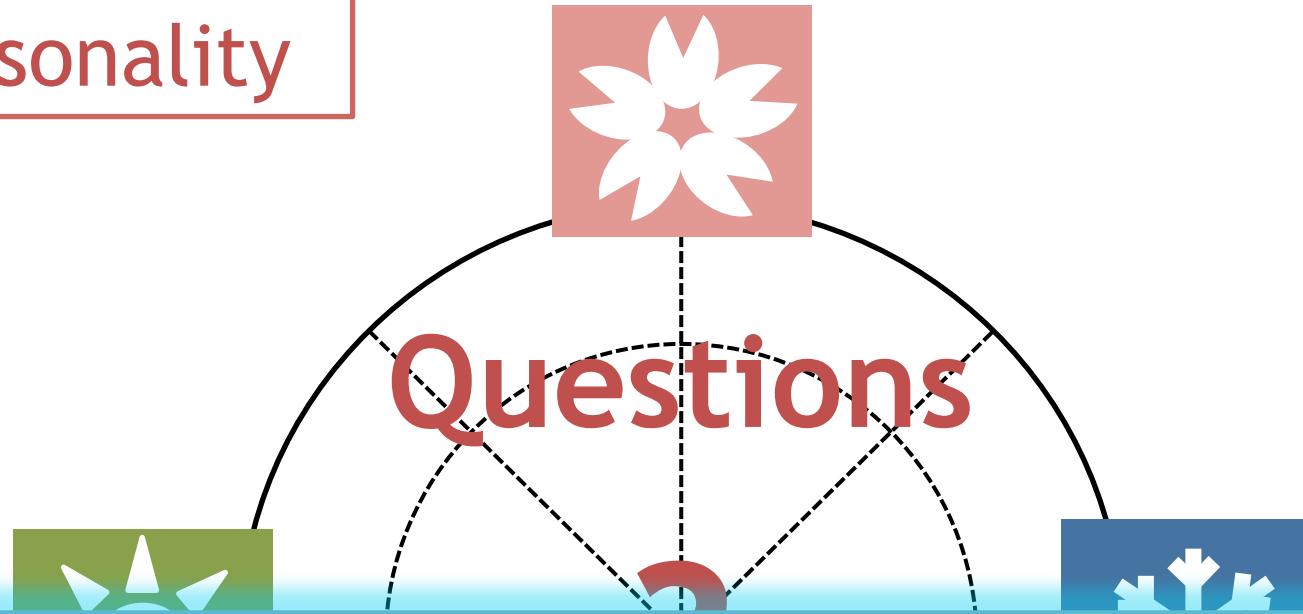
Q2

Q3

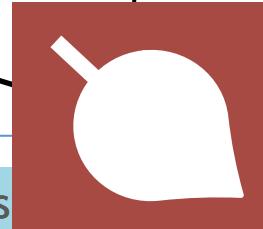
Q4

Q5

P1 Seasonality



Q: Does Influenza have seasonality? If yes, when?



Answers

Q1

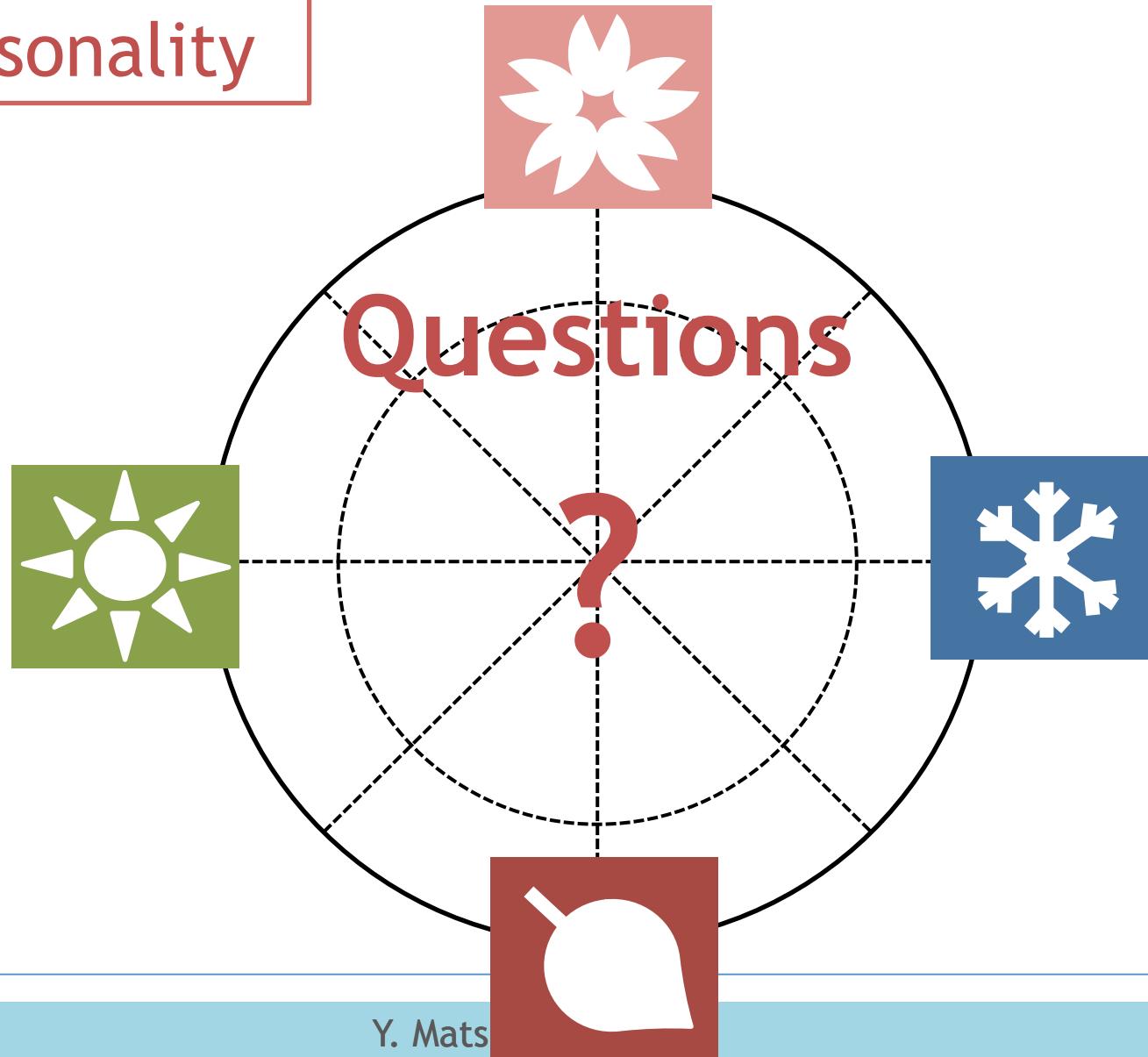
Q2

Q3

Q4

Q5

P1 Seasonality



Answers

Q1

Q2

Q3

Q4

Q5

P1 Seasonality

Influenza in Feb.
Detected by FUNNEL
(strong seasonality)

Rocky m.



Red fever

Lymedisease

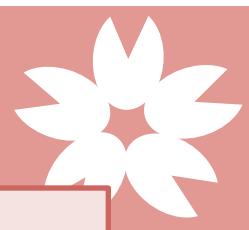
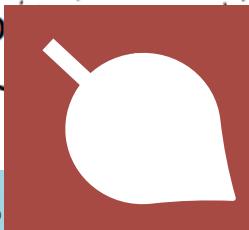
August (8)

Typhus fever

Crypt

September (9)

Y. Mats



Measles
0.4

0.3

Mumps

Chickenp

March

Detected!

February (2)

Influenza

Streptococ

Gonorrhea

Typhoidfever

(1)

December (12)

November (11)



Answers

Q1

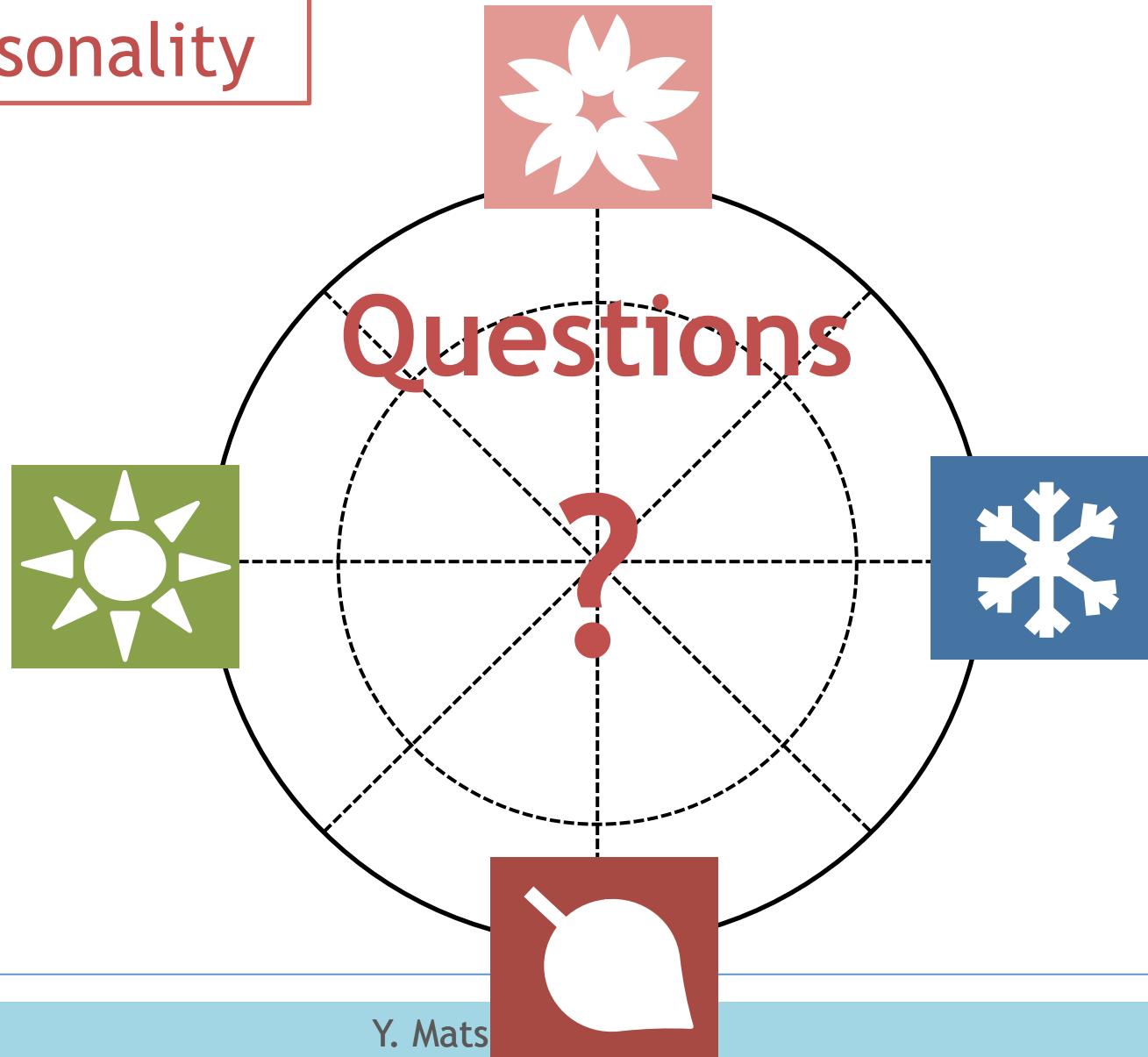
Q2

Q3

Q4

Q5

P1 Seasonality



Answers

Q1

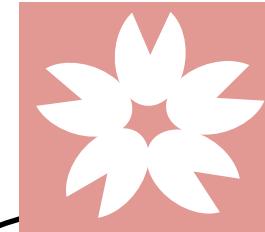
Q2

Q3

Q4

Q5

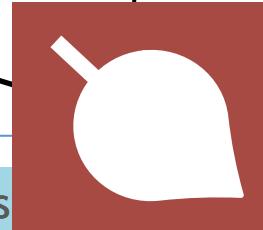
P1 Seasonality



Questions



Q: How about measles ?



Answers

Q1

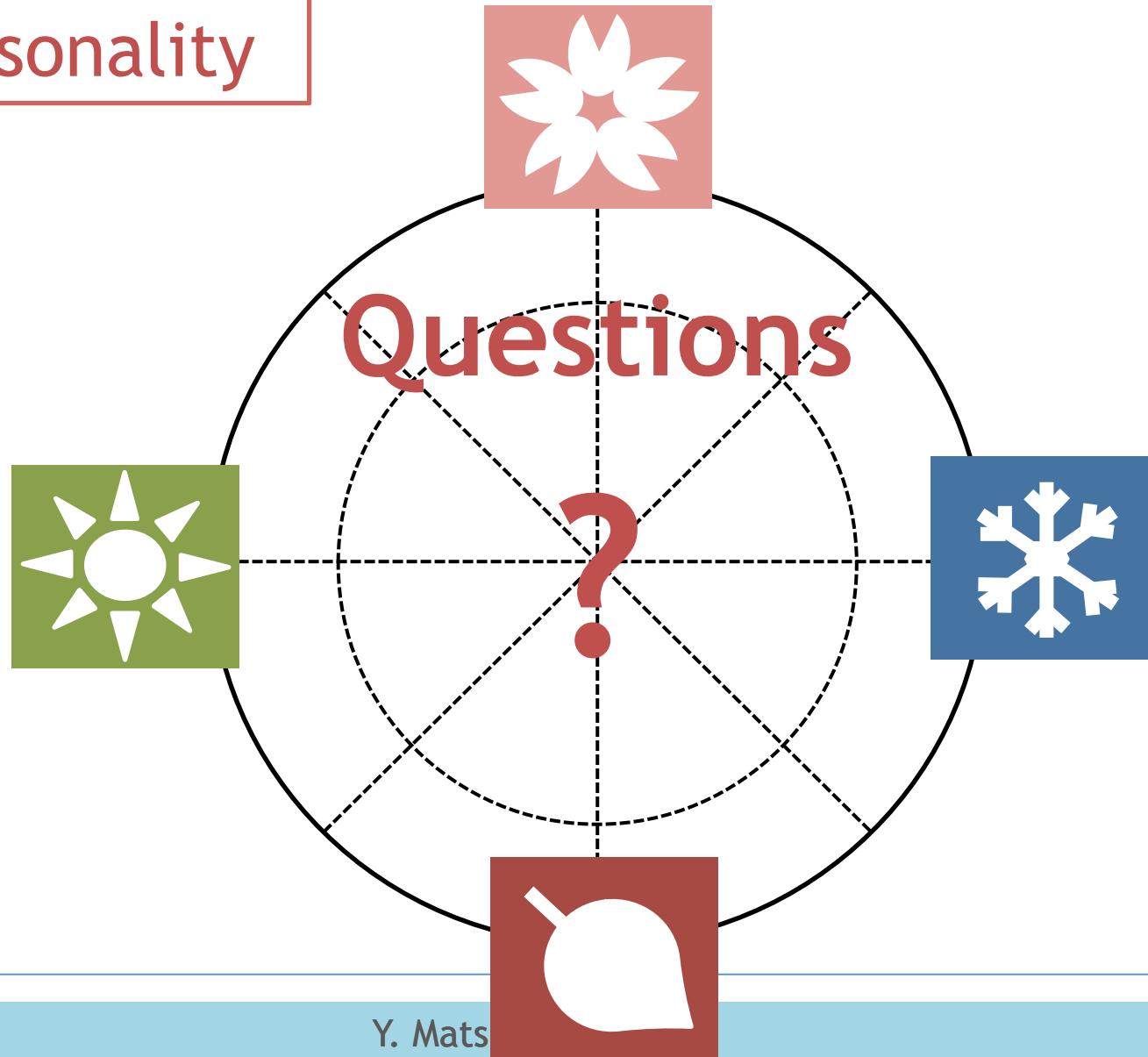
Q2

Q3

Q4

Q5

P1 Seasonality



Answers

Q1

Q2

Q3

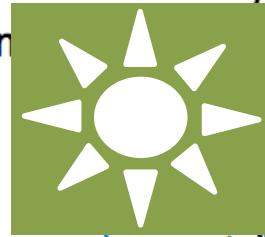
Q4

Q5

P1 Seasonality

Measles
(children's)
in spring

Rocky m.



August (8)

May (5)

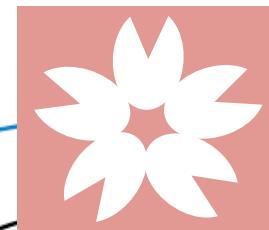
Red fever

Lymedisease

Typhus fever

Crypt

September (9)



Measles 0.4

Rubella 0.3

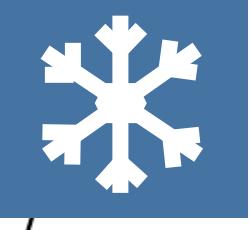
Smallpox 0.2

Measles 0.1

Measles 0.05

Detected!

February (2)



(1)

Chickenpox

Influenza

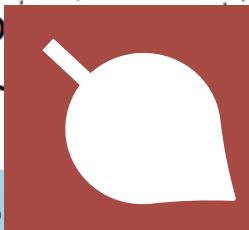
Streptococ

Gonorrhea

Typhoidfever

December (12)

November (11)



Y. Mats

Answers

Q1

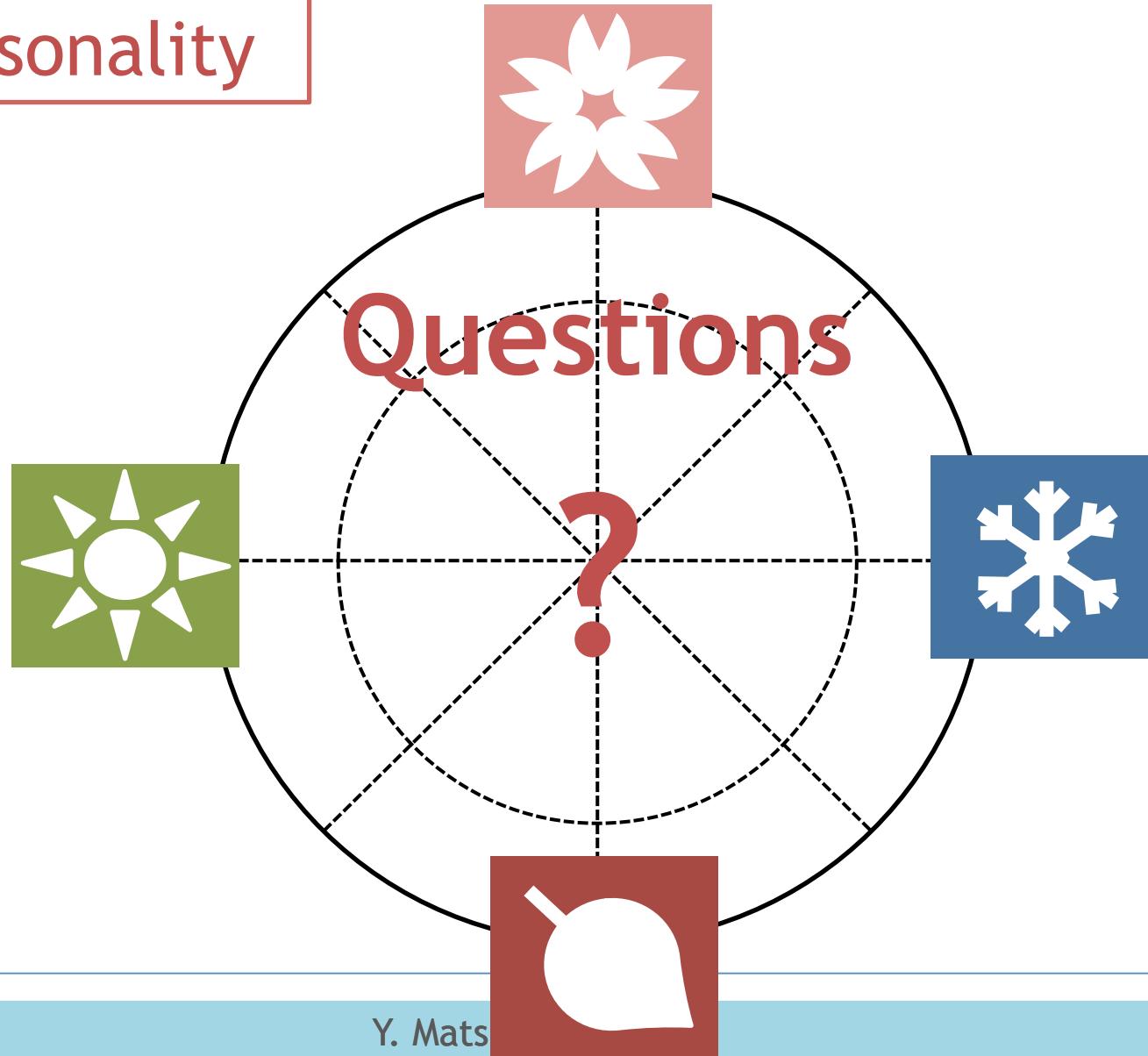
Q2

Q3

Q4

Q5

P1 Seasonality



Answers

Q1

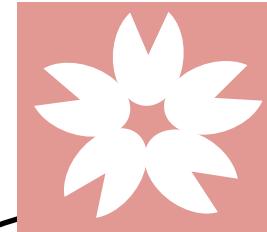
Q2

Q3

Q4

Q5

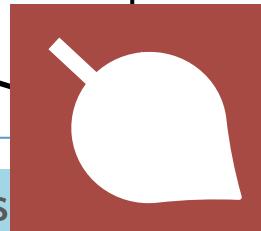
P1 Seasonality



Questions



Q: Which disease peaks in summer?



Answers

Q1

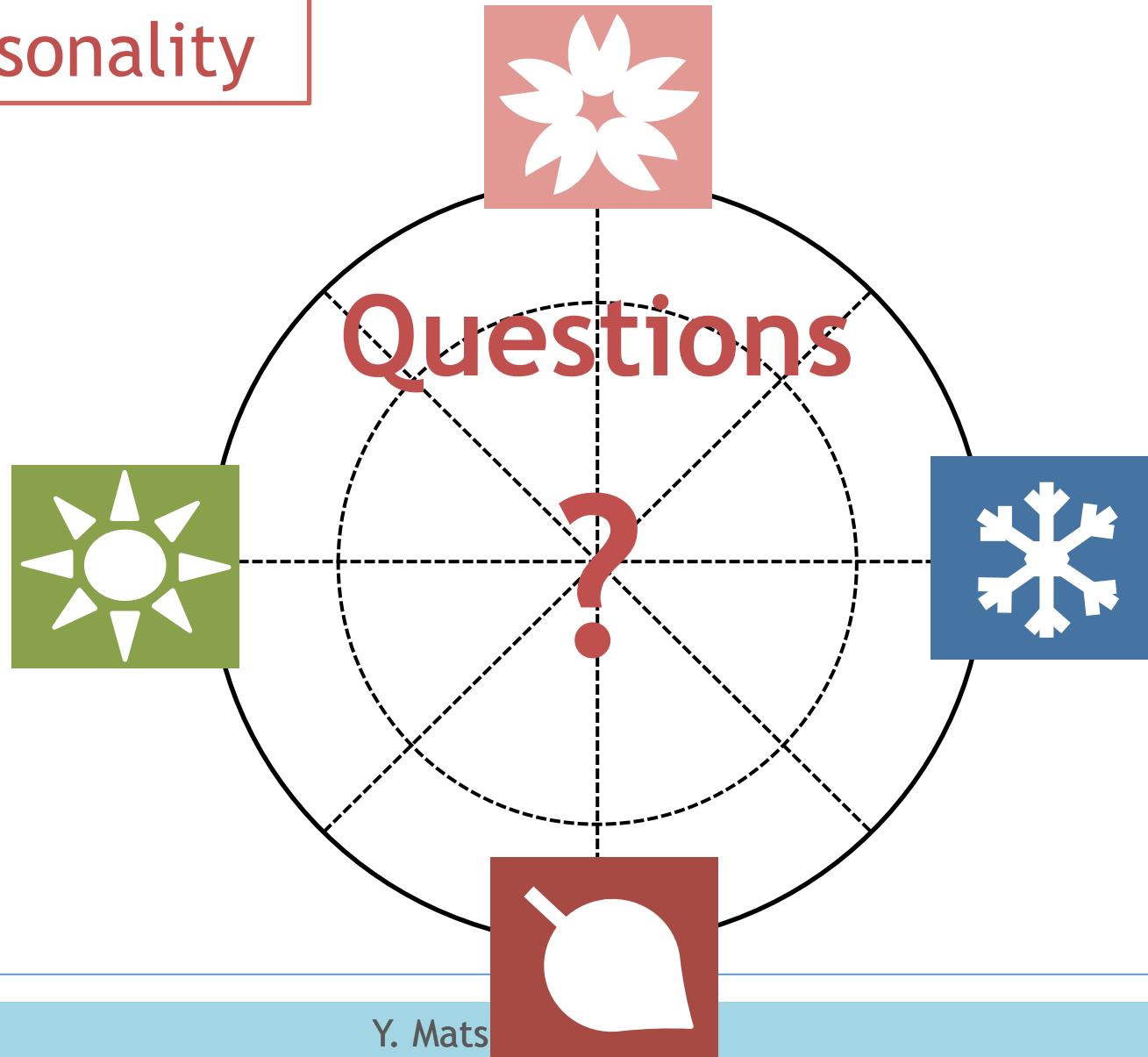
Q2

Q3

Q4

Q5

P1 Seasonality



Answers

Q1

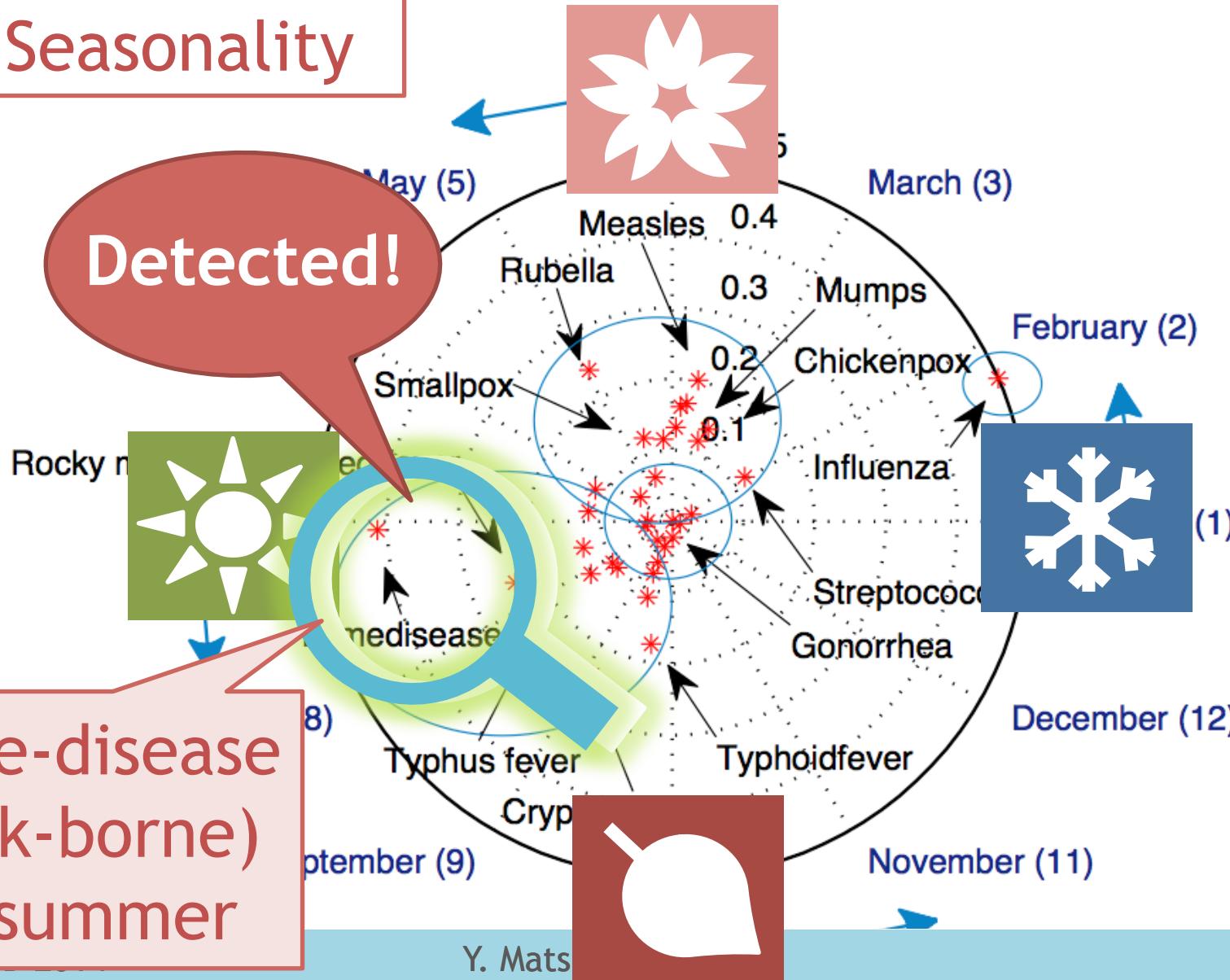
Q2

Q3

Q4

Q5

P1 Seasonality



Answers

Q1

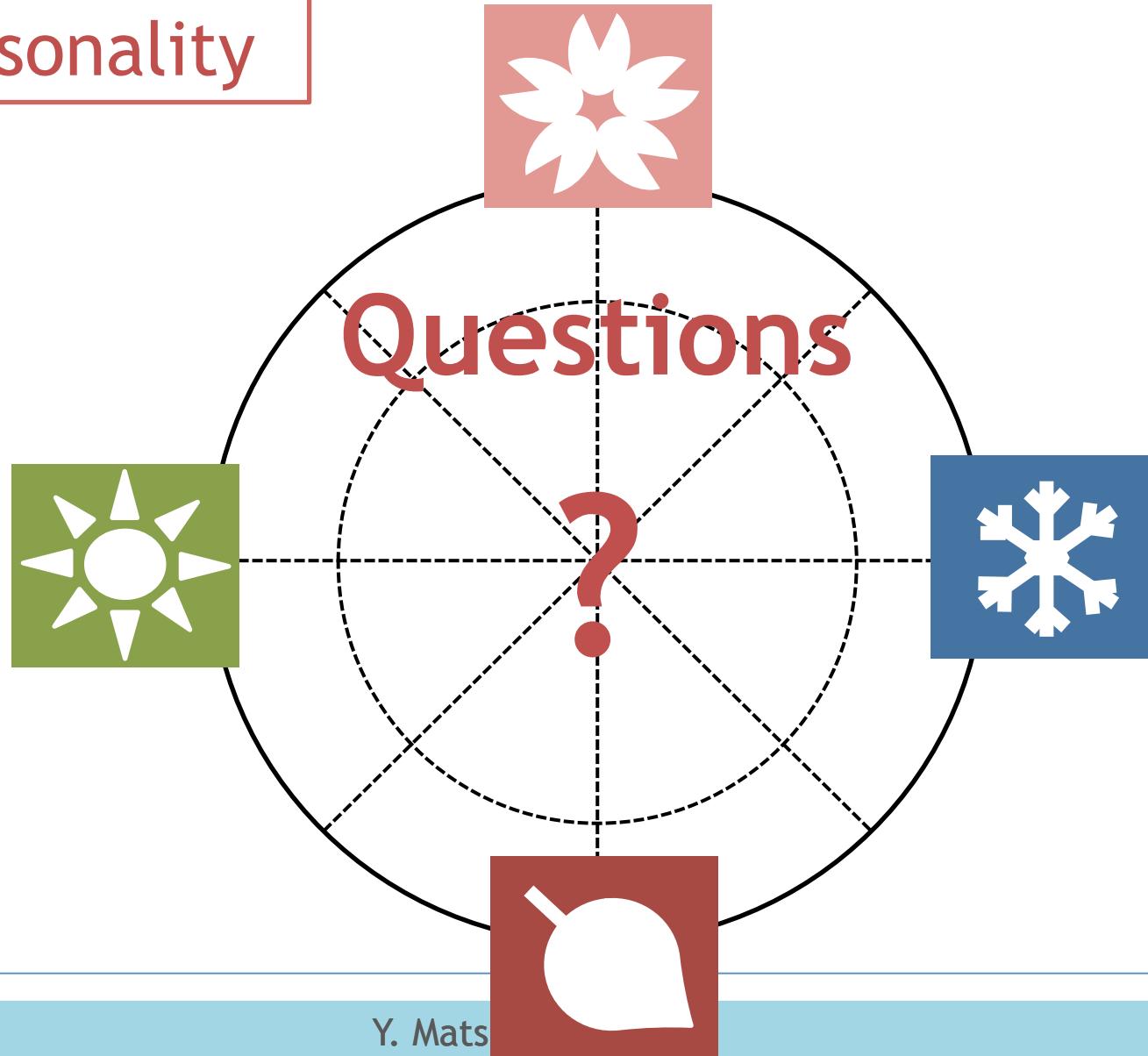
Q2

Q3

Q4

Q5

P1 Seasonality



Answers

Q1

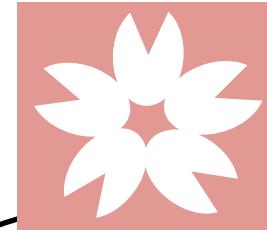
Q2

Q3

Q4

Q5

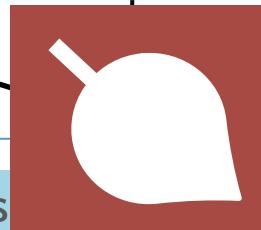
P1 Seasonality



Questions



Q: Which disease has no periodicity?



Answers

Q1

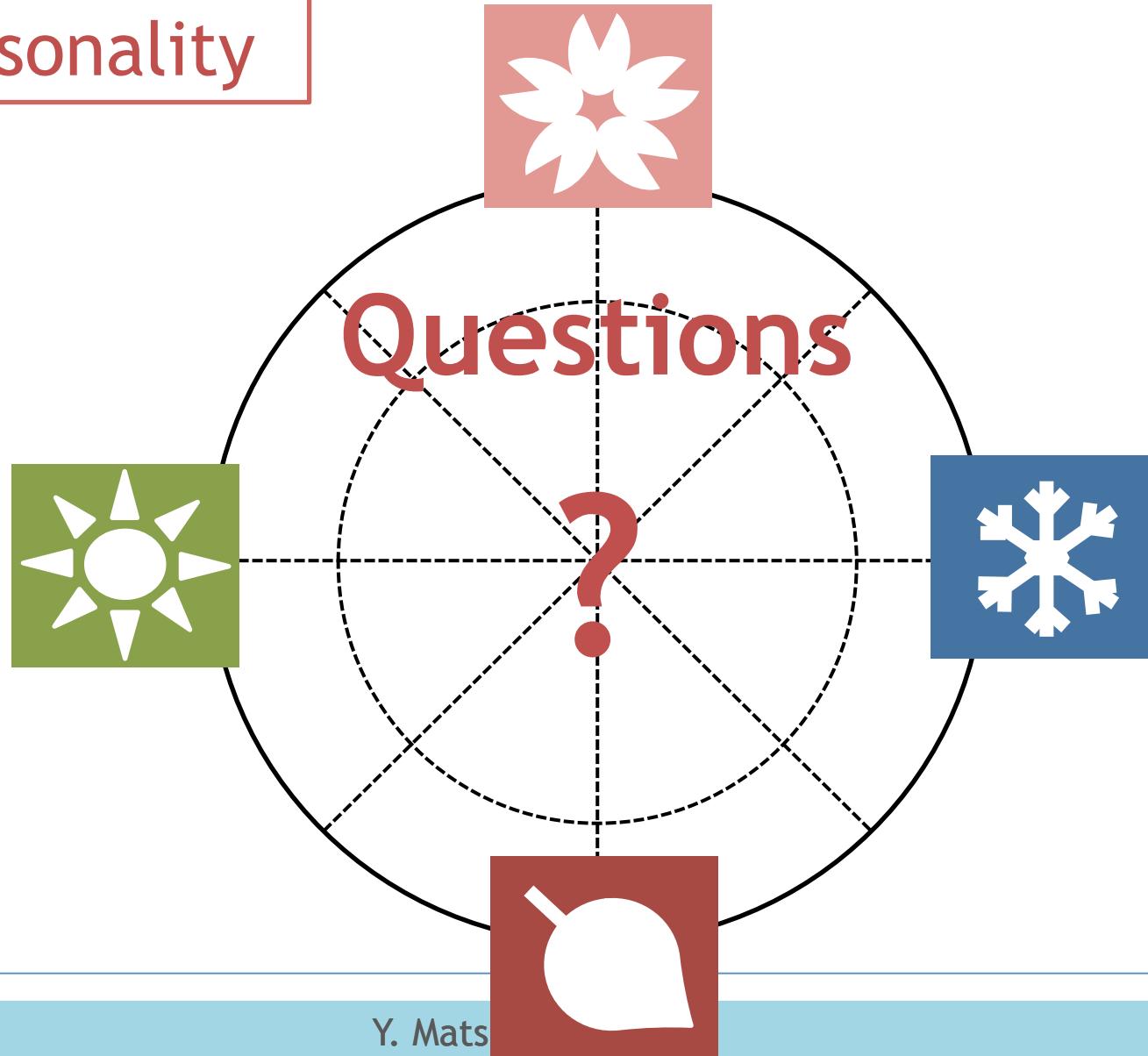
Q2

Q3

Q4

Q5

P1 Seasonality



Answers

Q1

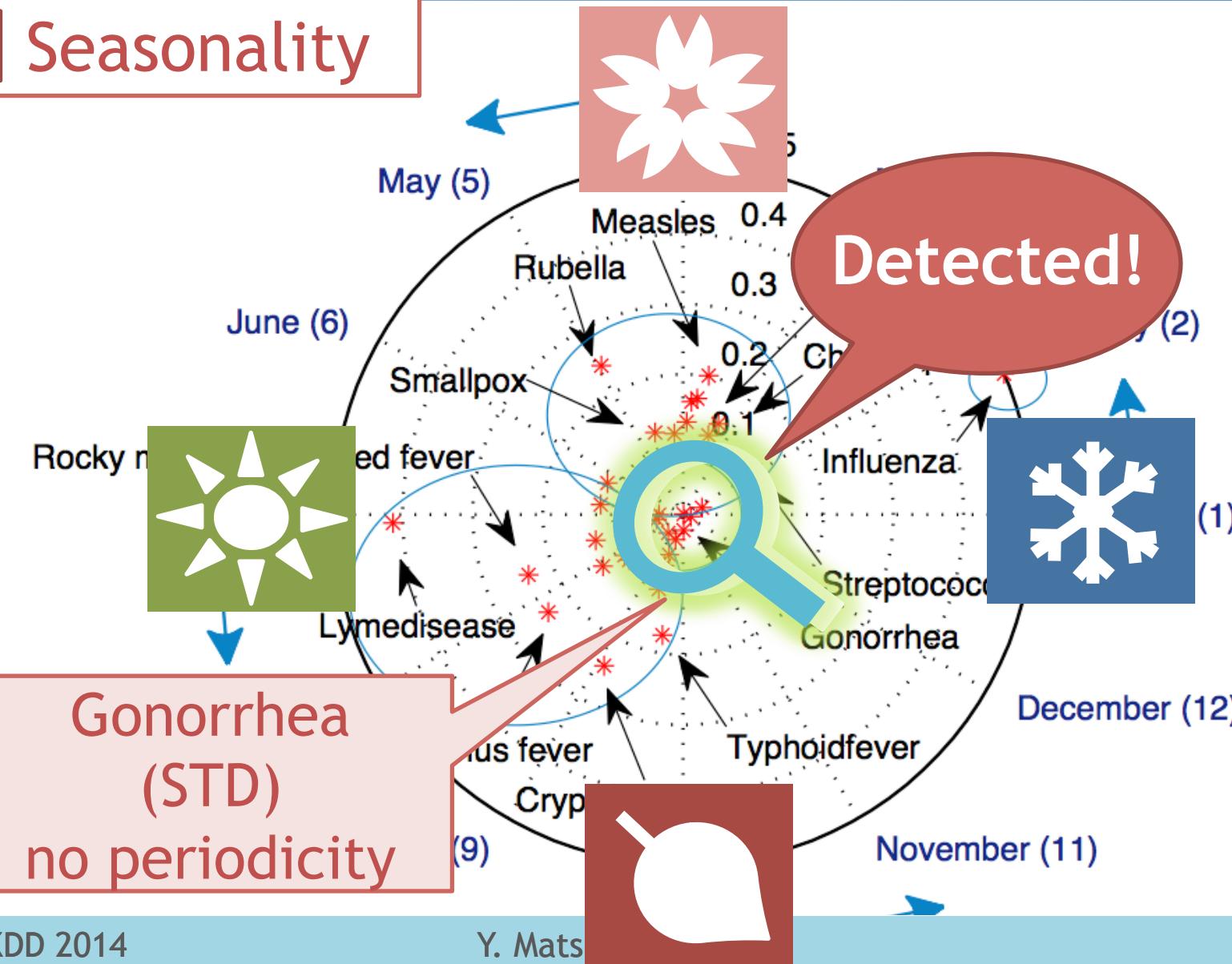
Q2

Q3

Q4

Q5

P1 Seasonality



Questions

Q1

Q2

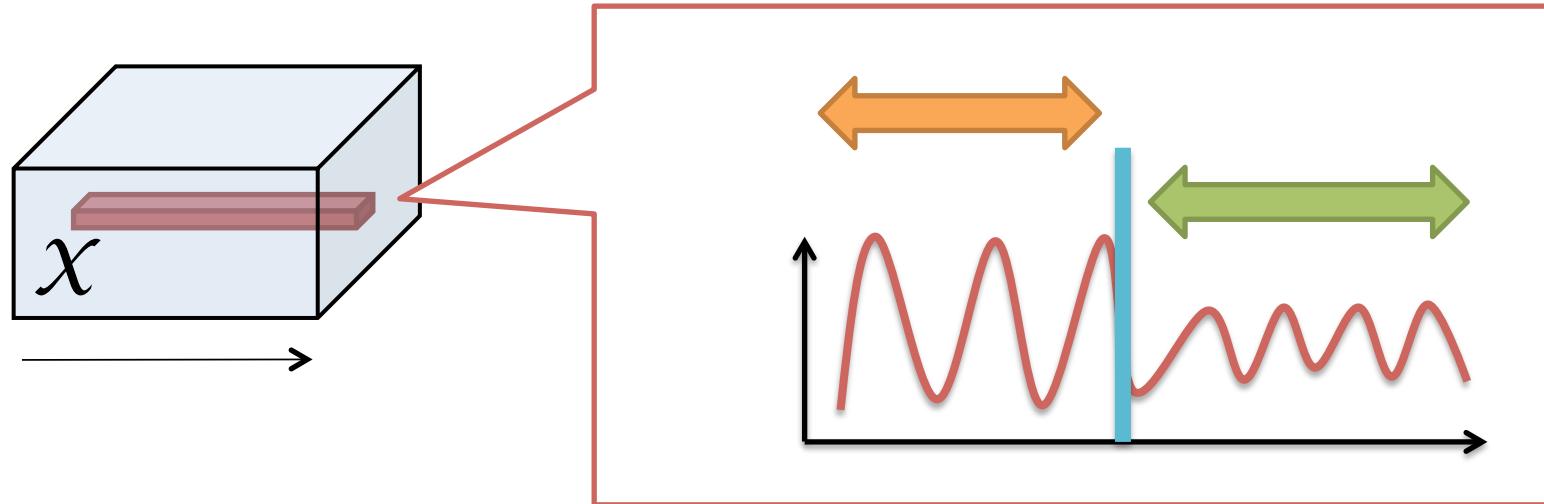
Q3

Q4

Q5

Q2

Can we see any discontinuities?



Answers

Q1

Q2

Q3

Q4

Q5

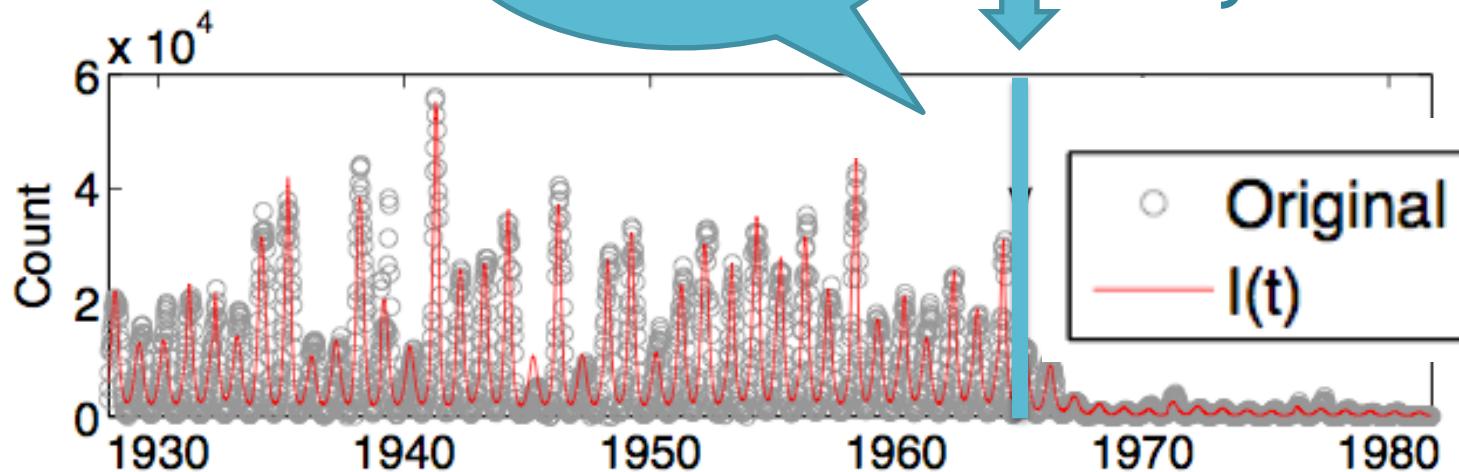
P2

Disease reduction effect

Measles

Detected!

1965: Detected by FUNNEL



1963:
Vaccine licensure

Questions

Q1

Q2

Q3

Q4

Q5

Q3

What's the difference between
measles in NY and in FL?



Answers

Q1

Q2

Q3

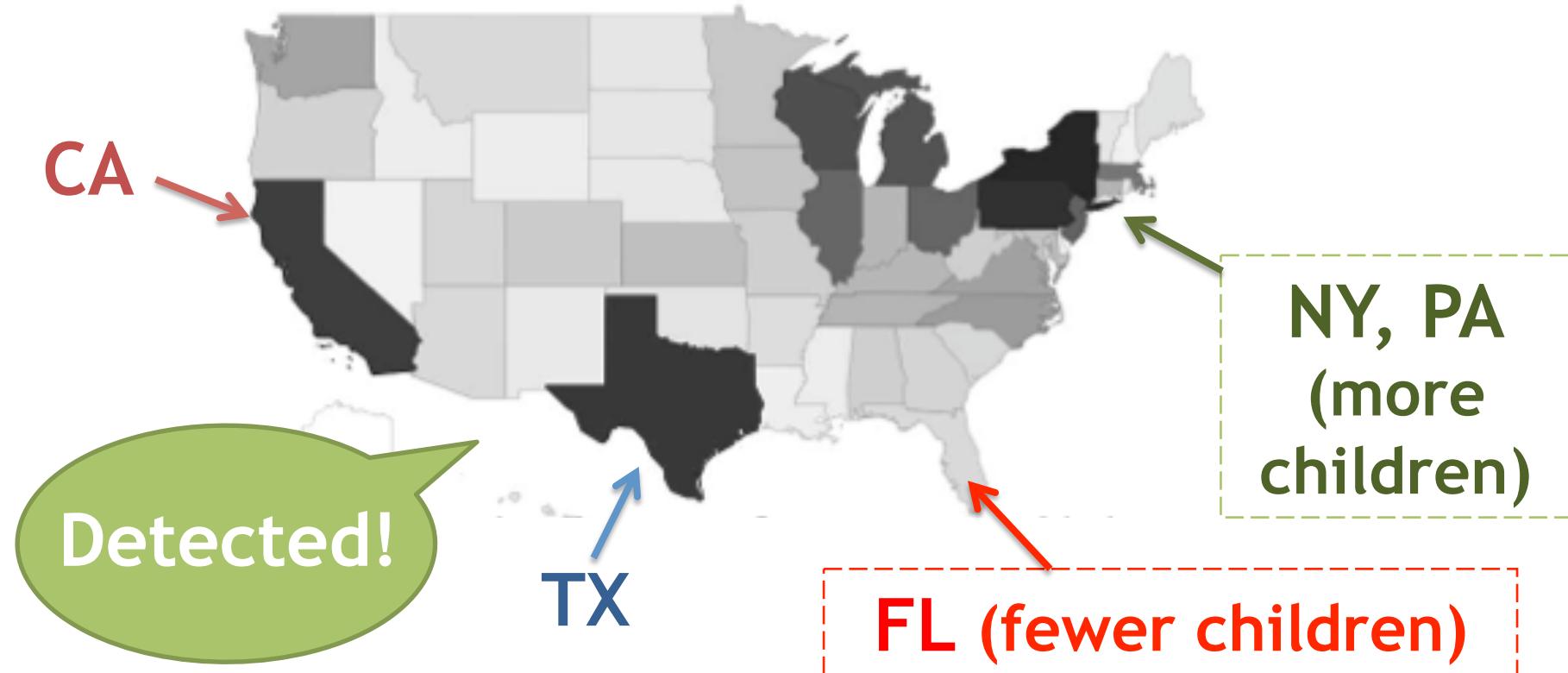
Q4

Q5

P3

area sensitivity

FUNNEL's guess of susceptibles (measles)



Questions

Q1

Q2

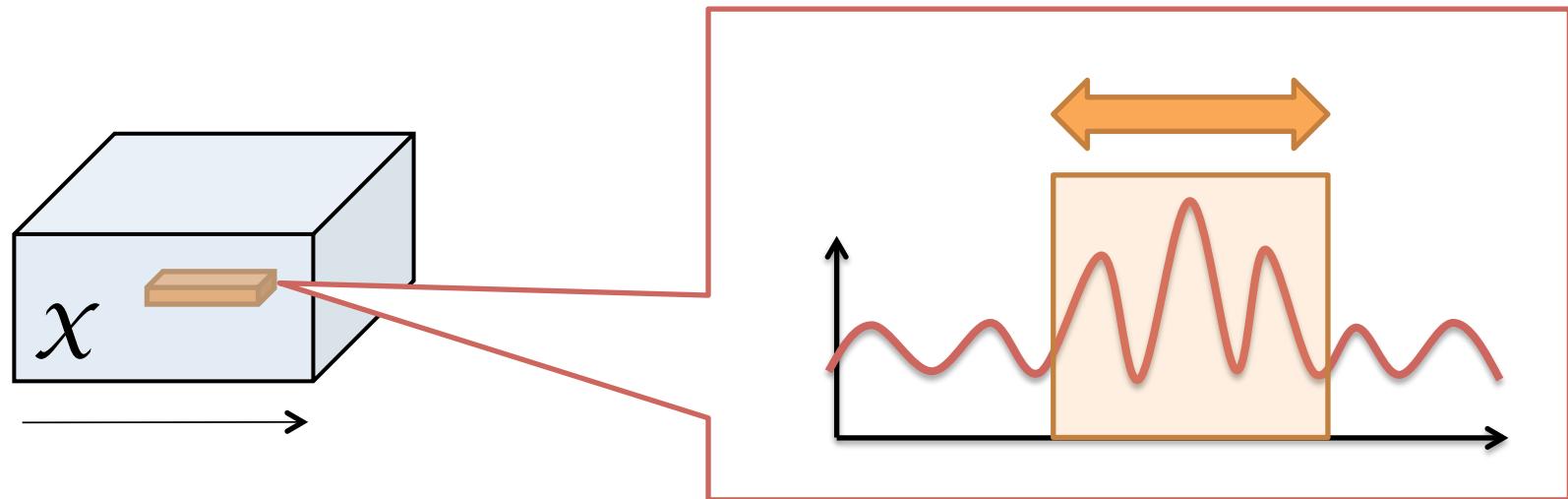
Q3

Q4

Q5

Q4

Are there any external
shock events, like wars?



Answers

Q1

Q2

Q3

Q4

Q5

P4

external shock events

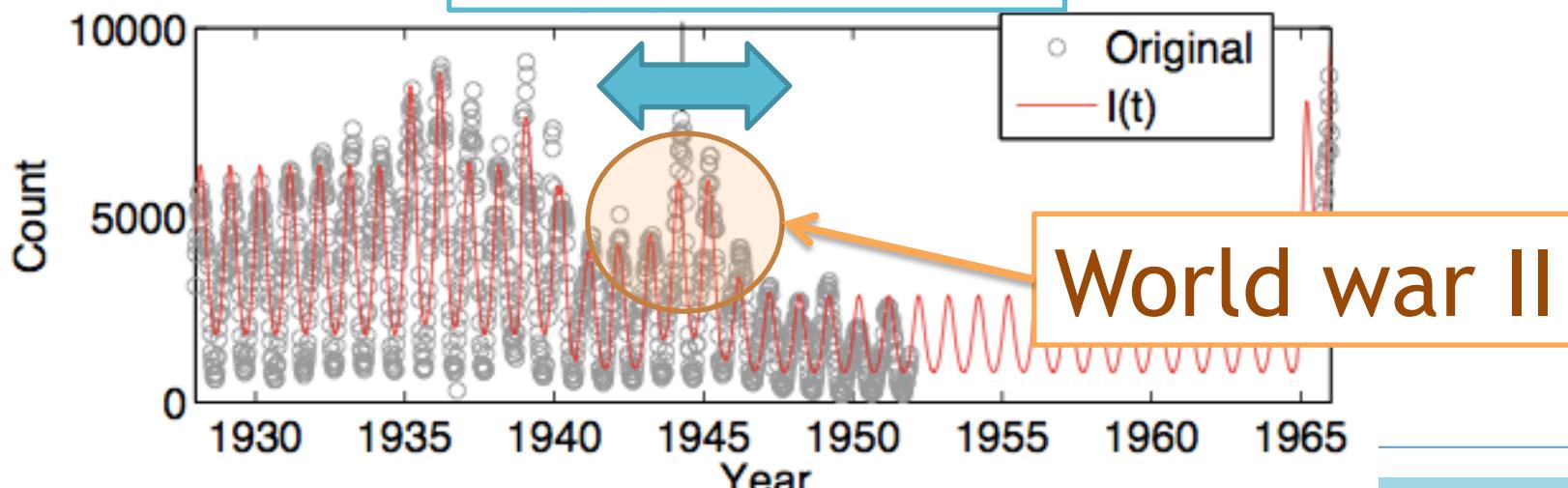
Funnel can detect external shocks

“fully-automatically” !

Scarlet fever

Detected by
FUNNEL

Detected!



Questions

Q1

Q2

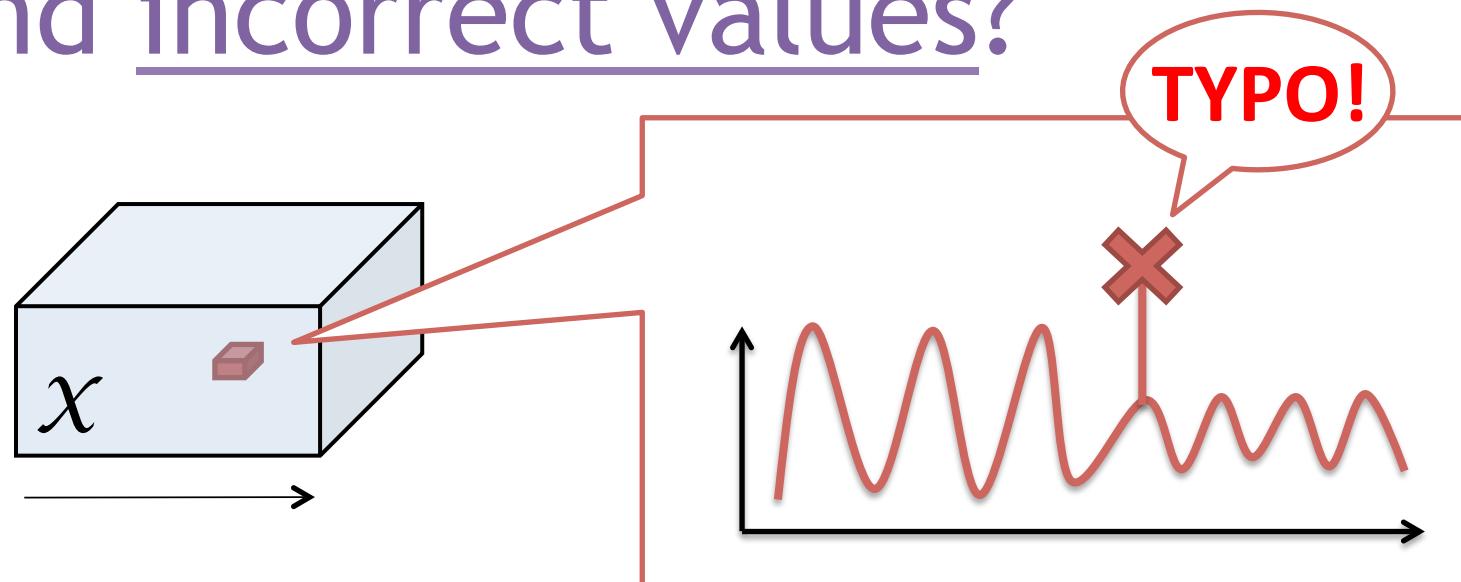
Q3

Q4

Q5

Q5

How can we remove mistakes
and incorrect values?



Answers

Q1

Q2

Q3

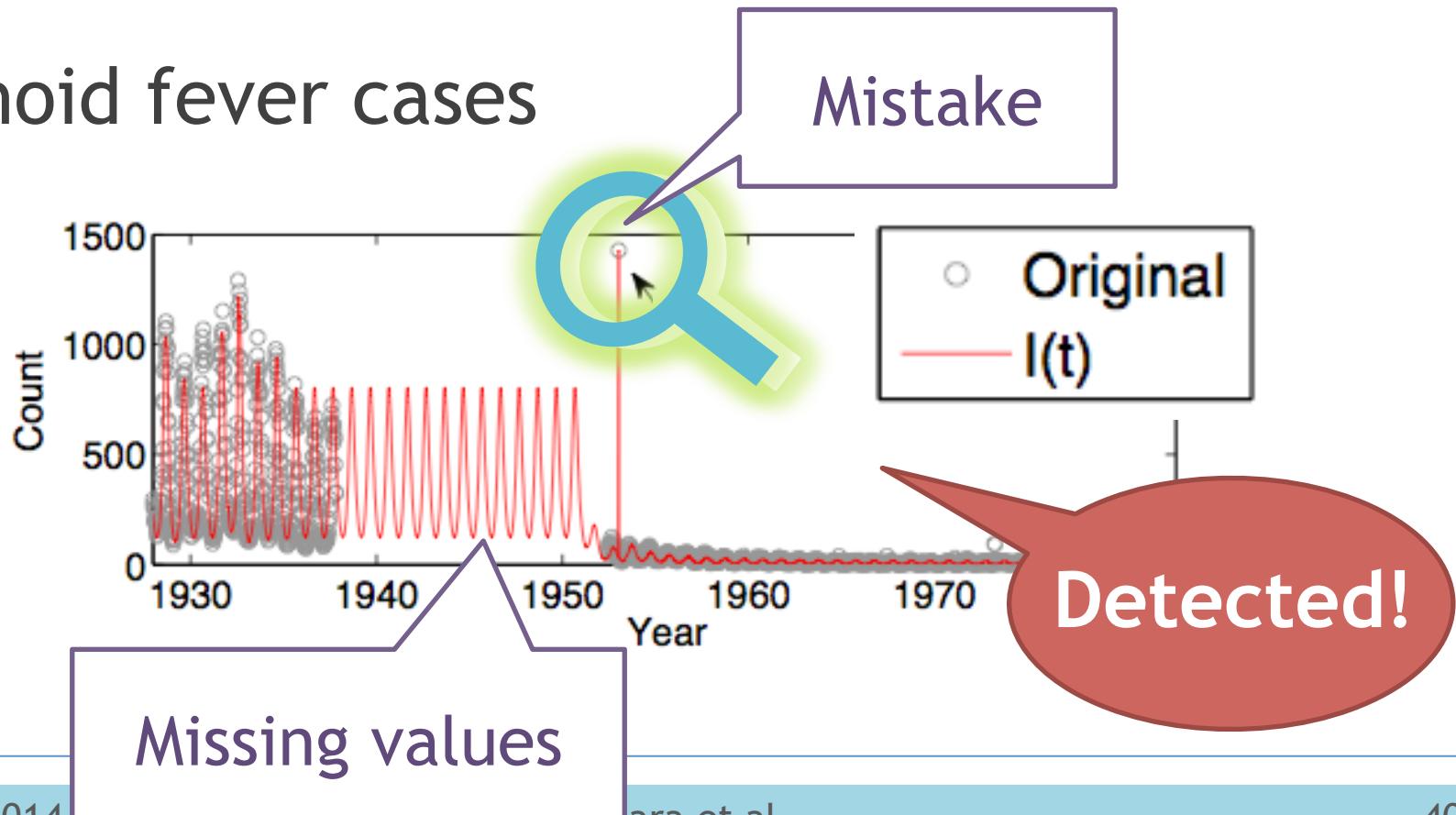
Q4

Q5

P5 mistakes

It can also detect typos, “**automatically**” !!

Typhoid fever cases



Modeling power of FUNNEL

Our model can capture 5 properties

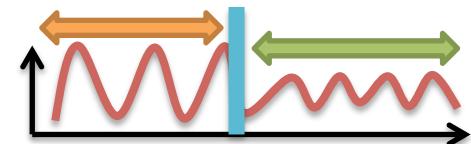
P1

Seasonality



P2

Disease reductions



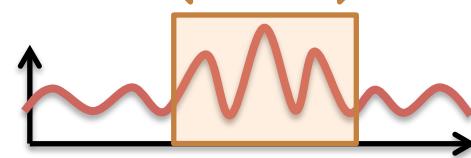
P3

Area sensitivity



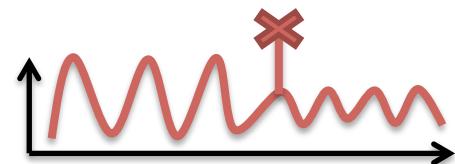
P4

External events



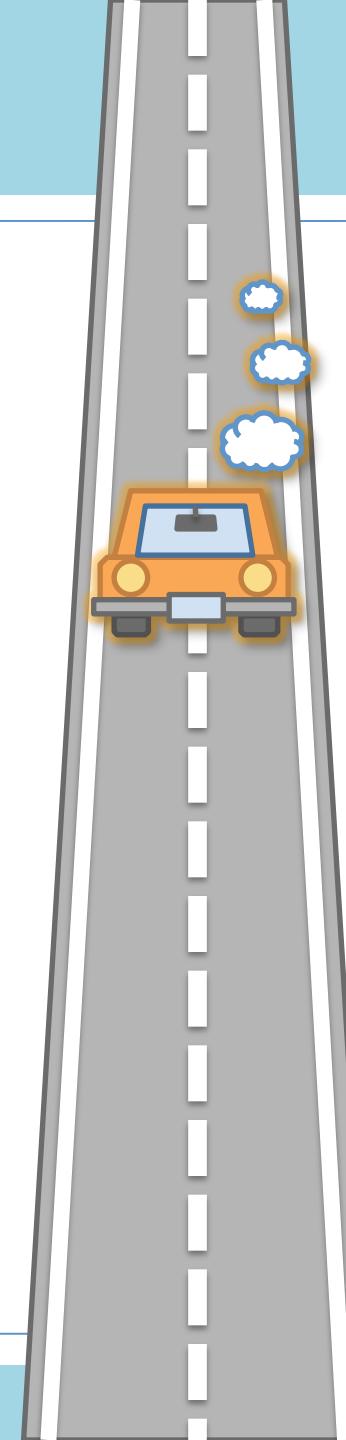
P5

Mistakes



Roadmap

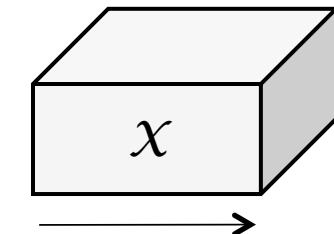
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- Overview - main ideas
- Proposed model - idea #1
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- Conclusions



Problem definition

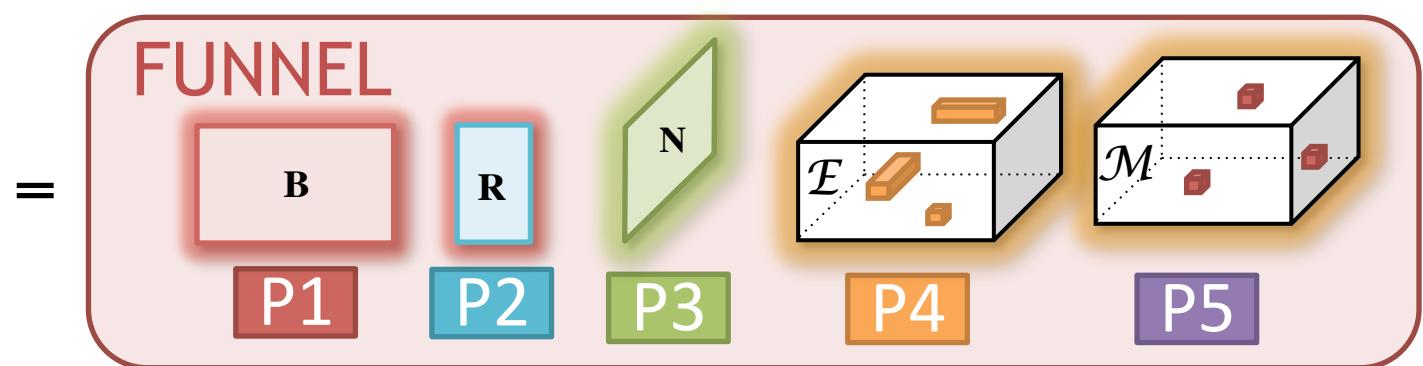
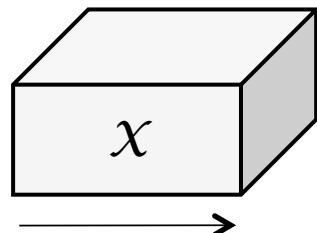
Given:

Tensor \mathcal{X} (**disease** x **state** x **time**)



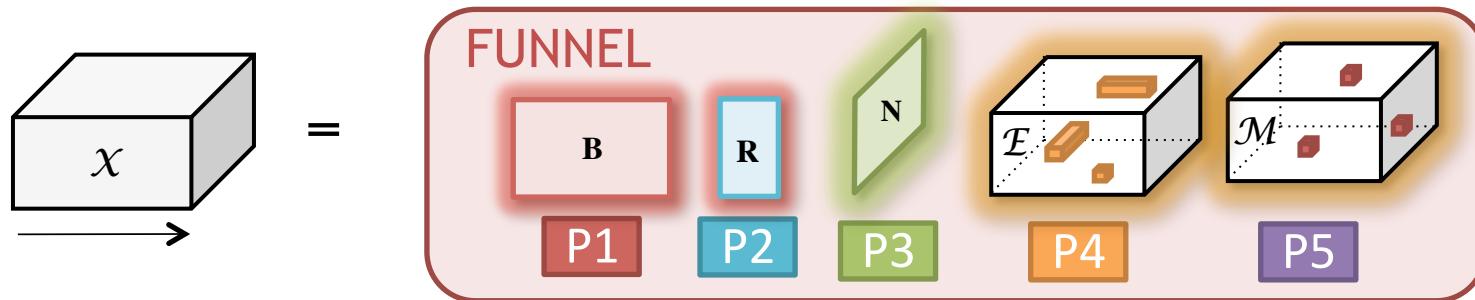
Find:

Compact description of \mathcal{X} , “*automatically*”



Two main ideas

Idea #1: Grey-box model



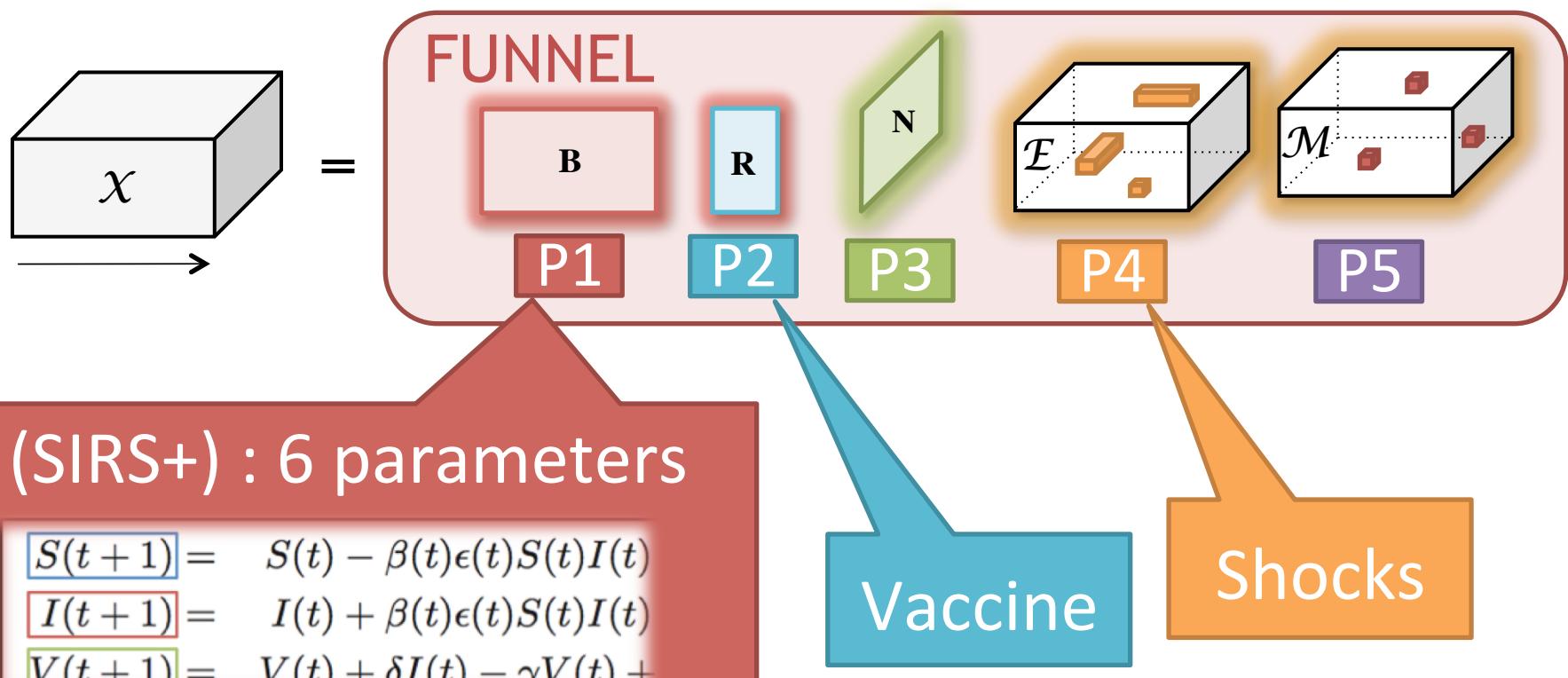
Idea #2: MDL for fitting

**NO magic numbers !
(parameter-free)**



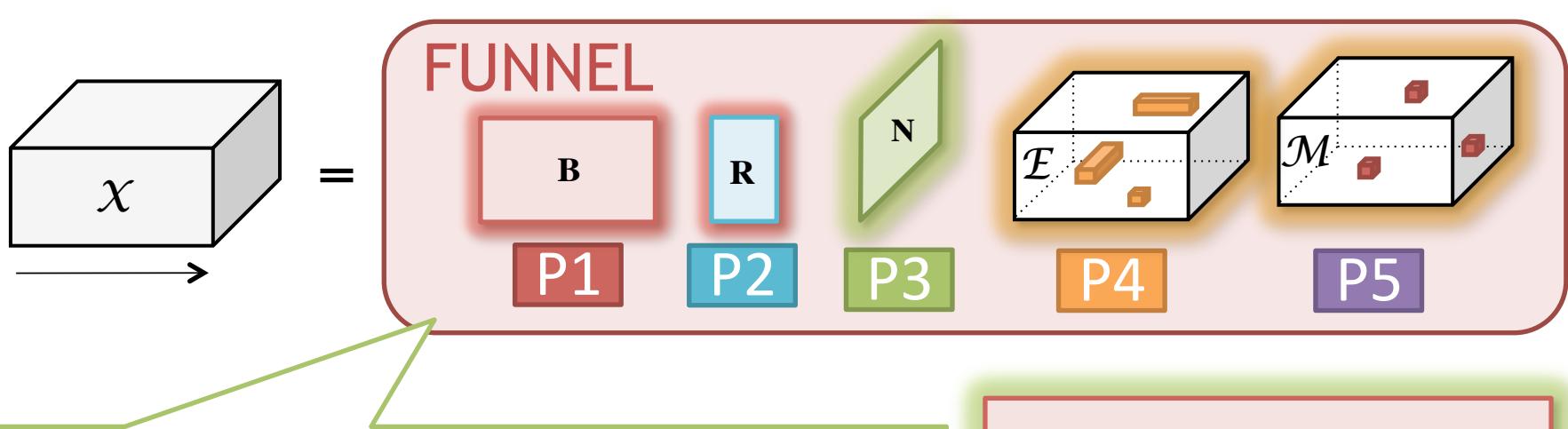
Two main ideas

Idea #1: Grey-box model - domain knowledge



Two main ideas

Idea #2: Fitting with MDL -> parameter free!



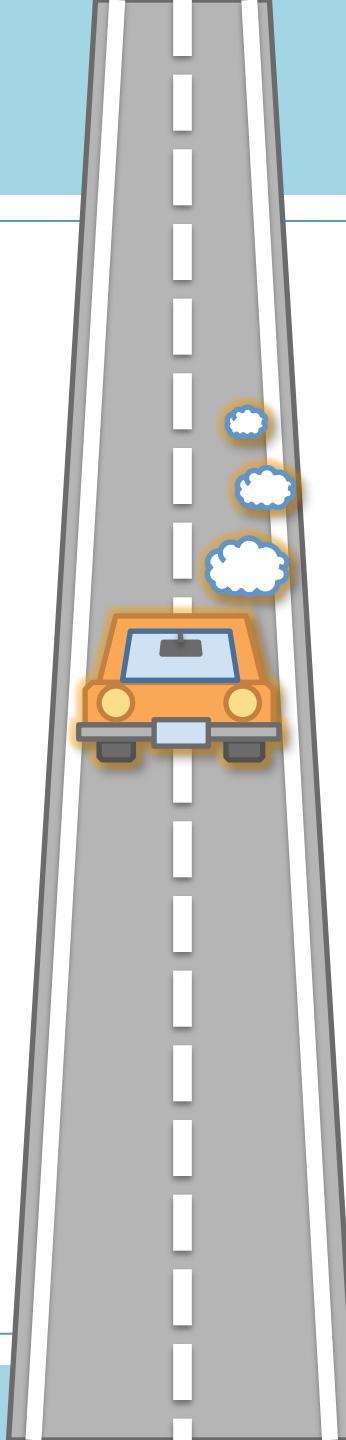
$$\begin{aligned} Cost_T(\mathcal{X}; \mathcal{F}) &= \log^*(d) + \log^*(l) + \log^*(n) \\ &+ Cost_M(\mathbf{B}) + Cost_M(\mathbf{R}) + Cost_M(\mathbf{N}) \\ &+ Cost_M(\mathcal{E}) + Cost_M(\mathcal{M}) + Cost_C(\mathcal{X}|\mathcal{F}) \end{aligned}$$

NO magic numbers
Parameter-free!

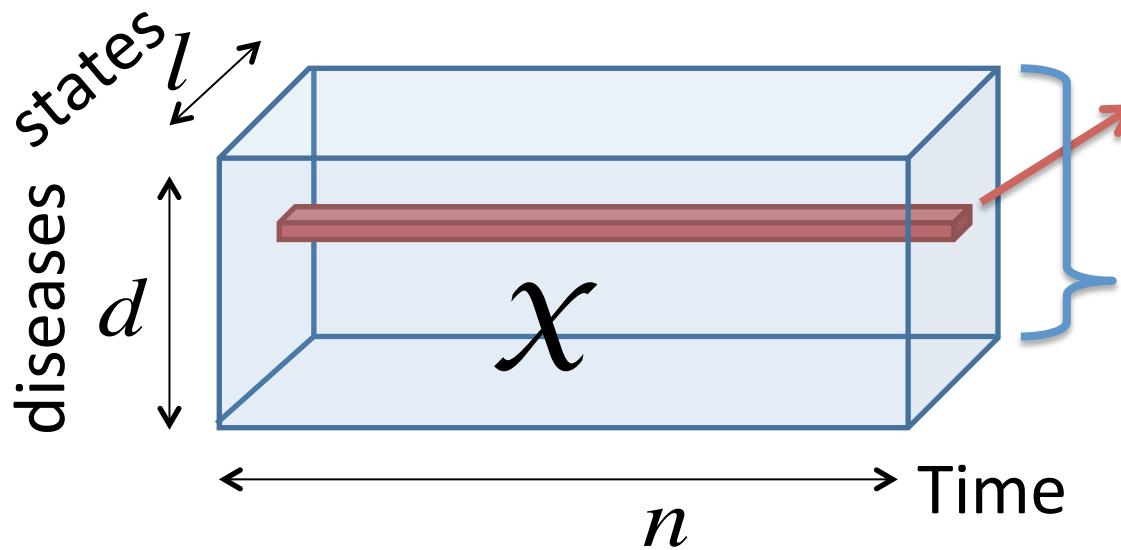
A red circle with a diagonal slash through it is positioned above a blue calculator icon. Five small, colorful stars (yellow, orange, green, blue, and red) are scattered around the calculator icon.

Roadmap

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- ✓ Overview - main ideas
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- Discussion
- Conclusions



Proposed model: FUNNEL

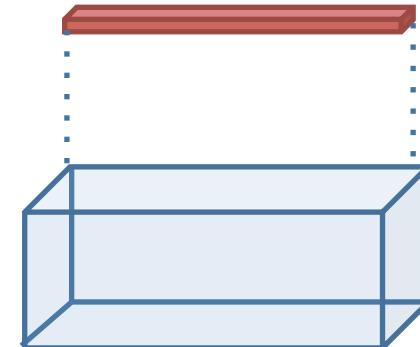


single epidemic

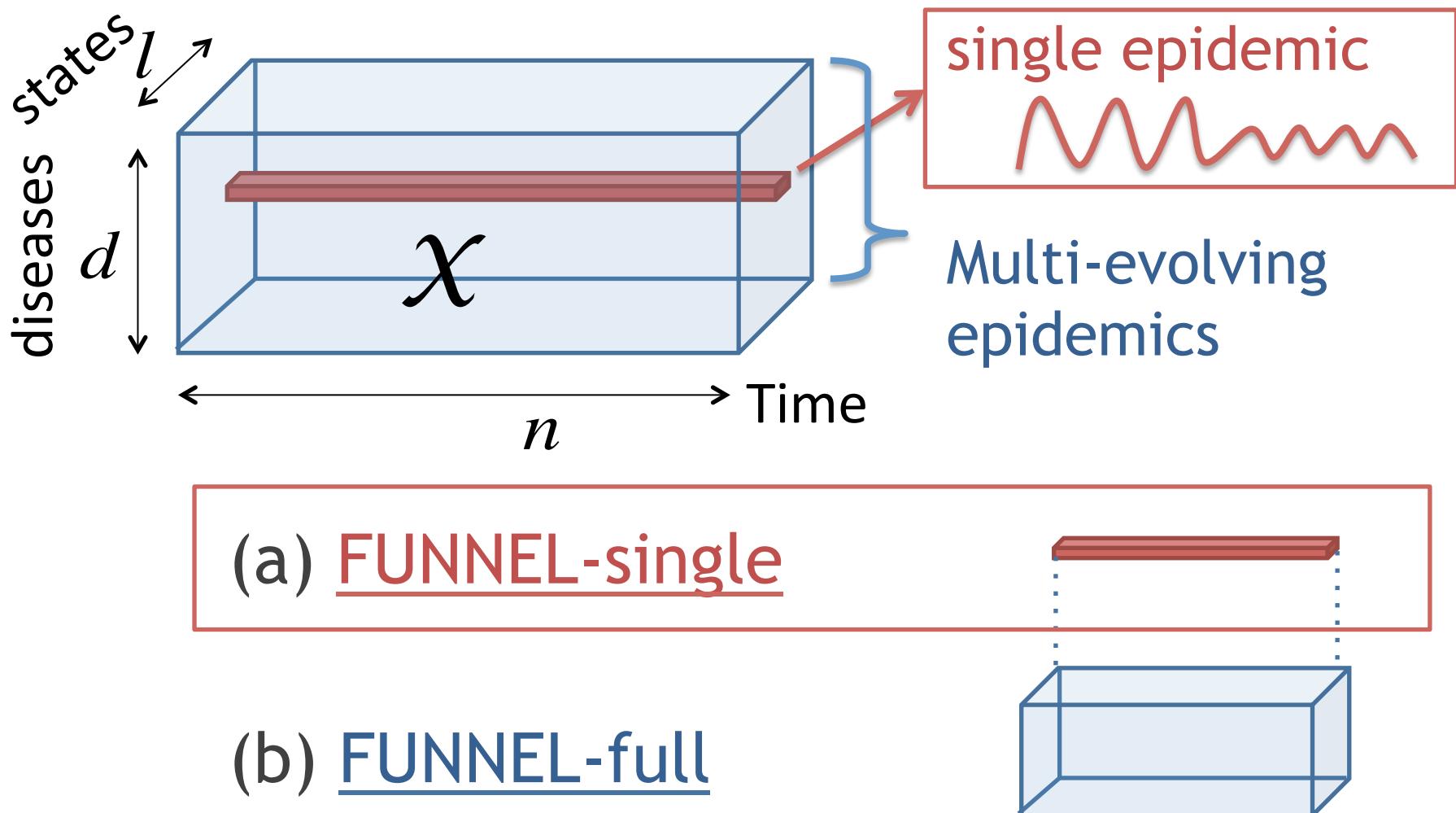
Multi-evolving epidemics

(a) FUNNEL-single

(b) FUNNEL-full



Proposed model: FUNNEL

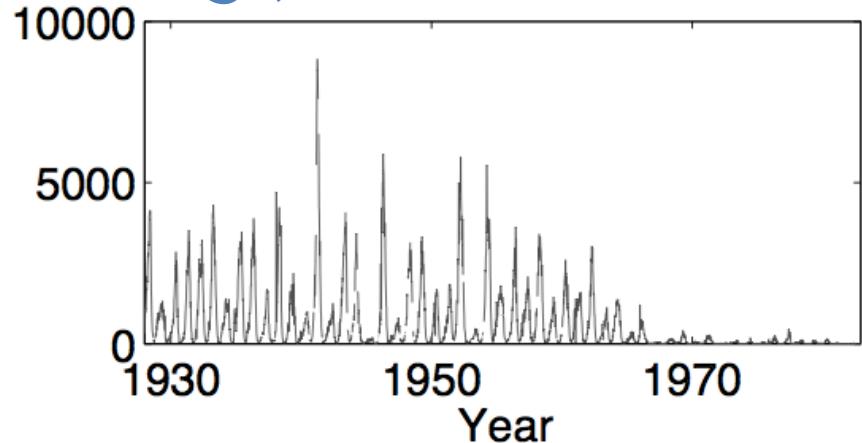


FUNNEL - with a single epidemic

Given:

“single” epidemic sequence

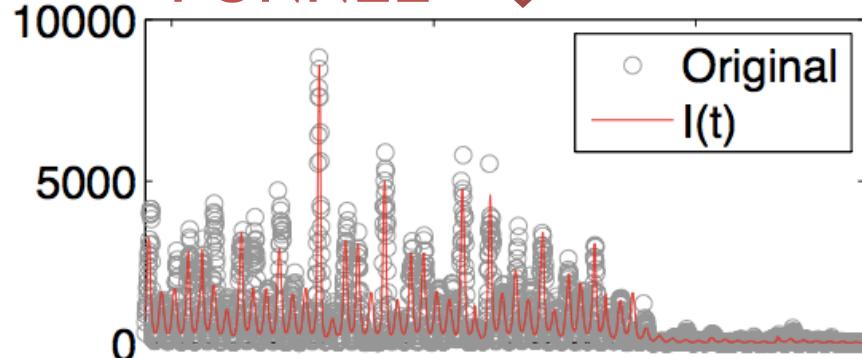
e.g., measles in NY



Find:

nonlinear equation,
model parameters

FUNNEL



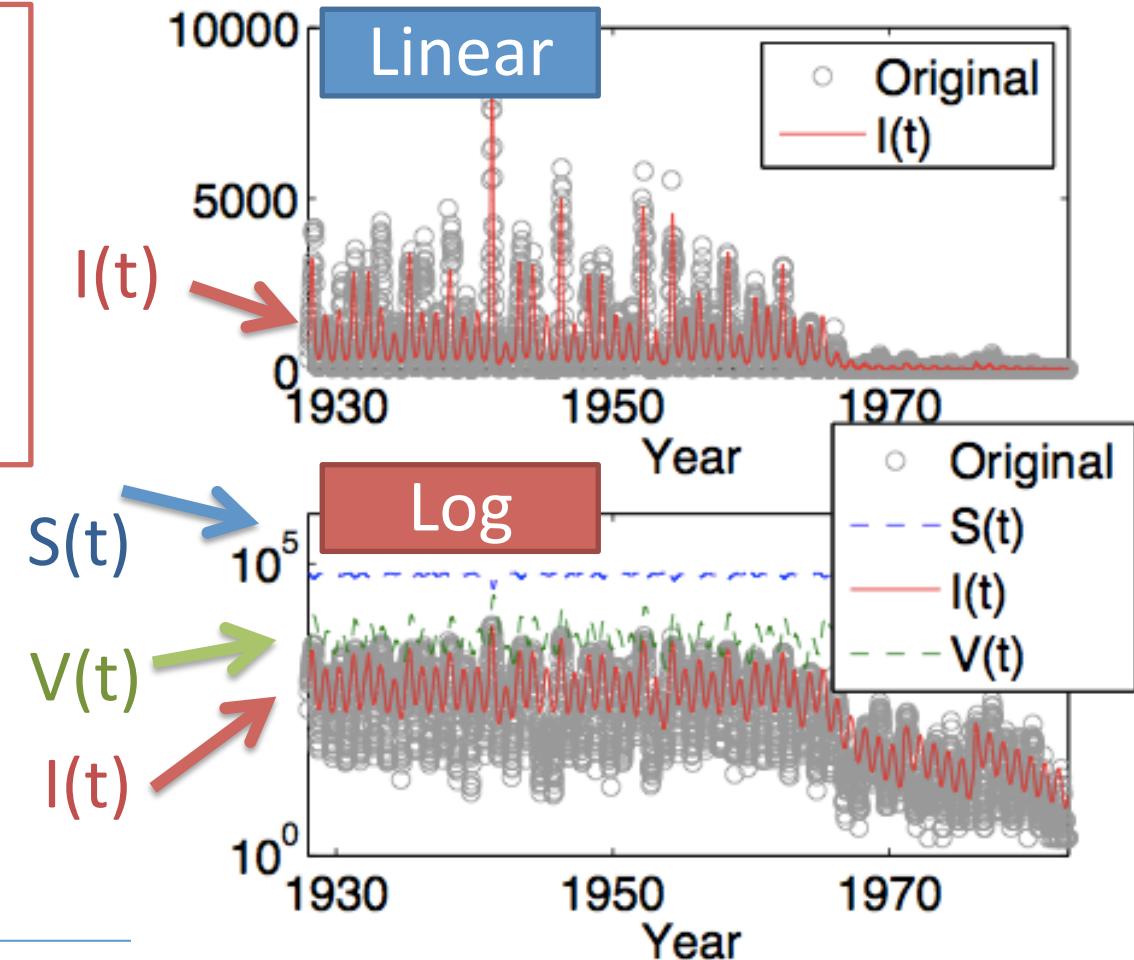
FUNNEL - with a single epidemic

Details

With a single epidemic: Funnel-RE

People of 3 classes

- S : Susceptible
- I : Infected
- V : Vigilant/
vaccinated



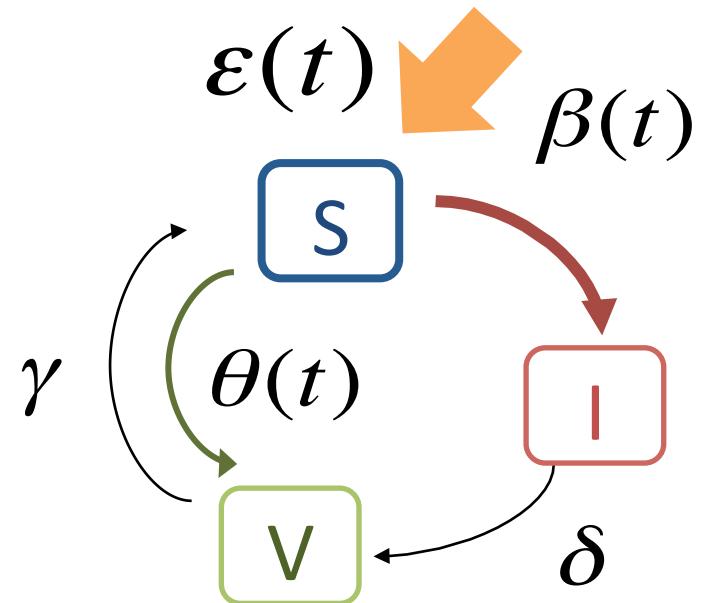
FUNNEL - with a single epidemic

Details

With a single epidemic: Funnel-RE

$$\begin{aligned} S(t+1) &= S(t) - \beta(t)\epsilon(t)S(t)I(t) + \gamma V(t) - \theta(t)S(t) \\ I(t+1) &= I(t) + \beta(t)\epsilon(t)S(t)I(t) - \delta I(t) \\ V(t+1) &= V(t) + \delta I(t) - \gamma V(t) + \theta(t)S(t) \end{aligned} \quad (3)$$

$S(t)$: susceptible
 $I(t)$: Infected
 $V(t)$: Vigilant /Vaccinated



FUNNEL - with a single epidemic

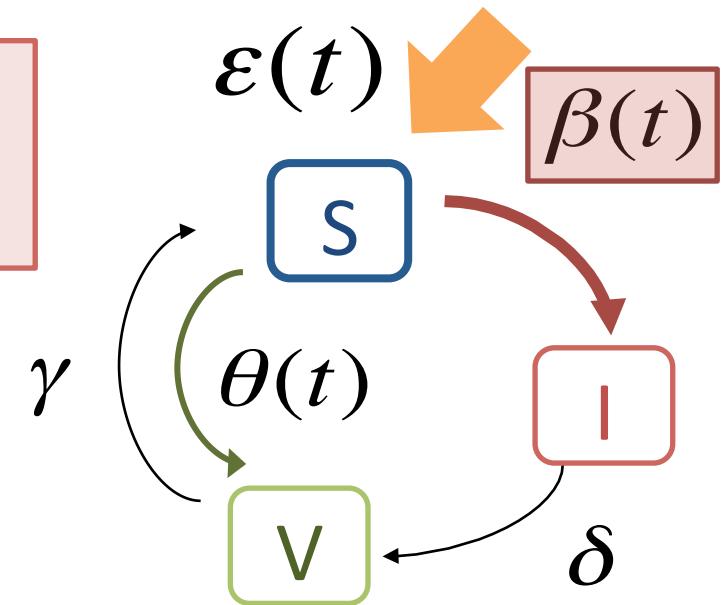
Details

With a single epidemic: Funnel-RE

$$\begin{aligned} S(t+1) &= S(t) - \beta(t)\epsilon(t)S(t)I(t) + \gamma V(t) - \theta(t)S(t) \\ I(t+1) &= I(t) + \beta(t)\epsilon(t)S(t)I(t) - \delta I(t) \\ V(t+1) &= V(t) + \delta I(t) - \gamma V(t) + \theta(t)S(t) \end{aligned} \quad (3)$$

$\beta(t)$: strength of infection
(yearly periodic func)

$$\beta(t) = \beta_0 \cdot \left(1 + P_a \cdot \cos\left(\frac{2\pi}{P_p}(t + P_s)\right)\right) \quad P_p = 52$$



FUNNEL - with a single epidemic

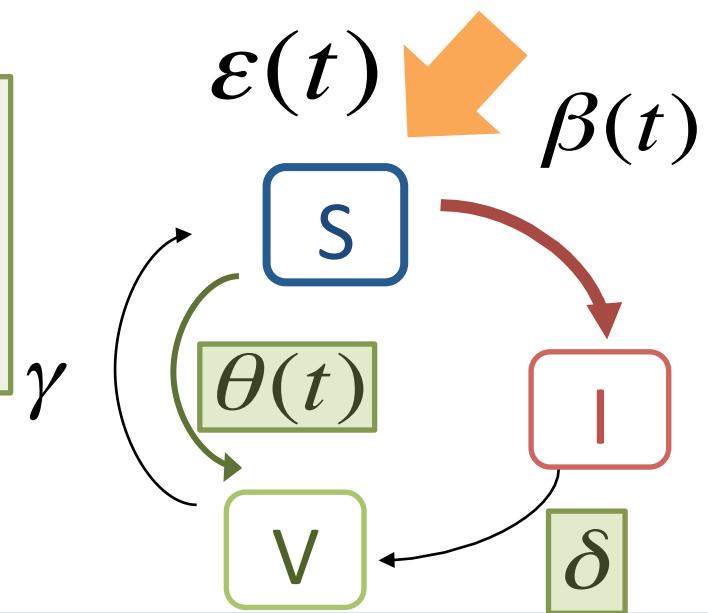
Details

With a single epidemic: Funnel-RE

$$\begin{aligned} S(t+1) &= S(t) - \beta(t)\epsilon(t)S(t)I(t) + \gamma V(t) - \theta(t)S(t) \\ I(t+1) &= I(t) + \beta(t)\epsilon(t)S(t)I(t) - \delta I(t) \\ V(t+1) &= V(t) + \delta I(t) - \gamma V(t) + \theta(t)S(t) \end{aligned} \quad (3)$$

δ : healing rate
 $\theta(t)$: disease reduction effect

$$\theta(t) = \begin{cases} 0 & (t < t_\theta) \\ \theta_0 & (t \geq t_\theta) \end{cases}$$



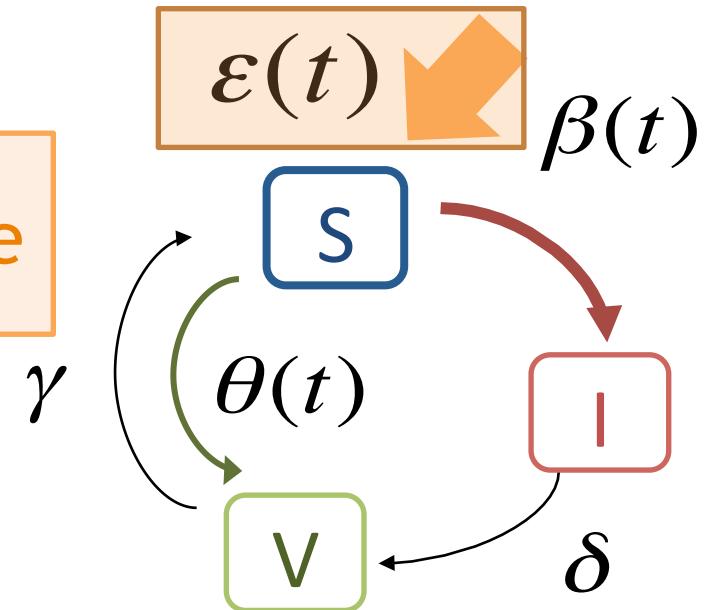
FUNNEL - with a single epidemic

Details

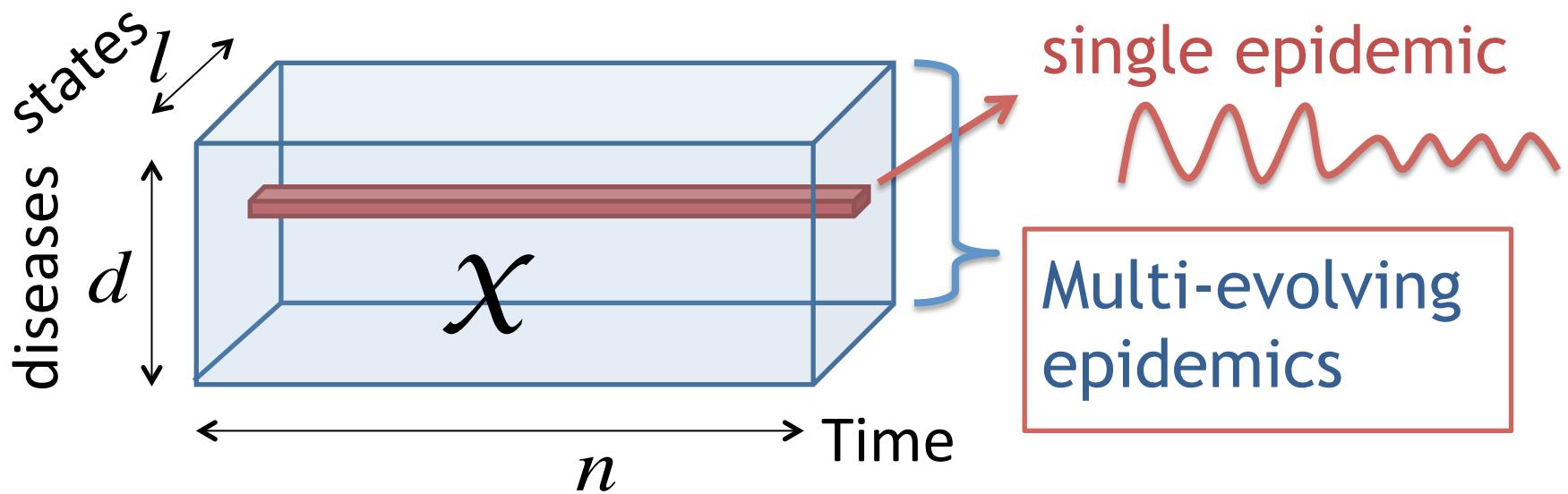
With a single epidemic: Funnel-RE

$$\begin{aligned} S(t+1) &= S(t) - \beta(t)\epsilon(t)S(t)I(t) + \gamma V(t) - \theta(t)S(t) \\ I(t+1) &= I(t) + \beta(t)\epsilon(t)S(t)I(t) - \delta I(t) \\ V(t+1) &= V(t) + \delta I(t) - \gamma V(t) + \theta(t)S(t) \end{aligned} \quad (3)$$

$\epsilon(t)$: temporal susceptible rate

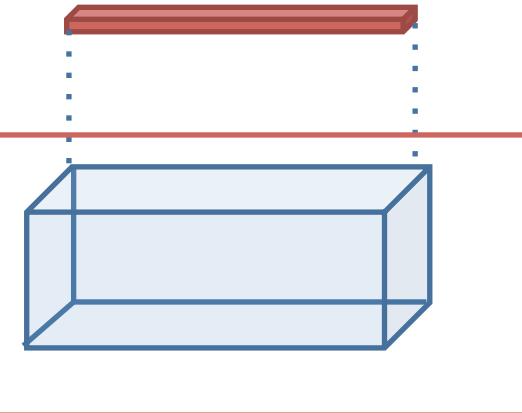


Proposed model: FUNNEL

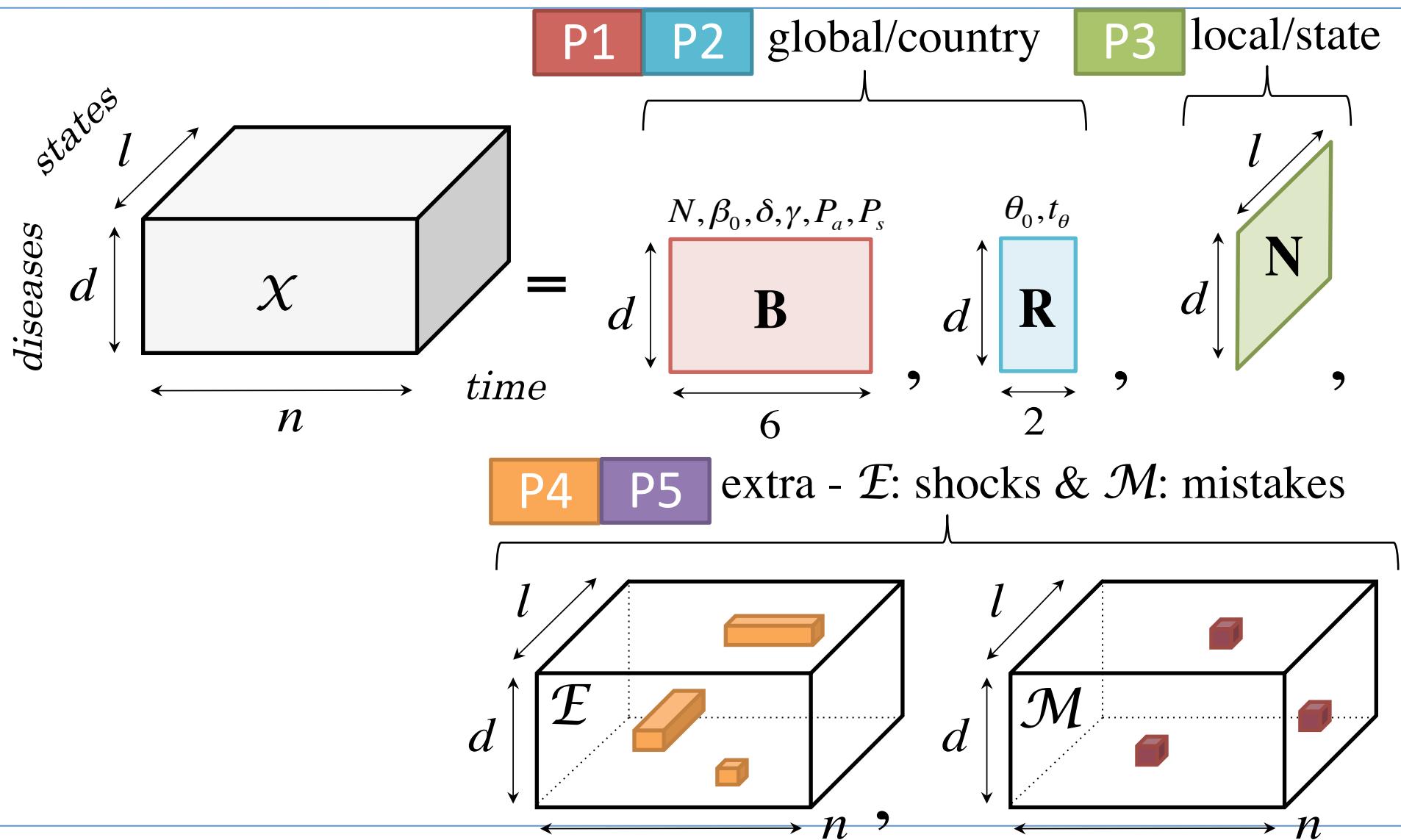


(a) FUNNEL-single

(b) FUNNEL-full

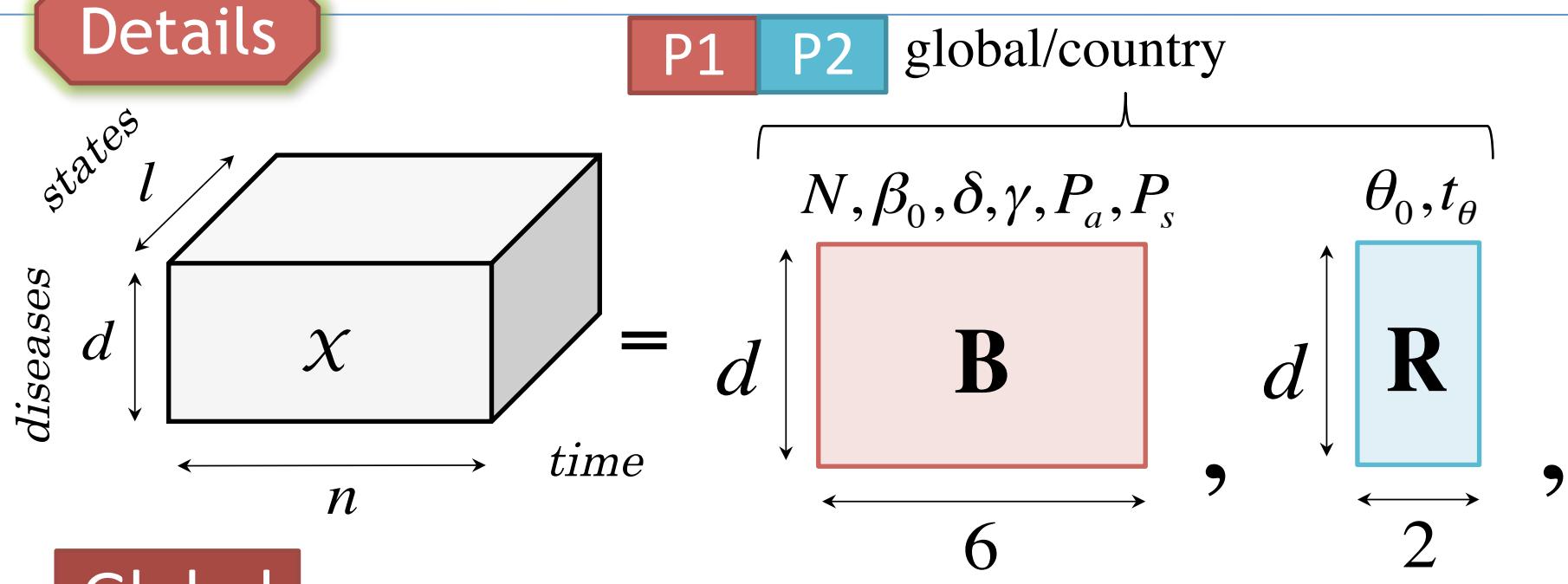


Proposed model: FUNNEL-full



Proposed model: FUNNEL-full

Details



Global

P1

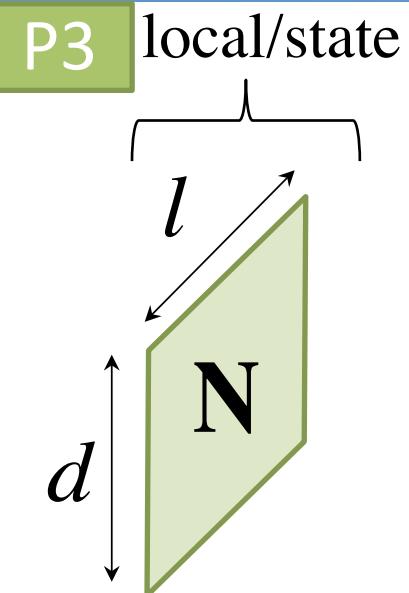
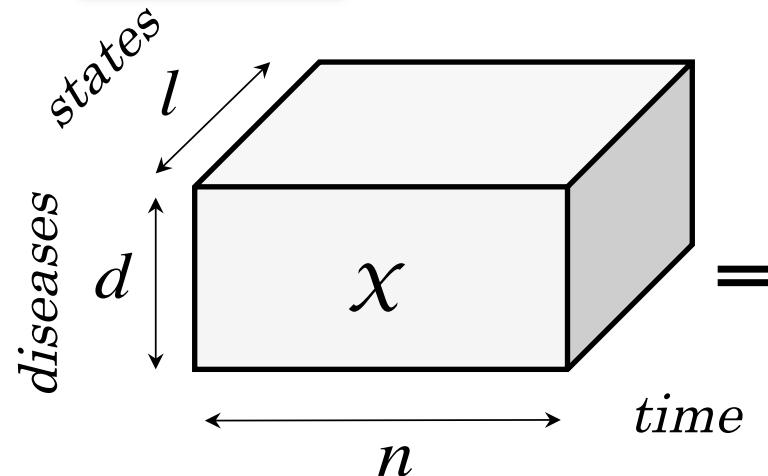
Base matrix B ($d \times 6$)

P2

Disease reduction matrix R ($d \times 2$)

Proposed model: FUNNEL-full

Details



Local

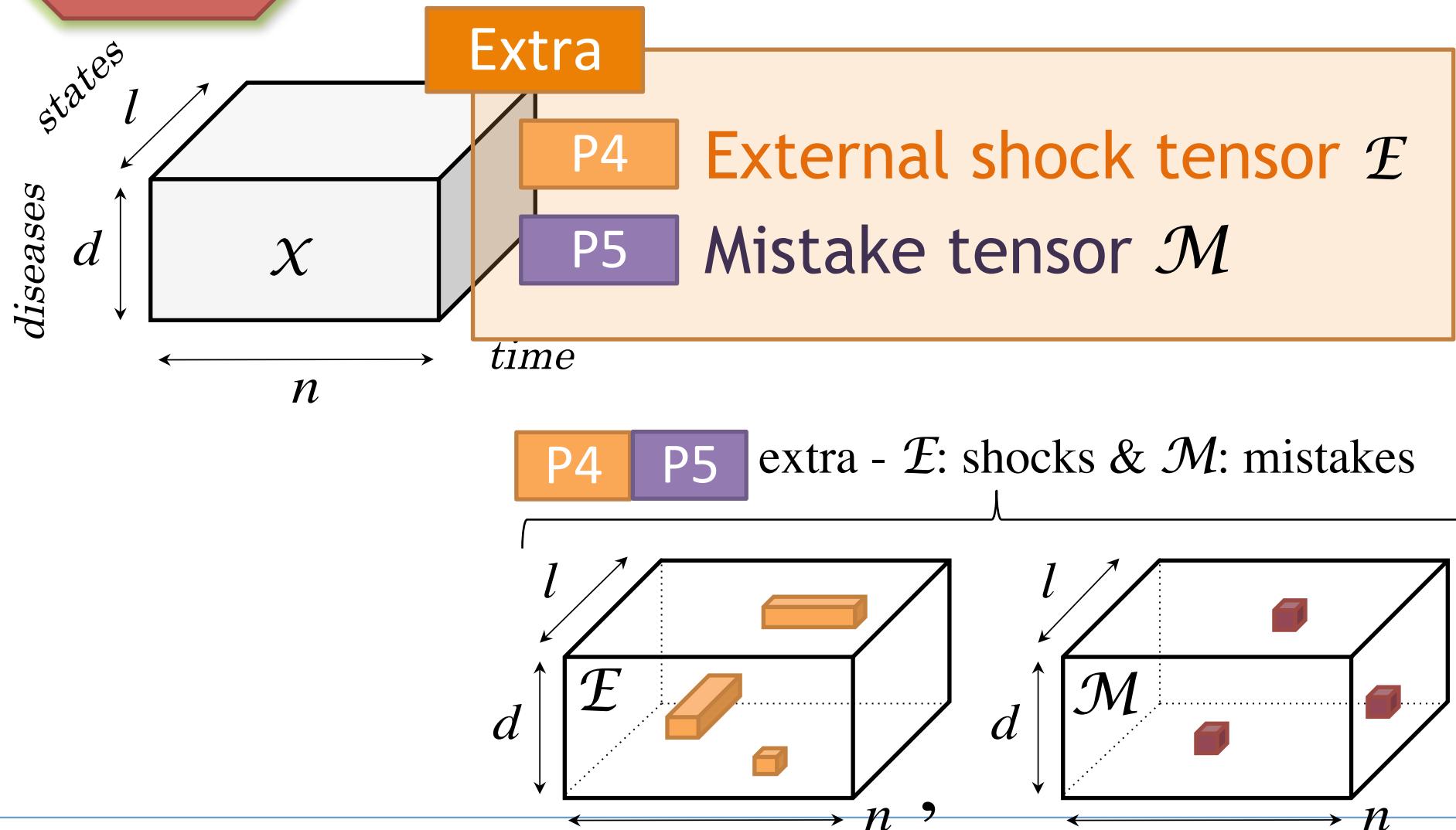
P3

Geo-disease matrix N ($d \times l$)

$N = \{N_{ij}\}_{i,j=1}^{d,l}$: potential population of disease i in state j

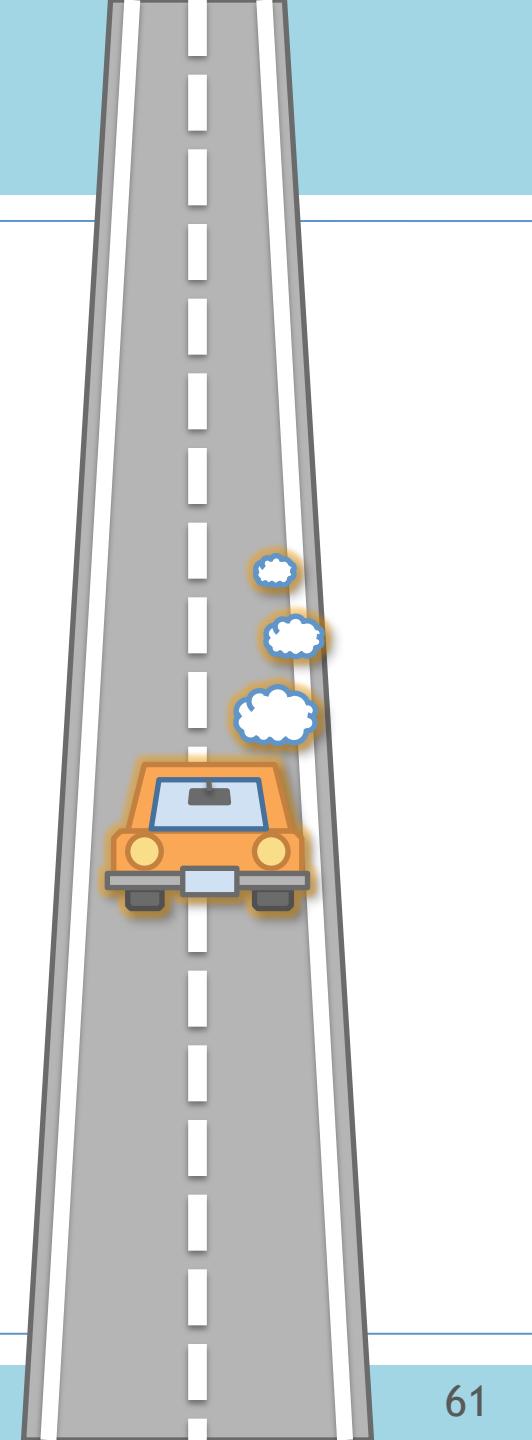
Proposed model: FUNNEL-full

Details



Roadmap

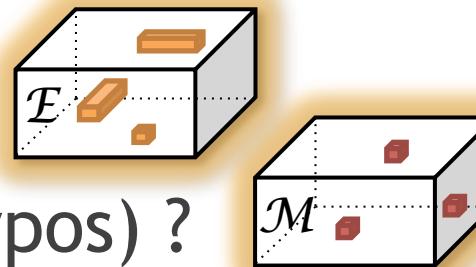
- ✓ Motivation
- ✓ Modeling power of FUNNEL
- ✓ Overview - main ideas
- ✓ Proposed model - idea #1
- Algorithm - idea #2
- Experiments
- Discussion
- Conclusions



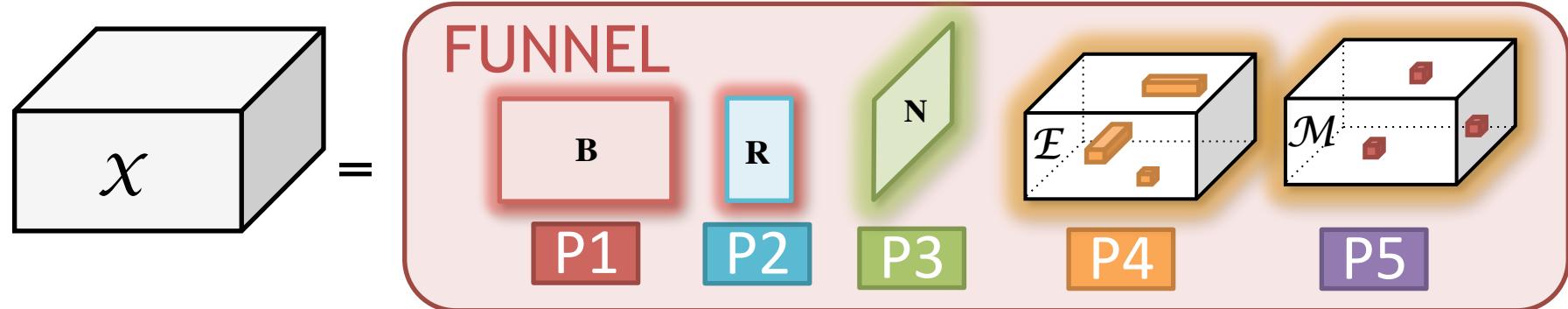
Challenges

Q1. How to automatically

- find “external shocks” ?
- ignore “mistakes” (i.e., typos) ?



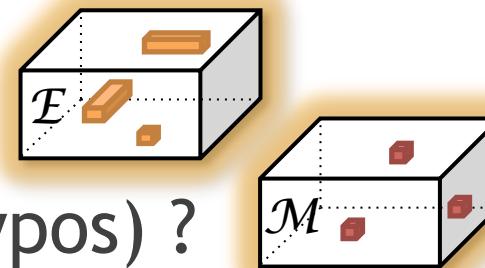
Q2. How to efficiently estimate model parameters ?



Challenges

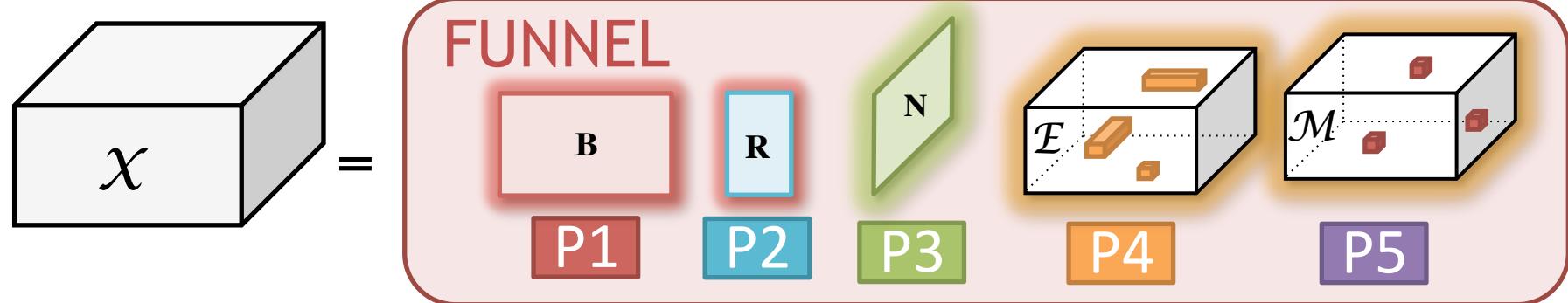
Q1. How to automatically

- find “external shocks” ?
- ignore “mistakes” (i.e., typos) ?



Idea (1) : Model description cost

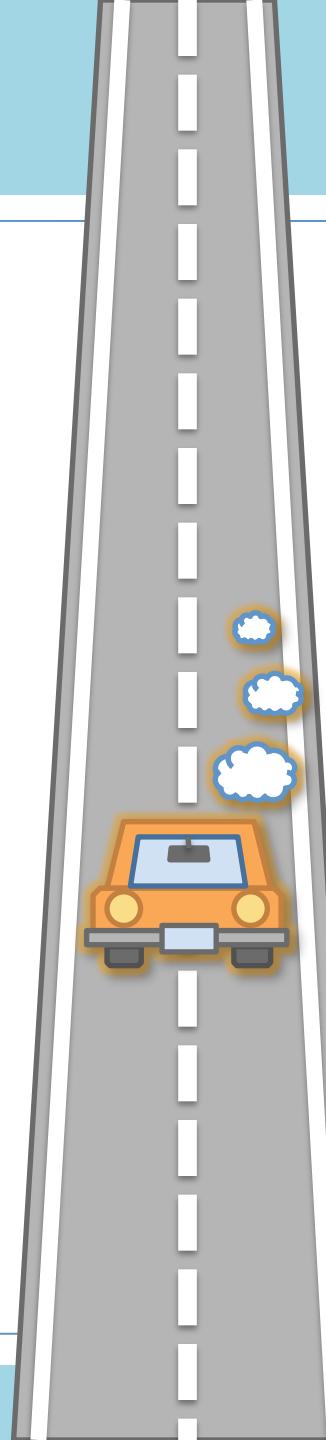
Q2. How to efficiently estimate model parameters ?



Idea (2): Multi-layer optimization (linear)

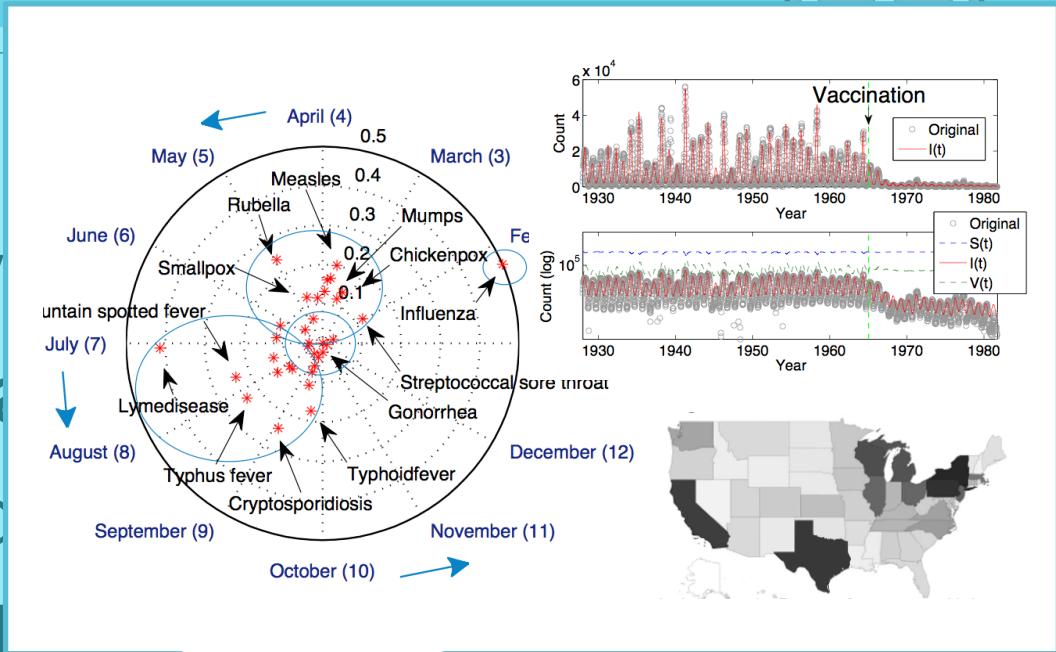
Roadmap

- ✓ Motivation
- ✓ Modeling power of FUNNEL
- ✓ Overview - main ideas
- ✓ Proposed model - idea #1
- ✓ Algorithm - idea #2
- Experiments
- Discussion
- Conclusions



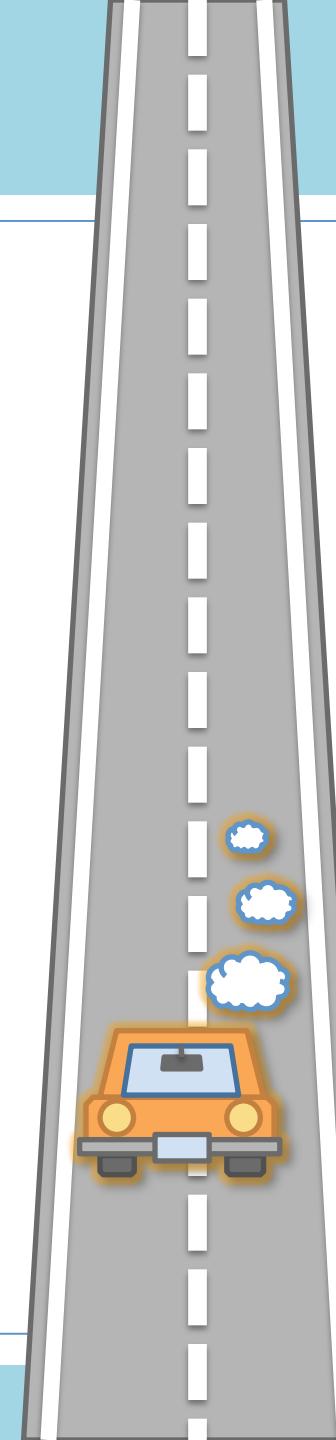
Roadmap

- ✓ Motivation
- ✓ Modeling power
- ✓ Overview - main idea
- ✓ Proposed model
- ✓ Algorithm - idea
- ✓ Experiments
- Discussion
- Conclusions



Roadmap

- ✓ Motivation
- ✓ Modeling power of FUNNEL
- ✓ Overview - main ideas
- ✓ Proposed model - idea #1
- ✓ Algorithm - idea #2
- ✓ Experiments
- Discussion
- Conclusions

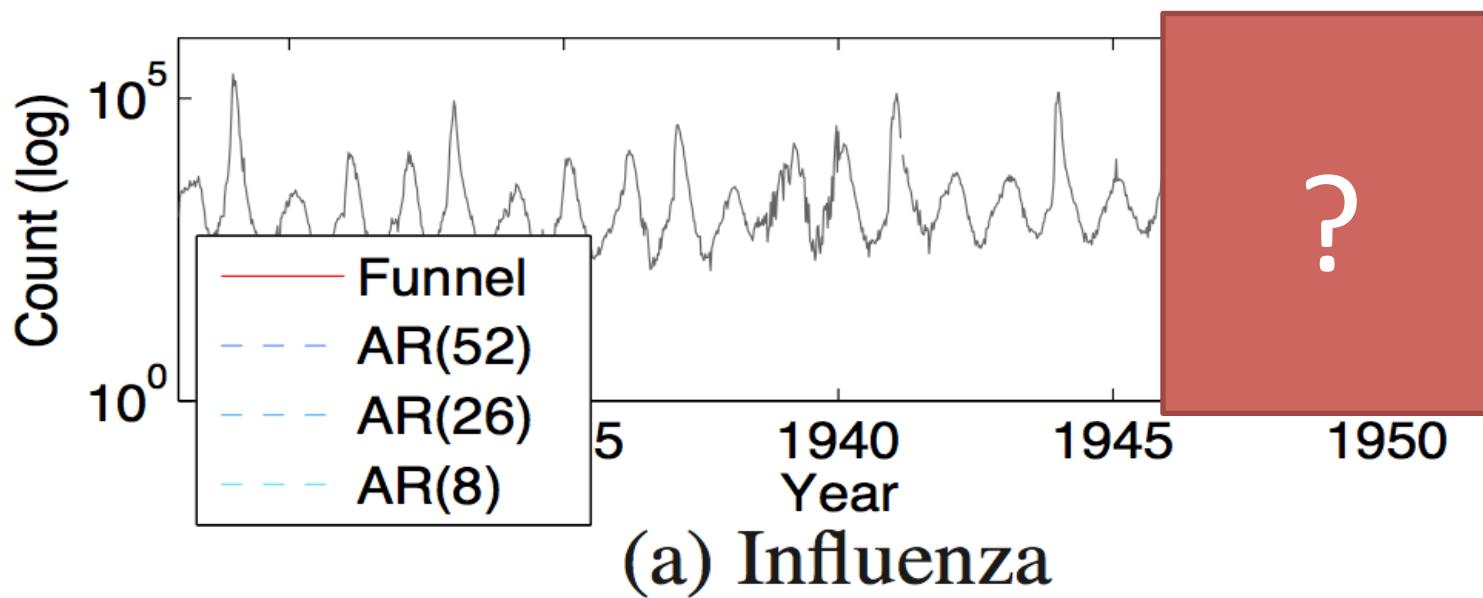


(a) FUNNEL at work - forecasting

Forecasting future epidemics

Train:
2/3 sequences

Forecast:
1/3 following years

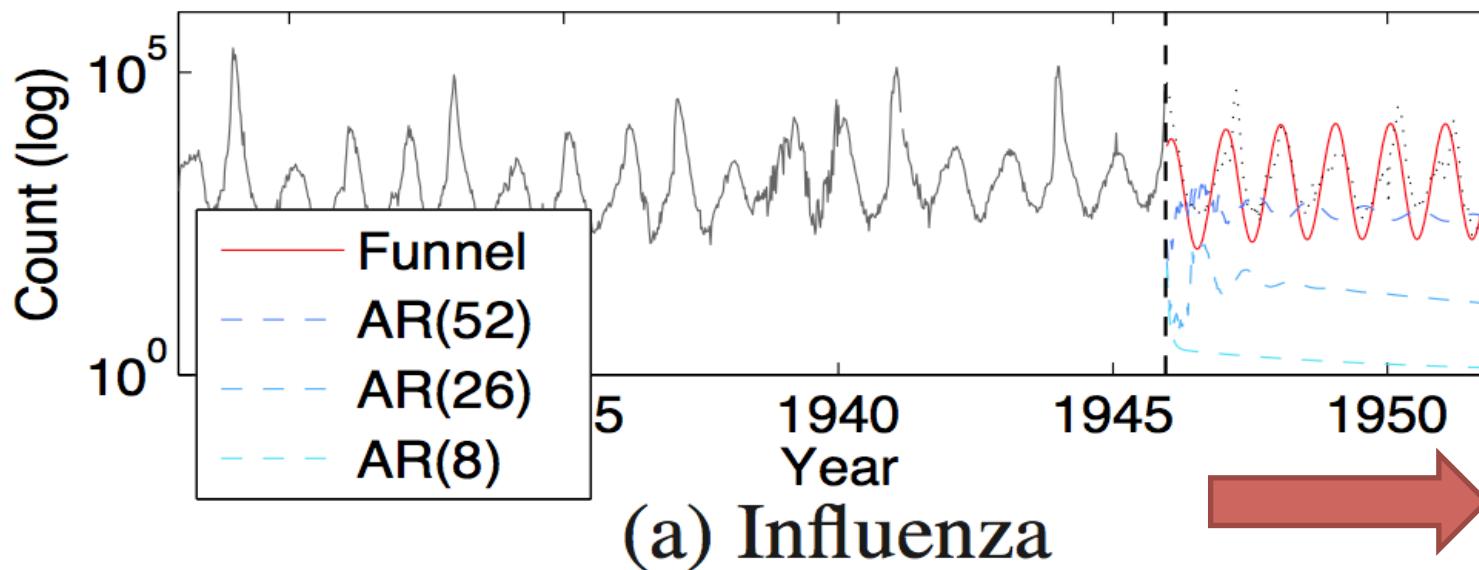


(a) FUNNEL at work - forecasting

Forecasting future epidemics

Train:
2/3 sequences

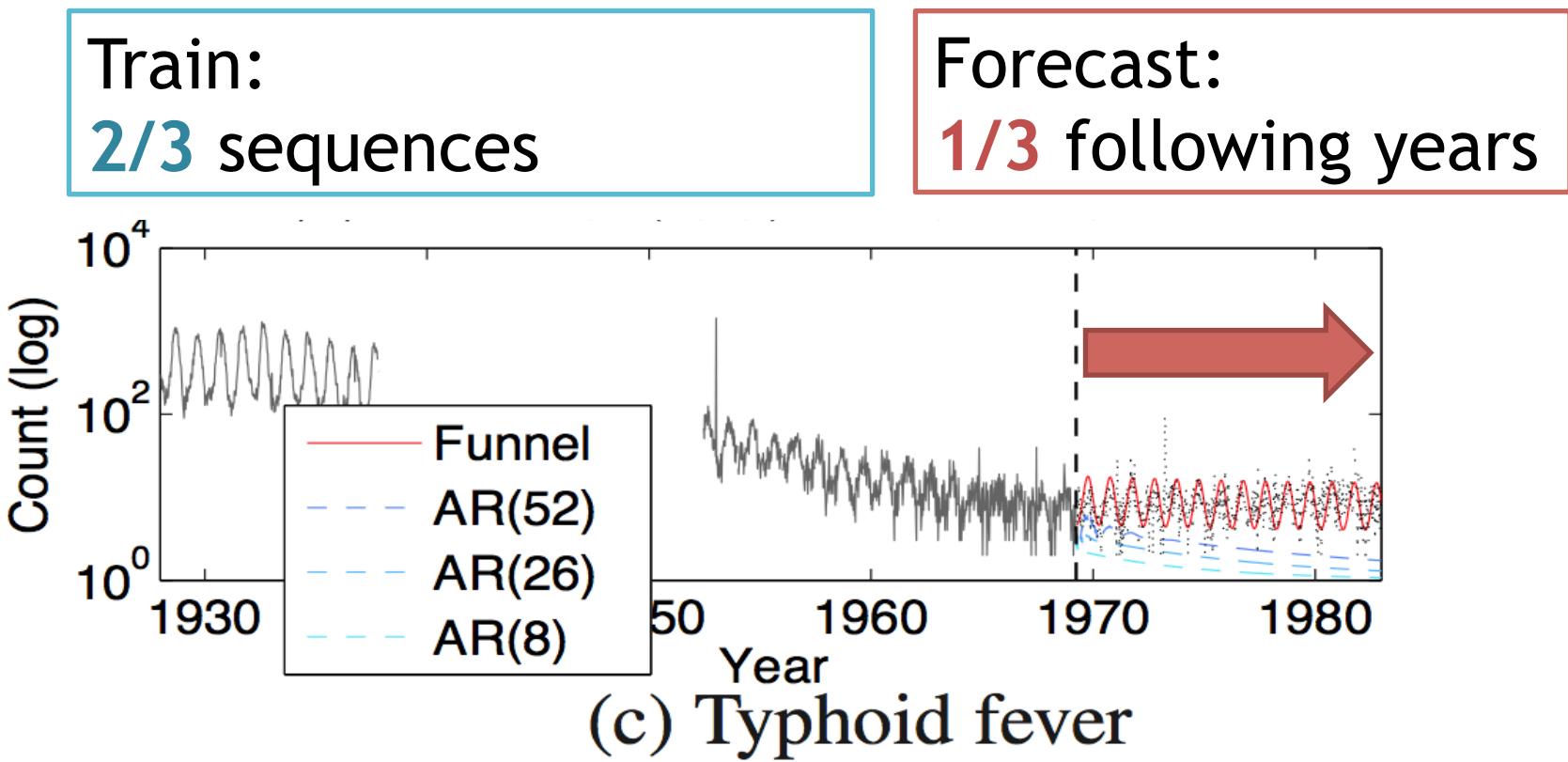
Forecast:
1/3 following years



Funnel can capture future epidemics (AR: fail)

(a) FUNNEL at work - forecasting

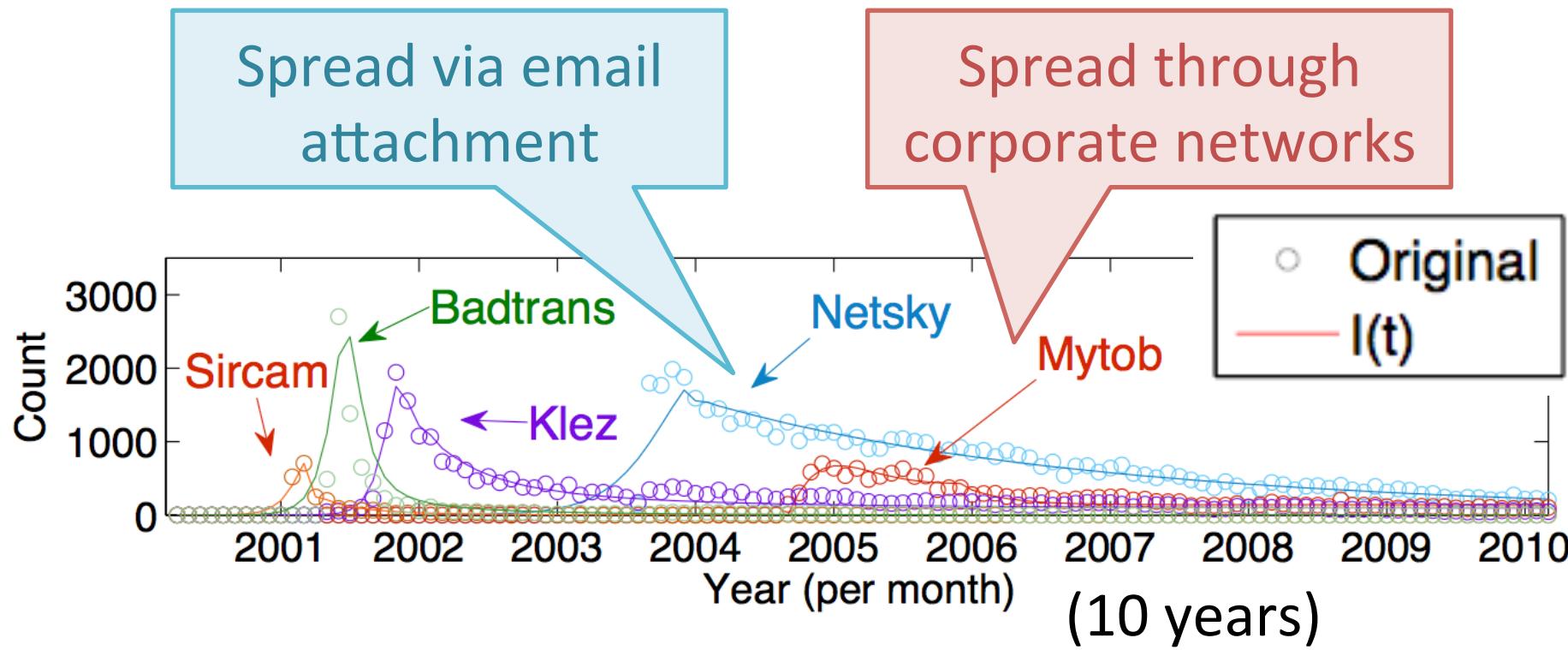
Forecasting future epidemics



Funnel can capture future epidemics (AR: fail)

(b) Generality of FUNNEL

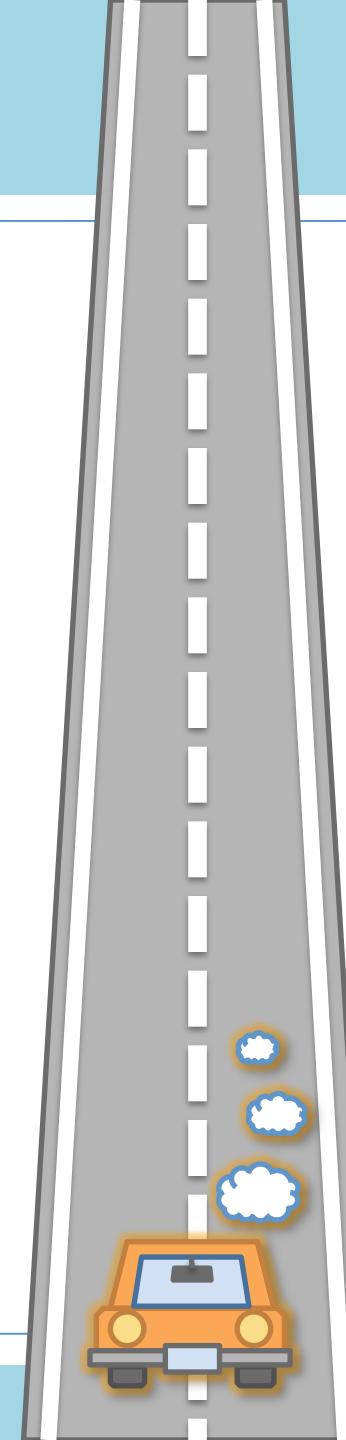
Epidemics on computer networks



Funnel is general: it fits computer virus very well!

Roadmap

- ✓ Motivation
- ✓ Modeling power of FUNNEL
- ✓ Overview - main ideas
- ✓ Proposed model - idea #1
- ✓ Algorithm - idea #2
- ✓ Experiments
- ✓ Discussion
- Conclusions



Conclusions

FUNNEL has the following advantages

✓ **Sense-making**

Captures all essential aspects:

✓ **Fully-automatic**

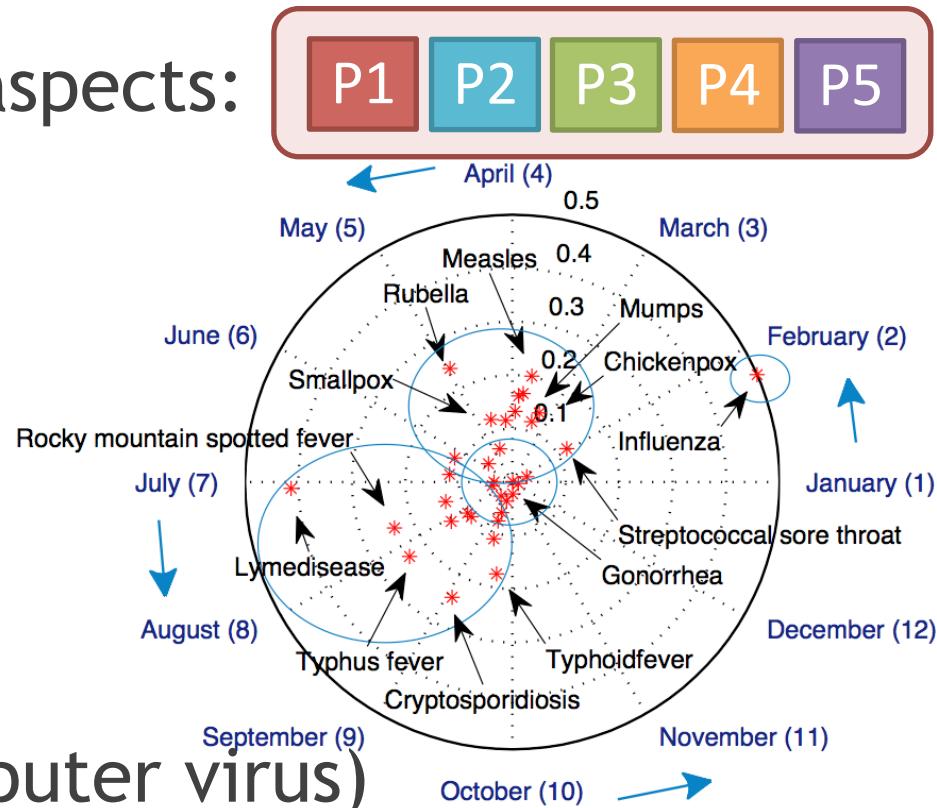
No training set

✓ **Scalable**

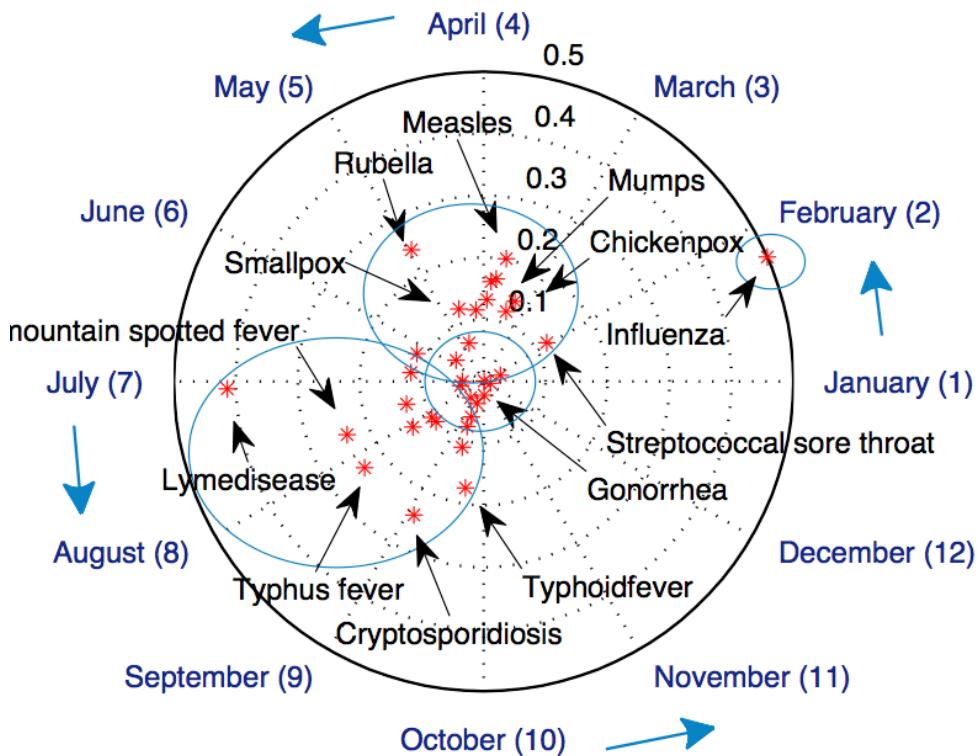
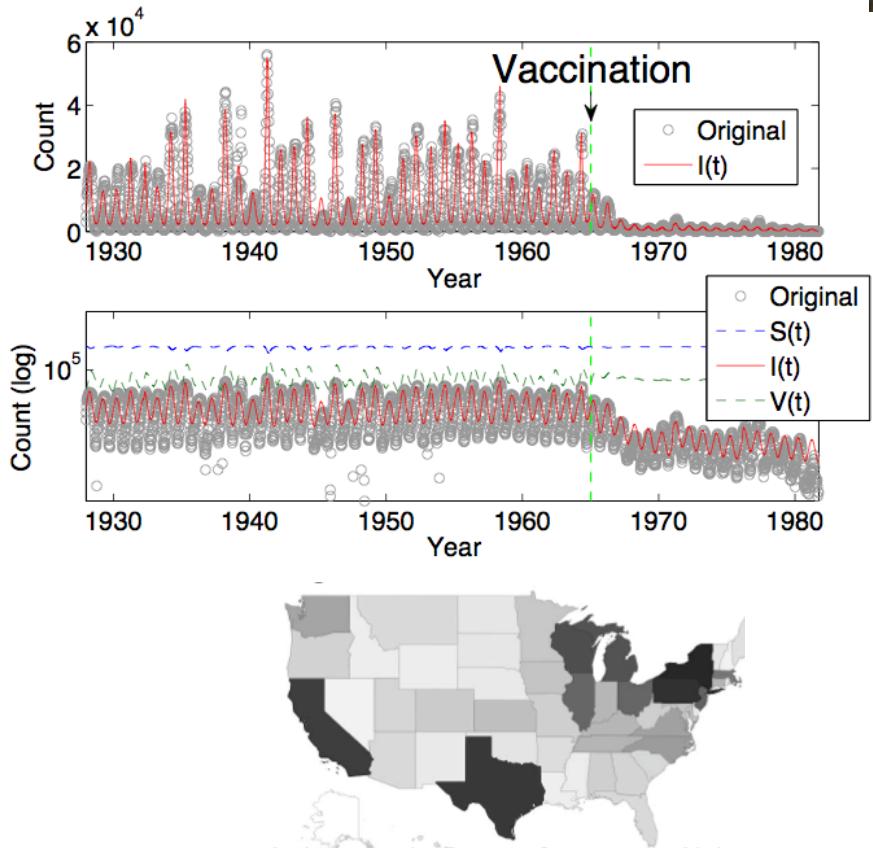
It scales linearly

✓ **General**

Real epidemics (+ computer virus)



Thank you!



Data: <http://www.tycho.pitt.edu/>

Code: <http://www.cs.kumamoto-u.ac.jp/~yasuko/software.html>

FUNNEL: Automatic Mining of Spatially Coevolving Epidemics

Yasuko Matsubara, Yasushi Sakurai (Kumamoto University)

Willem G. van Panhuis (University of Pittsburgh)

Christos Faloutsos (CMU)

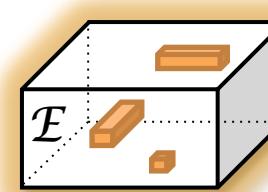


Proposed model: FUNNEL-full

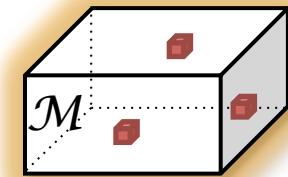
Details

What's the difference??

External shock vs. Mistake

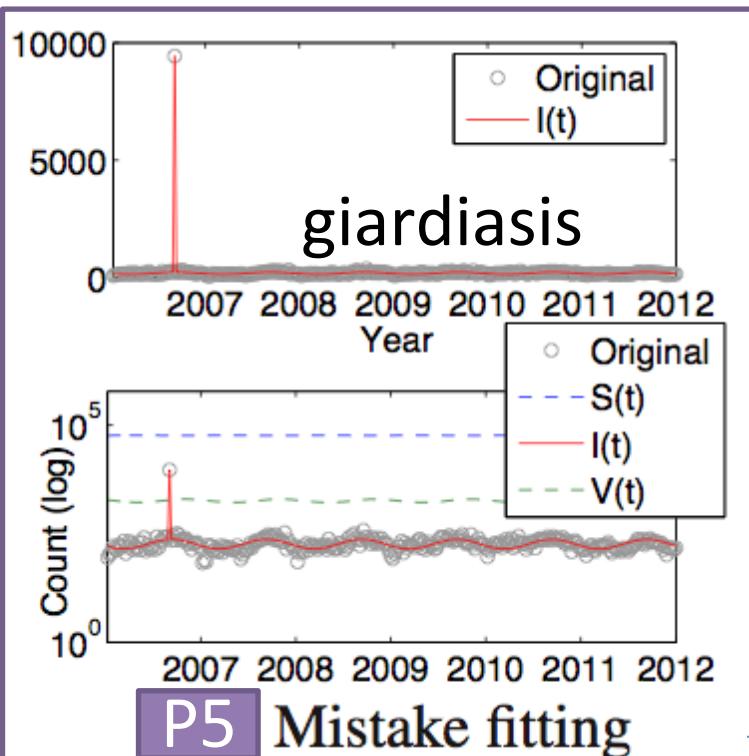
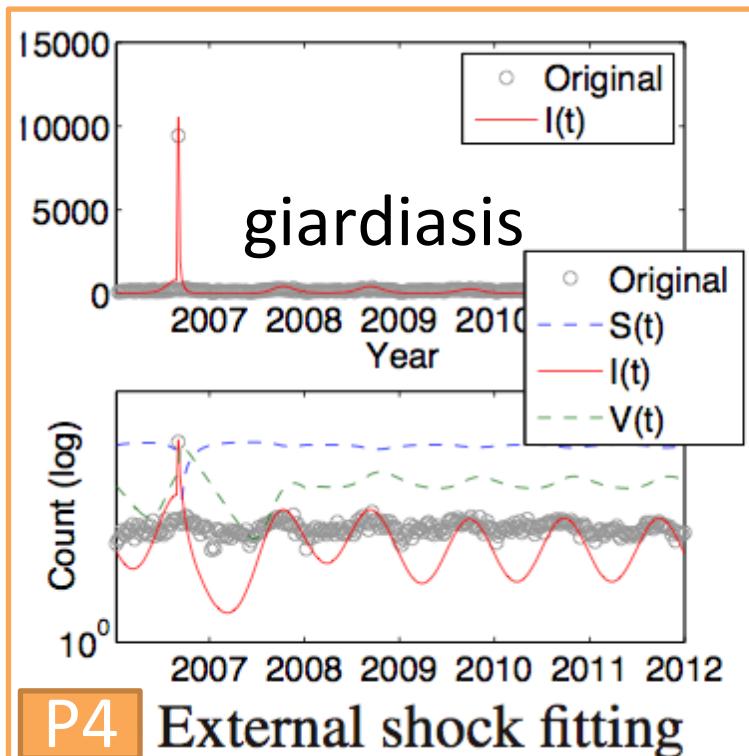


vs.



P4

P5

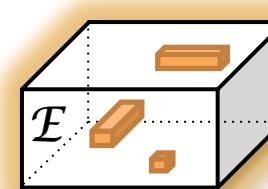


Proposed model: FUNNEL-full

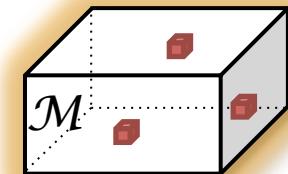
Details

What's the difference??

External shock vs. Mistake

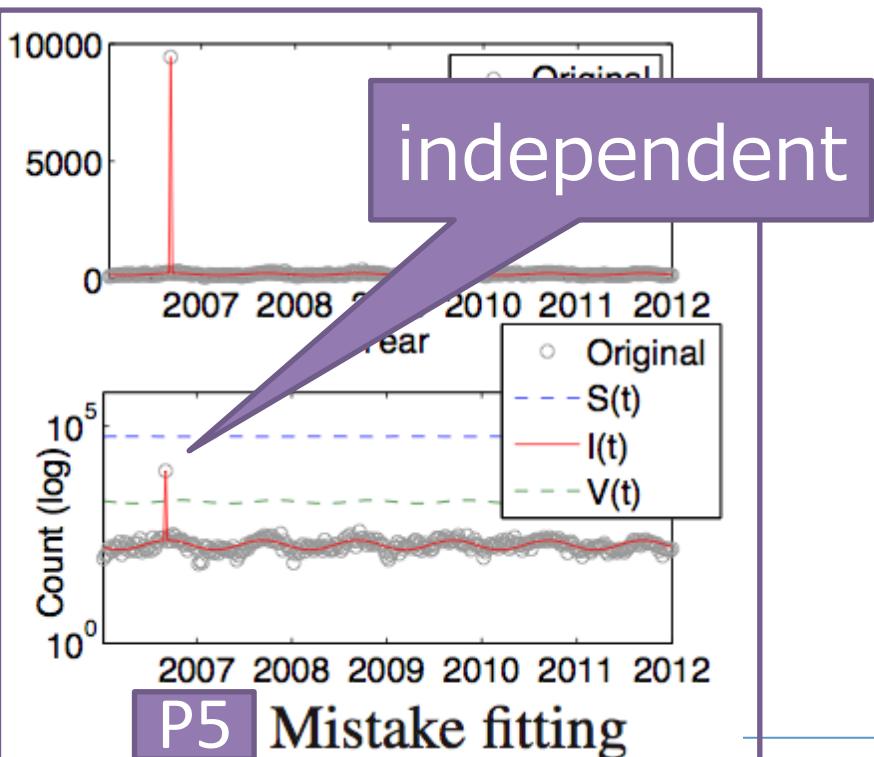
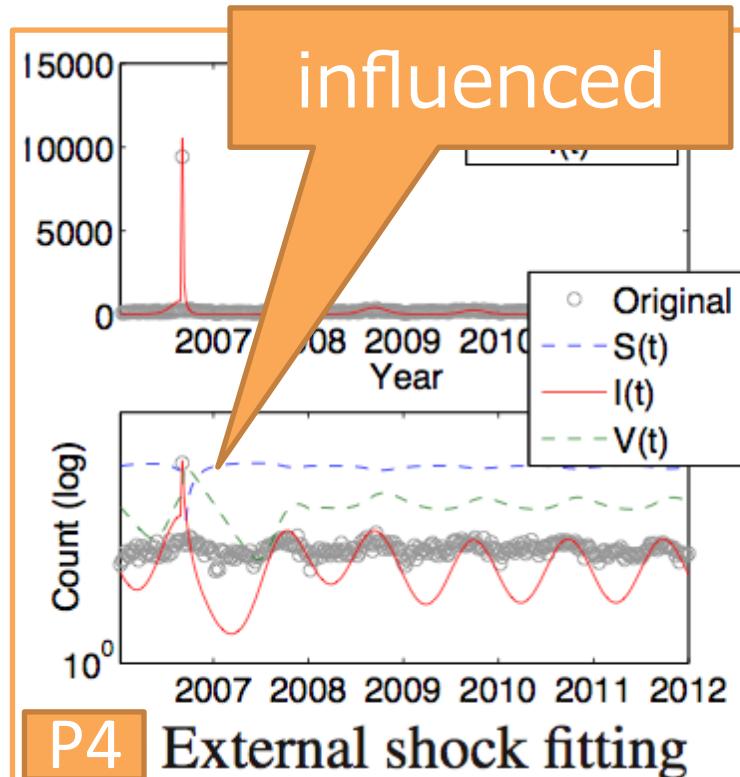


vs.



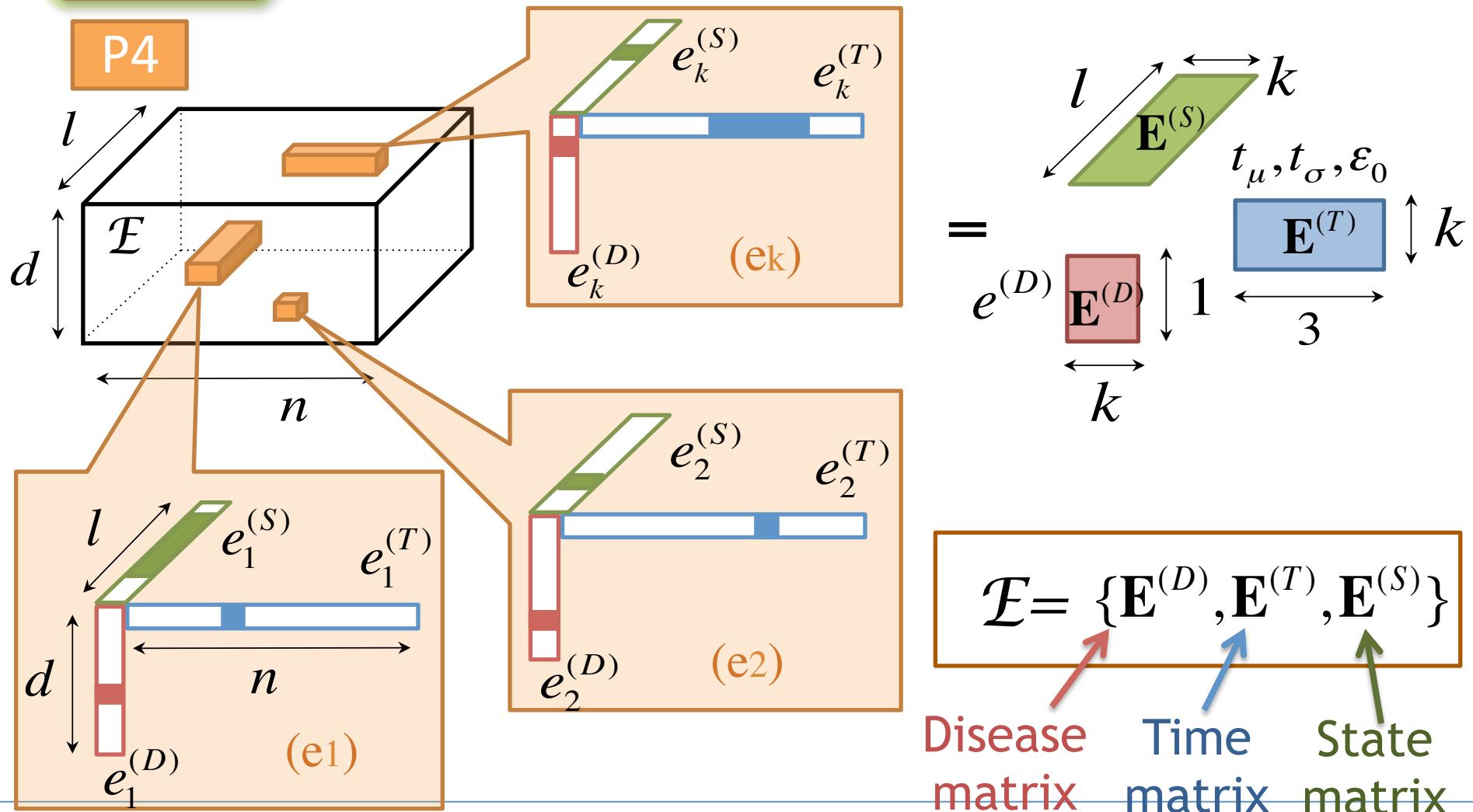
P4

P5



Proposed model: FUNNEL-full

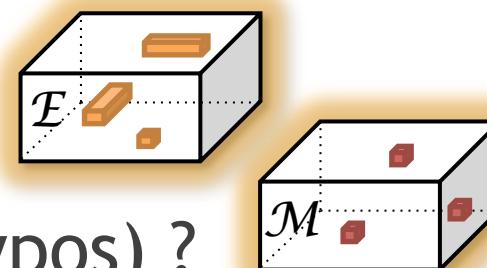
Details



Idea (1): Model description cost

Q1. How should we

- find “external shocks” ?
- ignore “mistakes” (i.e., typos) ?



Idea (1) : Model description cost

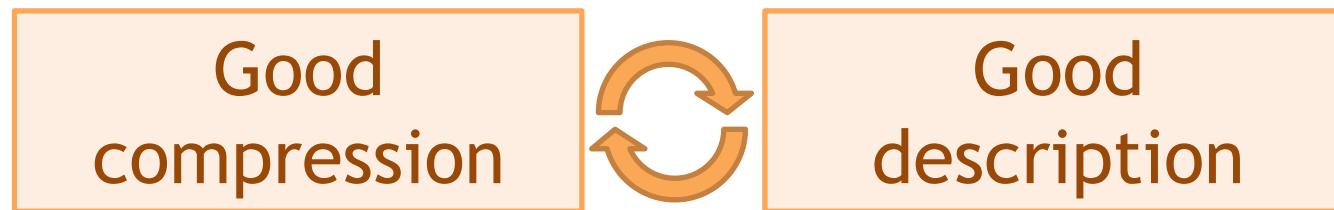
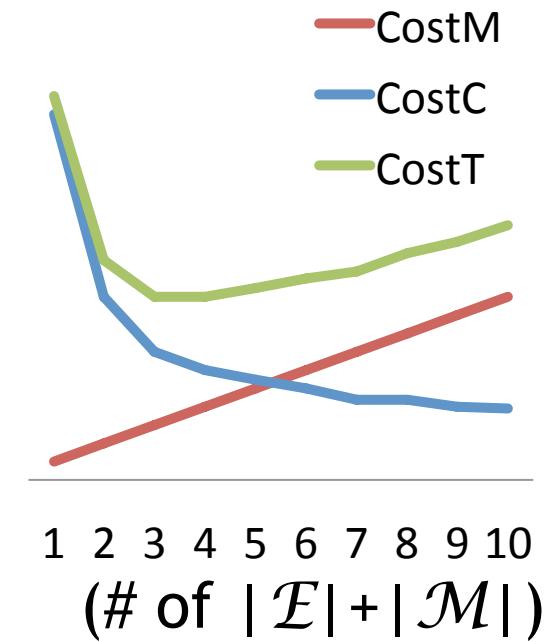
- Minimize coding cost
- find “optimal” # of externals/mistakes
- “automatically”

Idea (1): Model description cost

Idea: Minimize encoding cost!

$$\min (\text{Cost}_M(\mathcal{F}) + \text{Cost}_C(\mathcal{X}|\mathcal{F}))$$

Model cost Coding cost



Idea (1): Model description cost

Details

Total cost of tensor \mathcal{X} , given \mathcal{F}

$$\mathcal{F} = \{\mathbf{B}, \mathbf{R}, \mathbf{N}, \mathcal{E}, \mathcal{M}\}.$$

$$\begin{aligned} Cost_T(\mathcal{X}; \mathcal{F}) &= \log^*(d) + \log^*(l) + \log^*(n) \\ &+ Cost_M(\mathbf{B}) + Cost_M(\mathbf{R}) + Cost_M(\mathbf{N}) \\ &+ Cost_M(\mathcal{E}) + Cost_M(\mathcal{M}) + Cost_C(\mathcal{X}|\mathcal{F}) \end{aligned}$$

Idea (1): Model description cost

Total cost of tensor \mathcal{X} , given \mathcal{F}

Details

$$\mathcal{F} = \{\mathbf{B}, \mathbf{R}, \mathbf{N}, \mathcal{E}, \mathcal{M}\}$$

Dimensions of \mathcal{X}

$$Cost_T(\mathcal{X}; \mathcal{F}) = \log^*(d) + \log^*(l) + \log^*(n)$$

$$+ Cost_M(\mathbf{B}) + Cost_M(\mathbf{R}) + Cost_M(\mathbf{N})$$

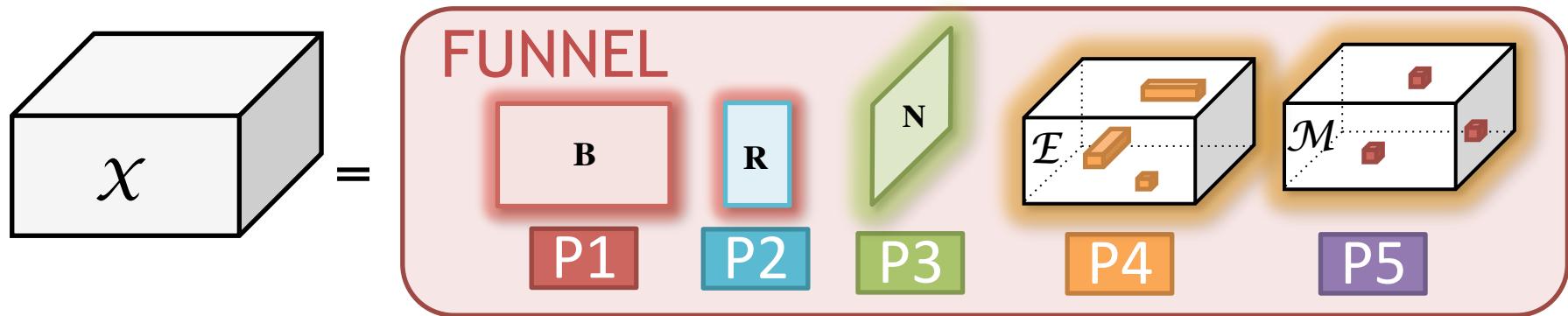
$$+ Cost_M(\mathcal{E}) + Cost_M(\mathcal{M}) + Cost_C(\mathcal{X} | \mathcal{F})$$

Model description
cost of \mathcal{F}

Coding cost
of \mathcal{X} given \mathcal{F}

Idea (2): Multi-layer optimization

Q2. How to efficiently estimate model parameters ?

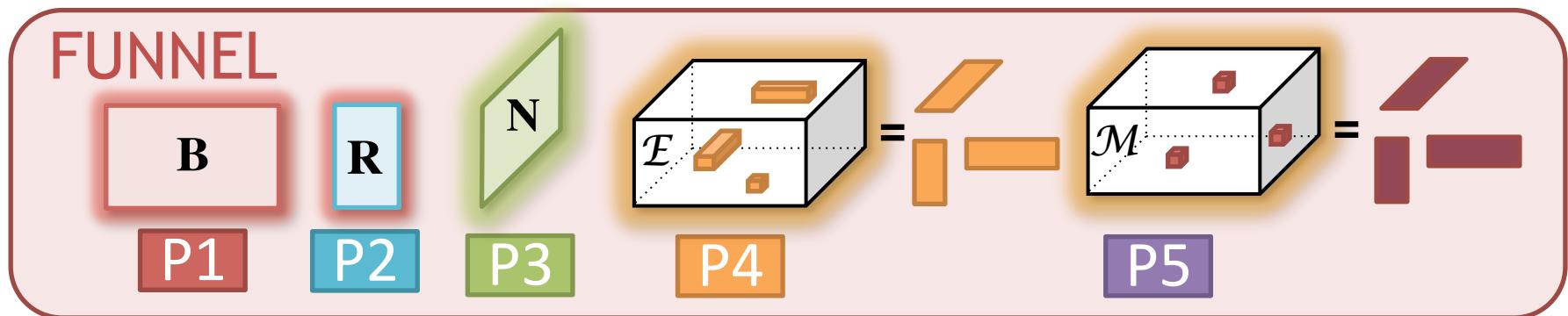


Idea (2): Multi-layer optimization

- Find “optimal” solution w.r.t.
 - Global level parameters
 - Local level parameters

Idea (2): Multi-layer optimization

Find “optimal” solution w.r.t. Global Local

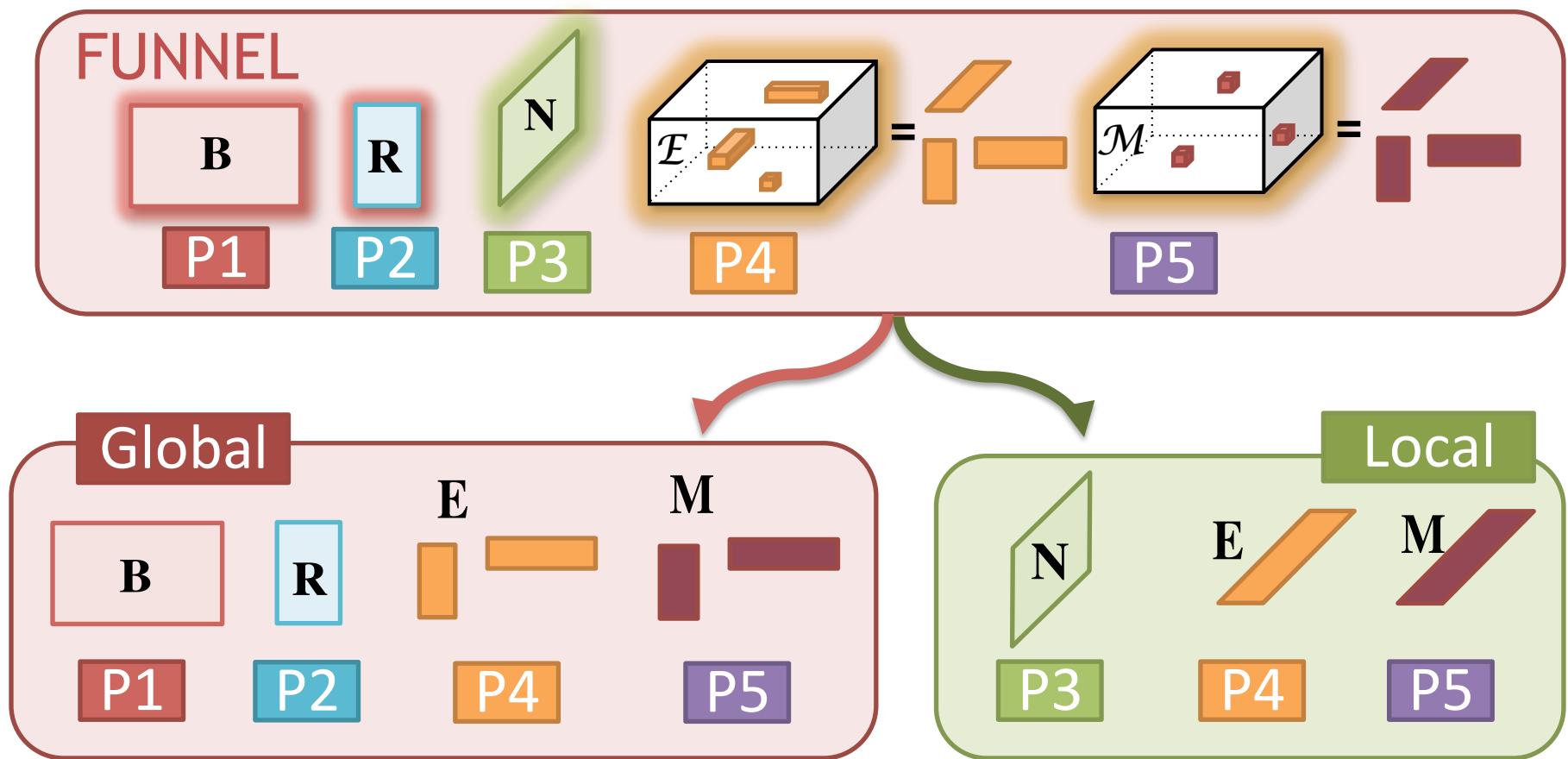


Global

Local

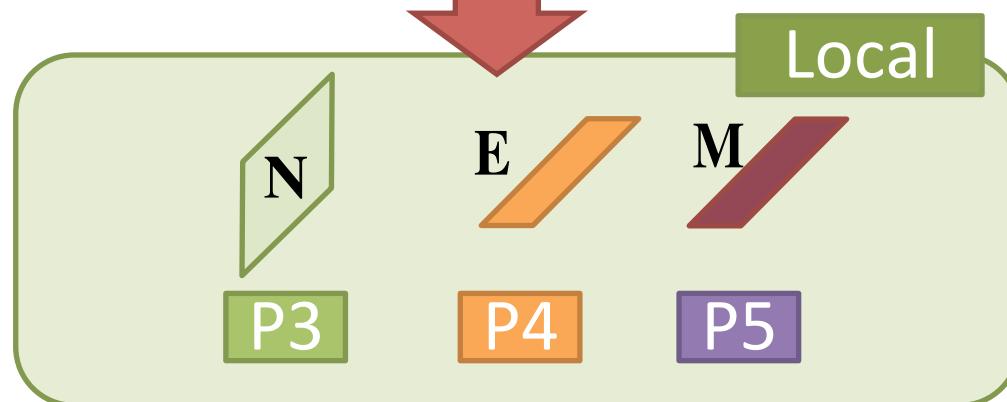
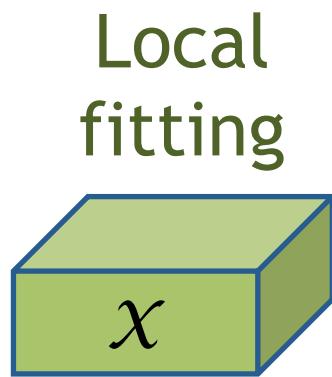
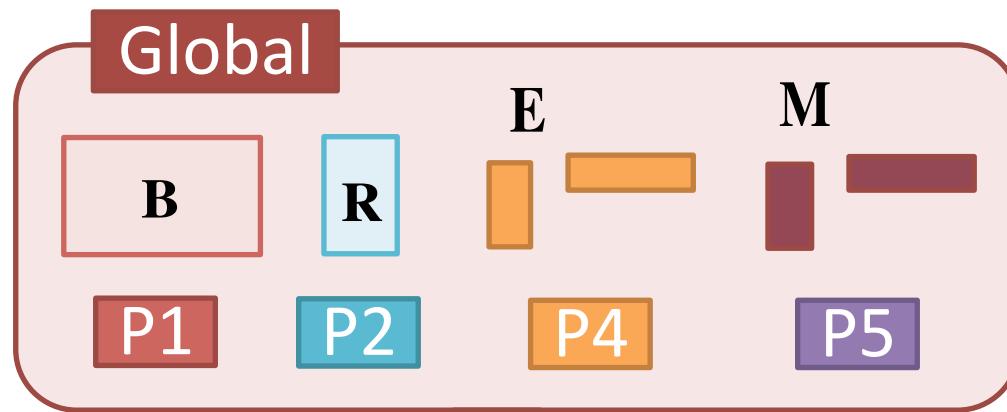
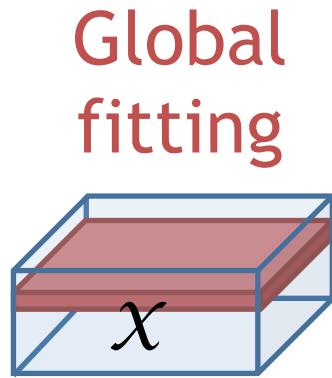
Idea (2): Multi-layer optimization

Find “optimal” solution w.r.t. Global Local



Idea (2): Multi-layer optimization

Multi-layer fitting algorithm



Experiments

We answer the following questions...

Q1. Sense-making

Can it help us understand the given epidemics?

Q2. Accuracy

How well does it match the data?

Q3. Scalability

How does it scale in terms of computational time?

Q1. Sense-making

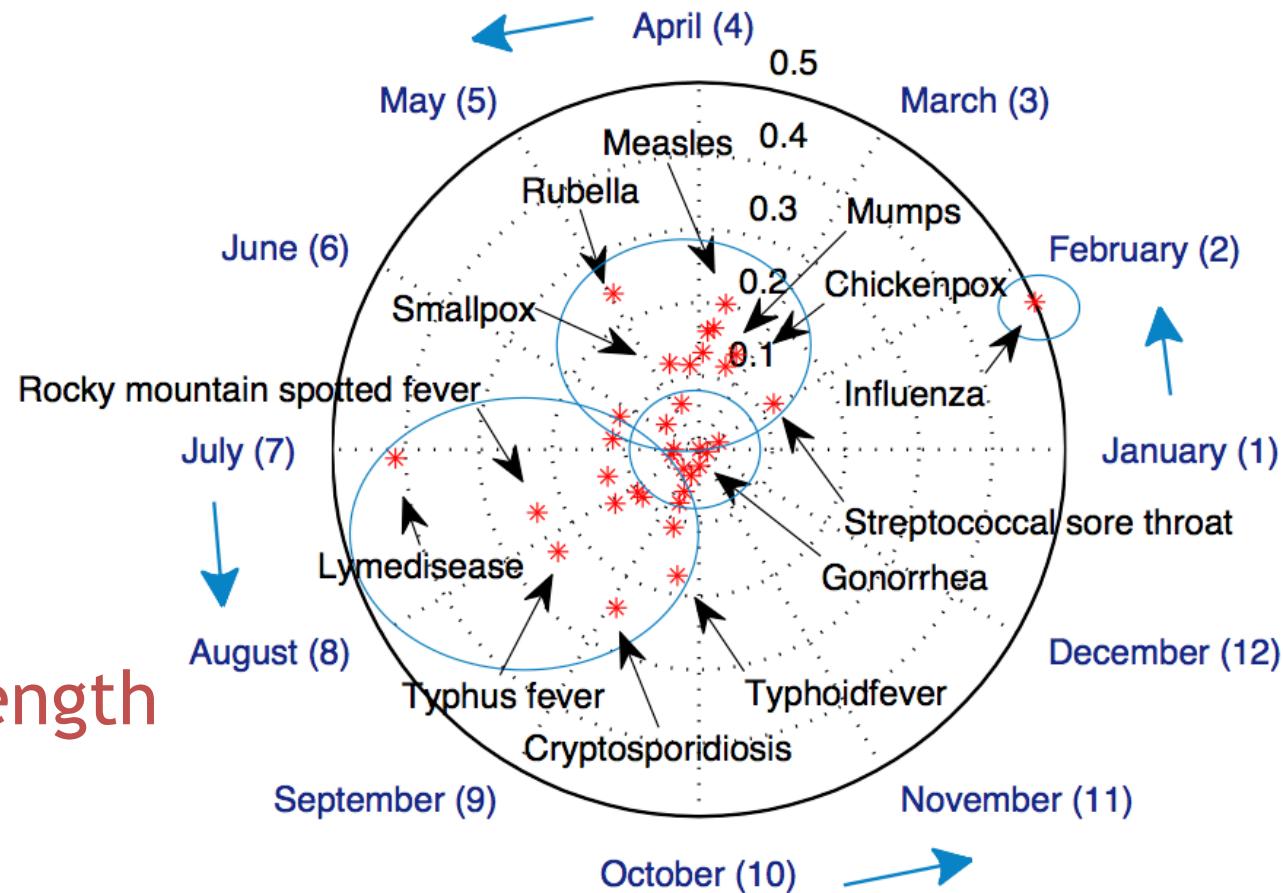
Our preliminary observations:

- P1 yearly periodicity
- P2 disease reduction effects
- P3 area specificity and sensitivity
- P4 external shock events
- P5 mistakes, incorrect values

Q1. Sense-making

P1

Disease seasonality



Radius:

seasonality strength

Angle:

peak season

Q1. Sense-making

P1

Disease seasonality

Children's
in spring

Rocky mountain spotted fever

July (7)

August (8)

Tick-borne
in summer

May (5)

June (6)

Smallpox

Lymedisease

Typhus fever

Cryptosporidiosis

October (10)

April (4)

0.5

0.4

0.3

0.2

0.1

Measles

Rubella

March (1)

Mumps

Chickenpox

Influenza

Influenza
in Feb.

February (2)

January (1)

Streptococcal sore throat

Gonorrhea

December (12)

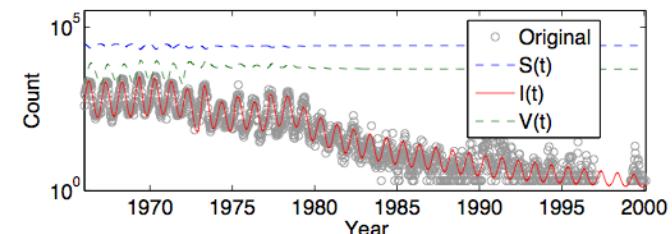
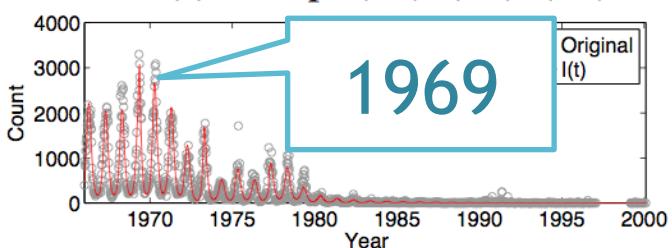
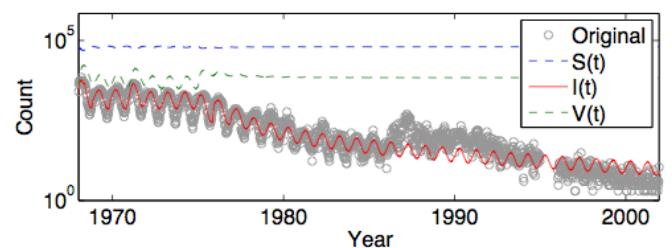
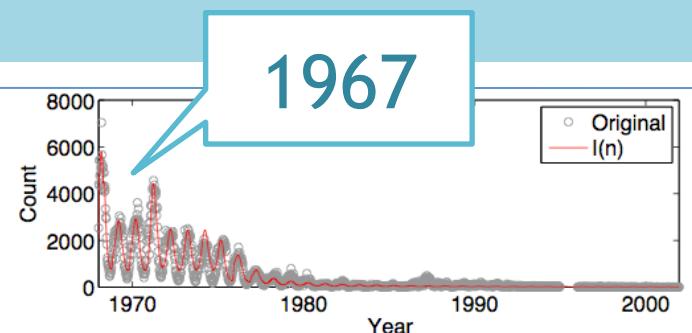
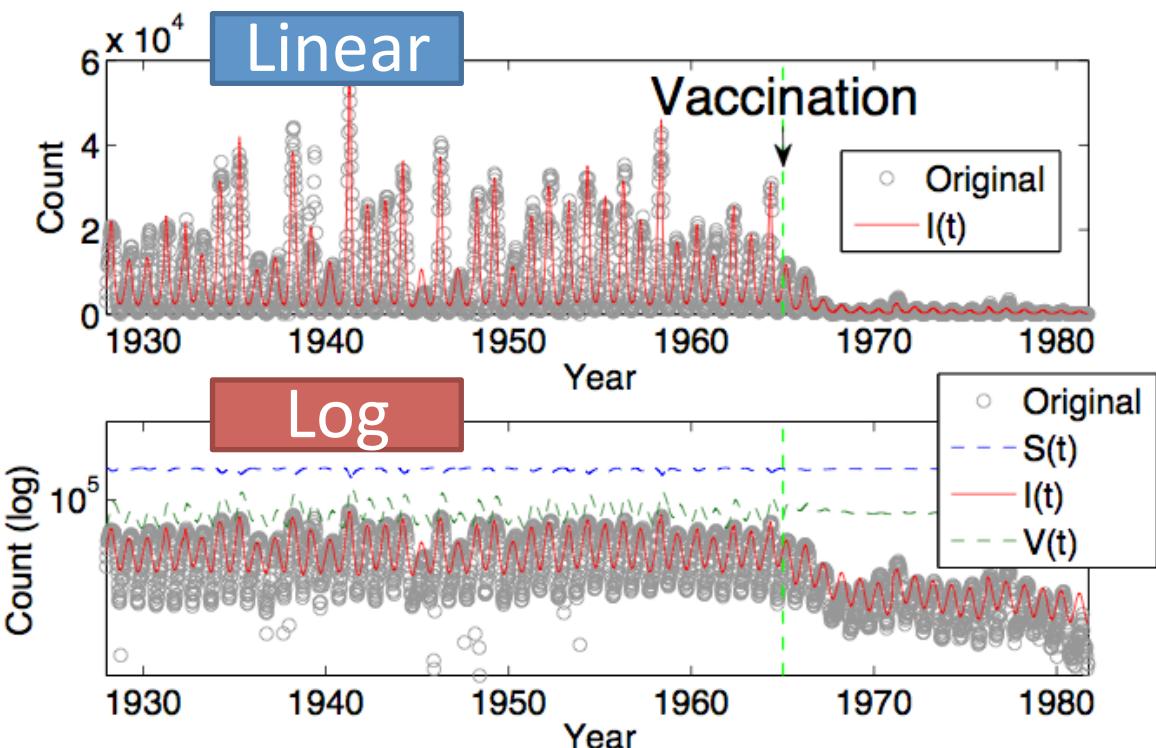
Gonorrhea
no periodicity

Q1. Sense-making

P2

Disease reduction effect

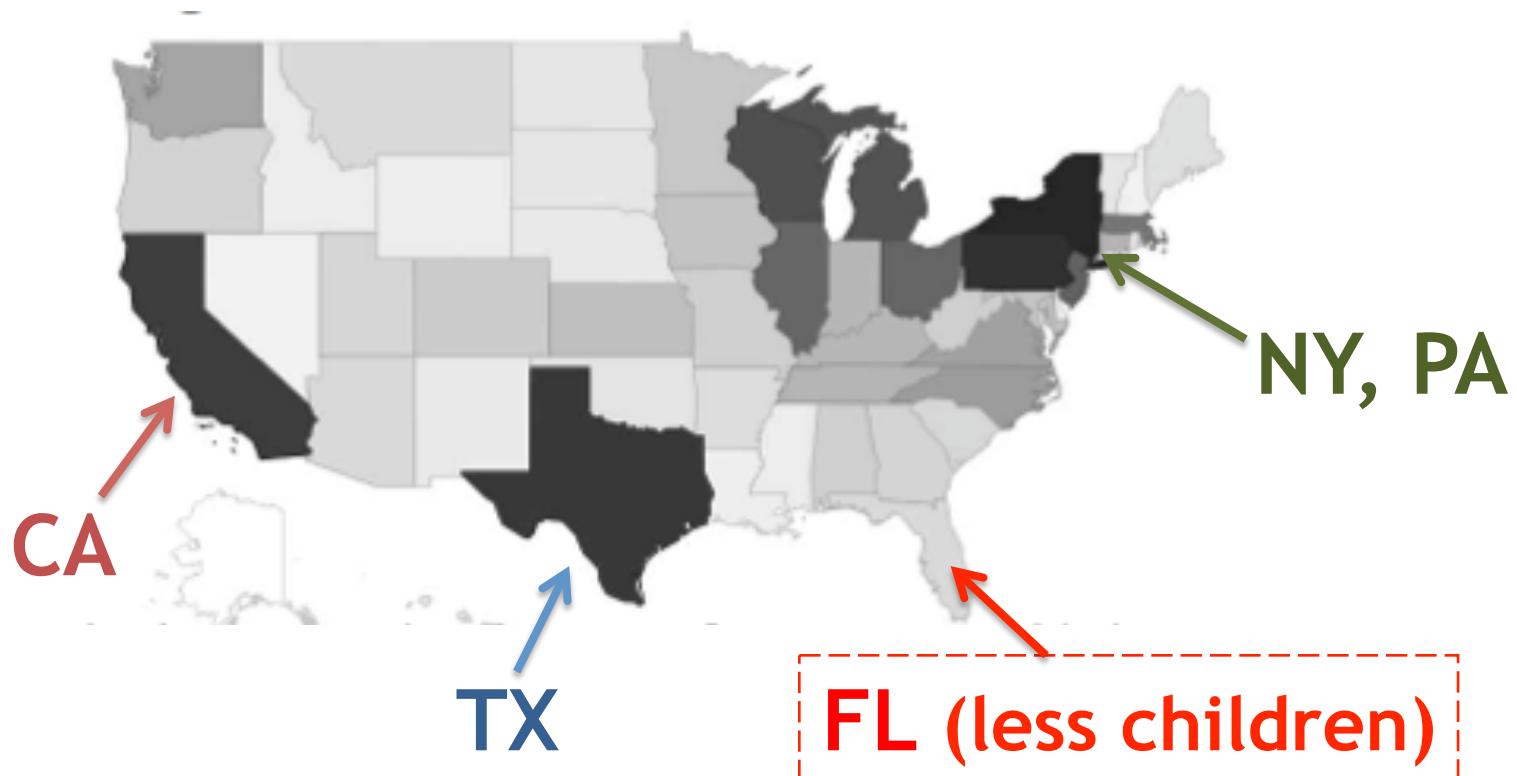
Measles (vaccine licensure: 1963)



Q1. Sense-making

P3 area specificity and sensitivity

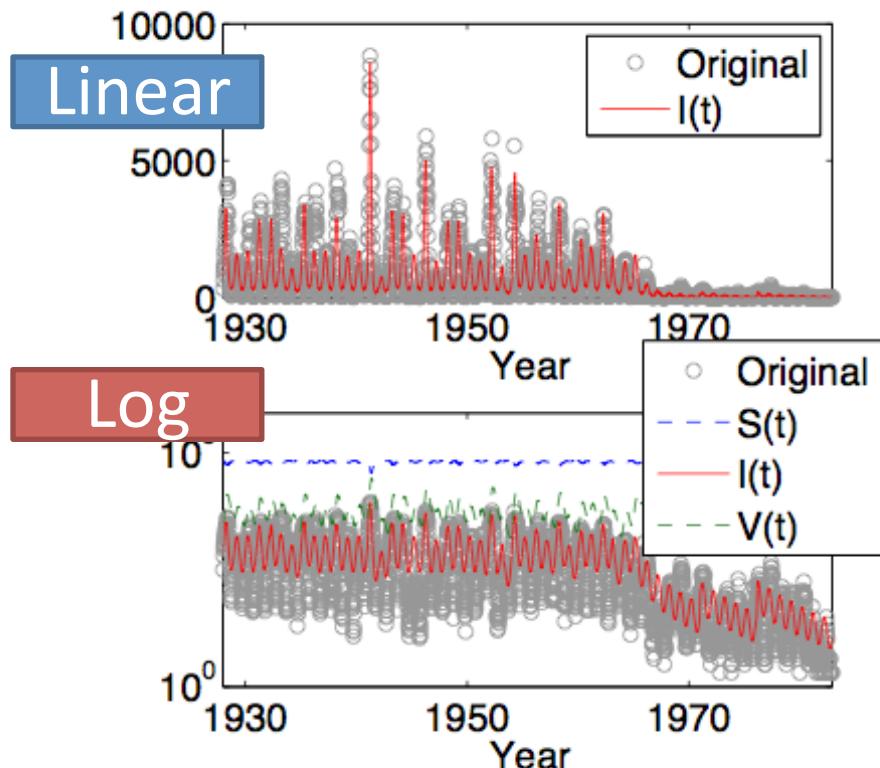
Potential population of susceptibles (measles)



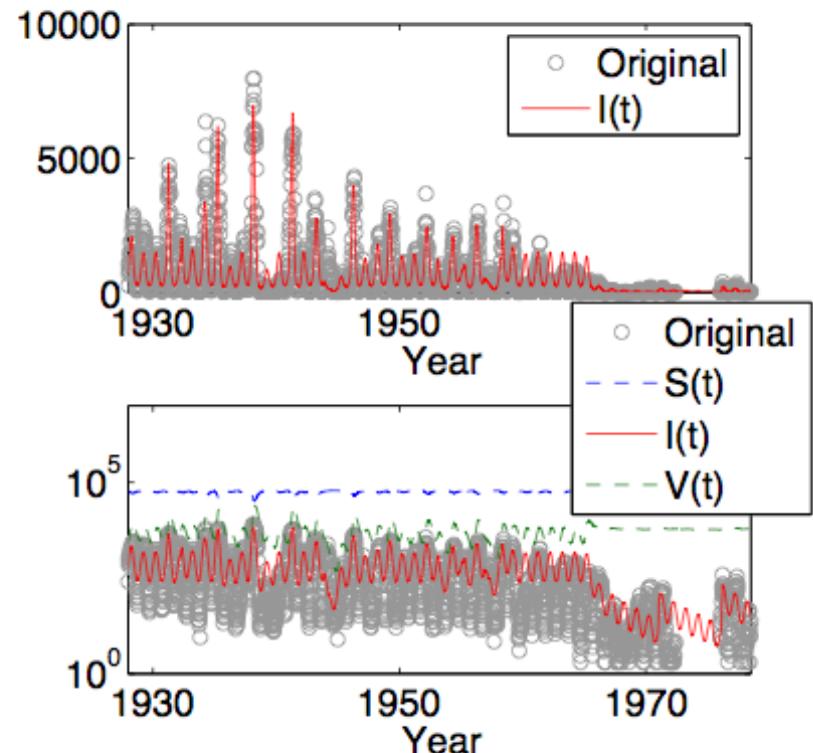
Q1. Sense-making

P3 area specificity and sensitivity

Measles in NY and PA



(a) New York State (NY)



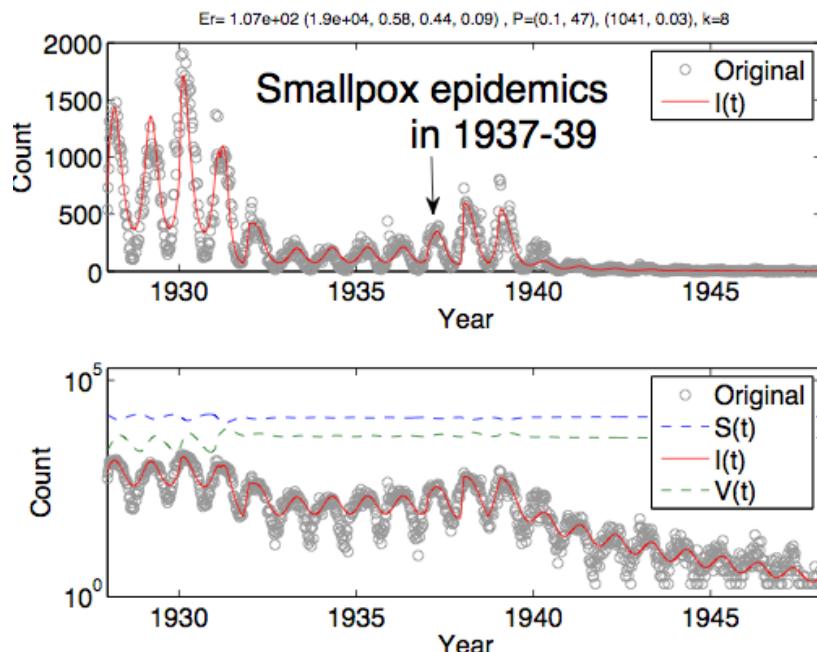
(b) Pennsylvania (PA)

Q1. Sense-making

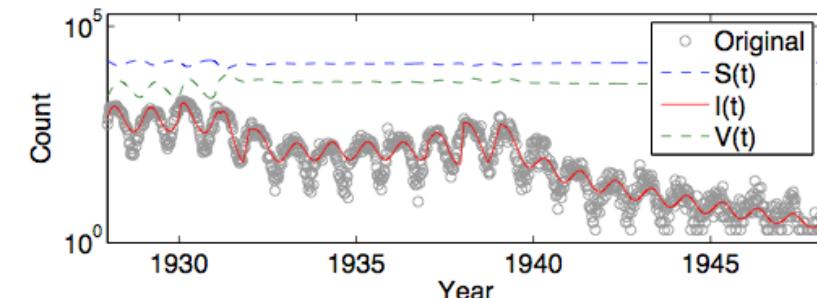
P4

external shock events

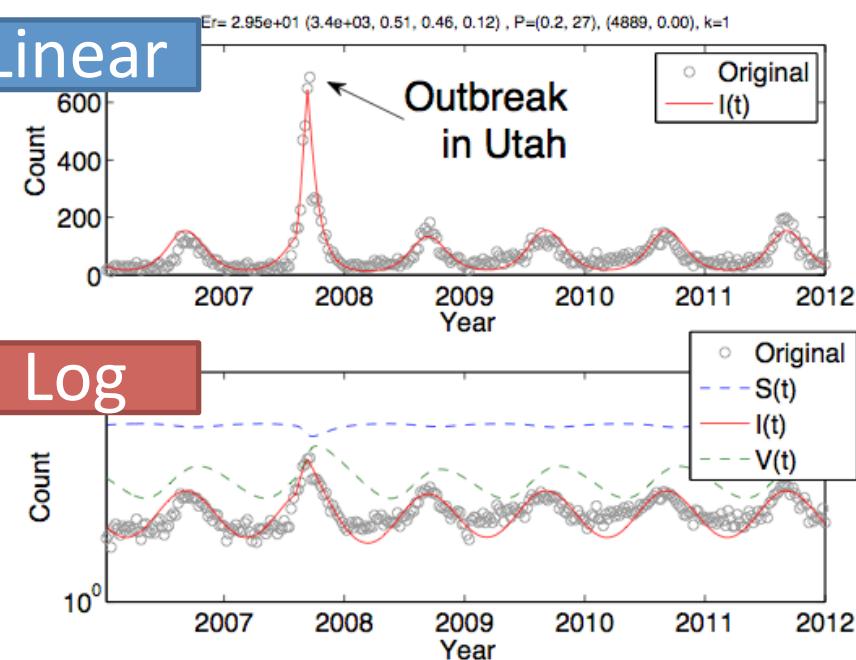
We can detect external shocks “automatically” !!



Linear



Log



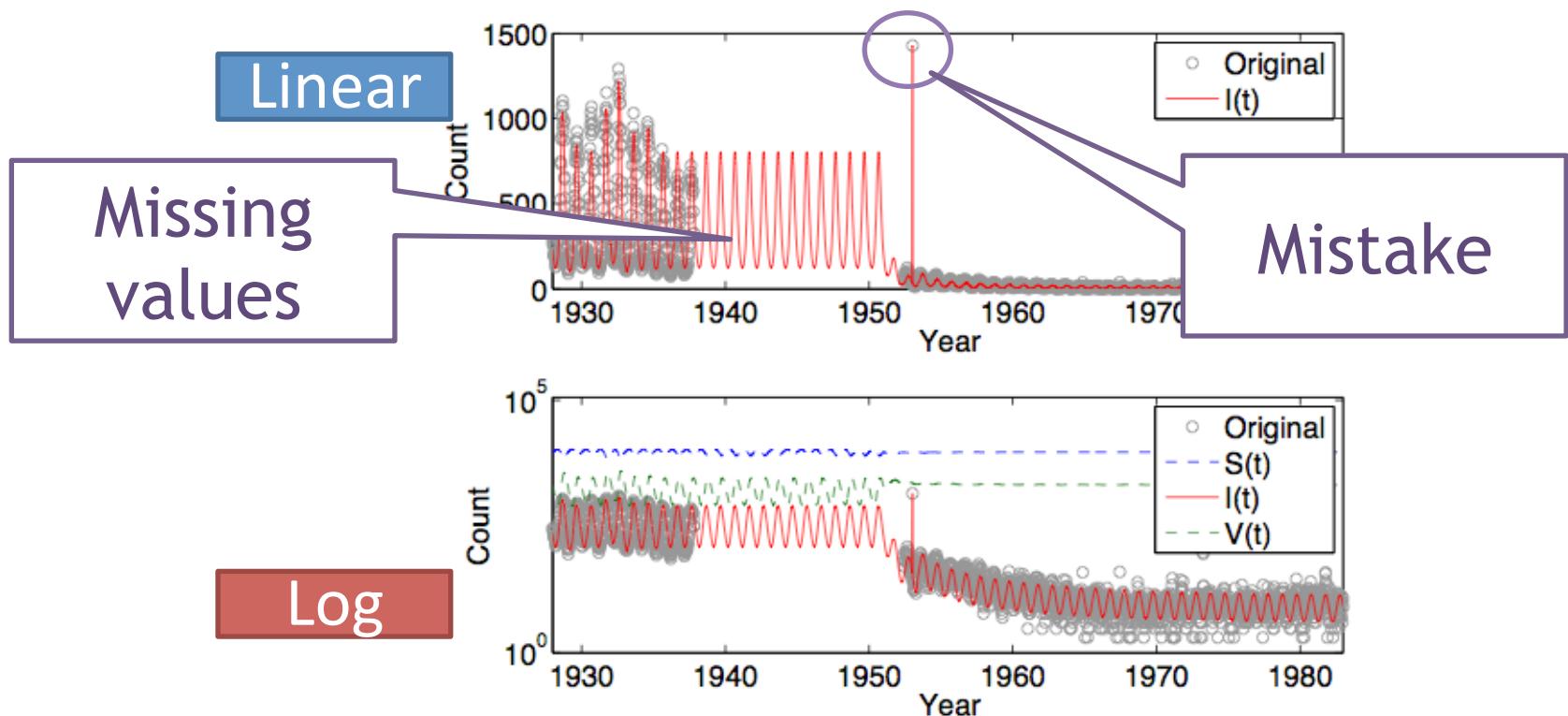
(h) Smallpox (P1), (P2), (P4)

(j) Cryptosporidiosis (P1), (P3), (P4)

Q1. Sense-making

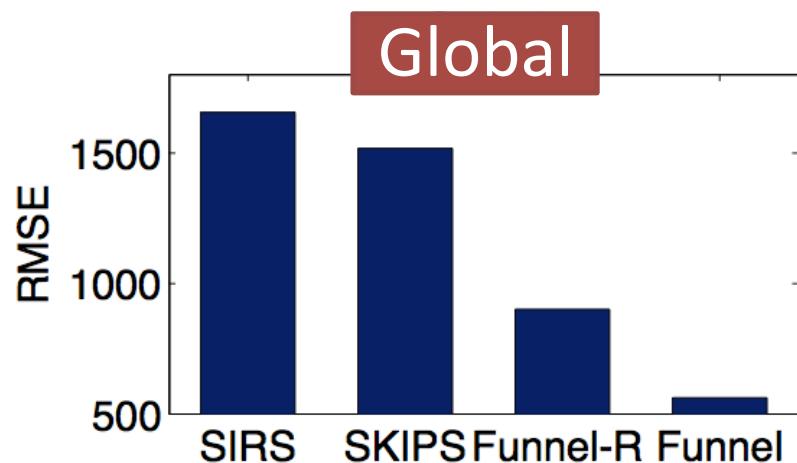
P5 mistakes, incorrect values

We can also detect typos “automatically” !!

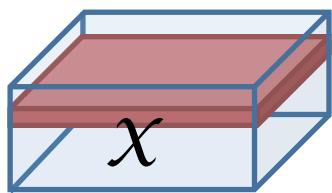


Q2. Accuracy

Fitting accuracy for Global Local sequences
(lower is better)

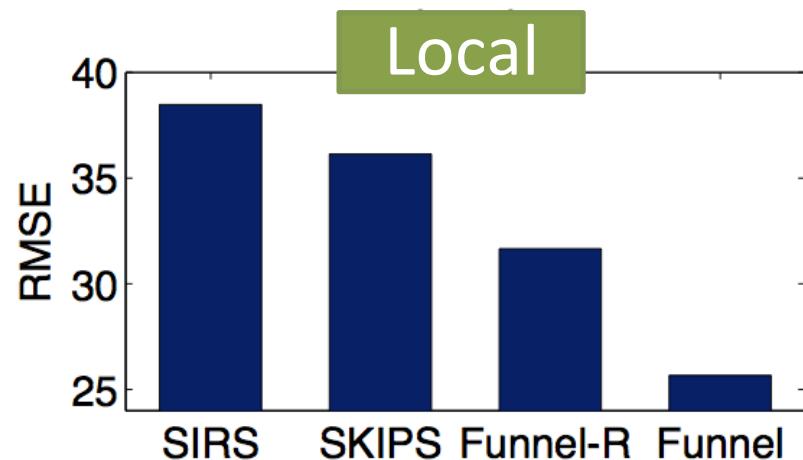


(a) Global fitting

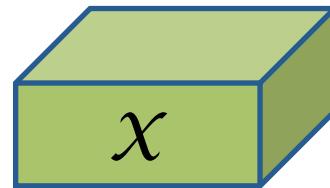


$$\{\bar{x}_i(t)\}_{i,t}^{d,n}$$

$$\bar{x}_i(t) = \sum_{j=1}^l x_{ij}(t)$$



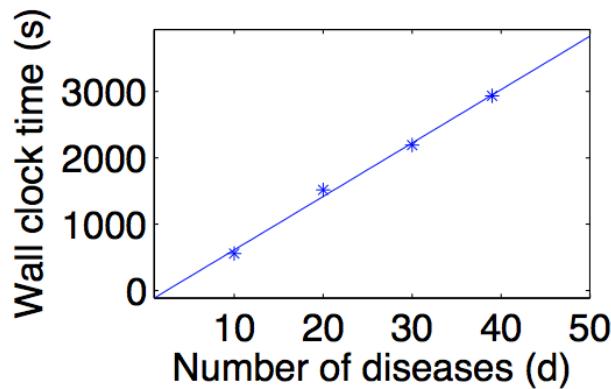
(b) Local fitting



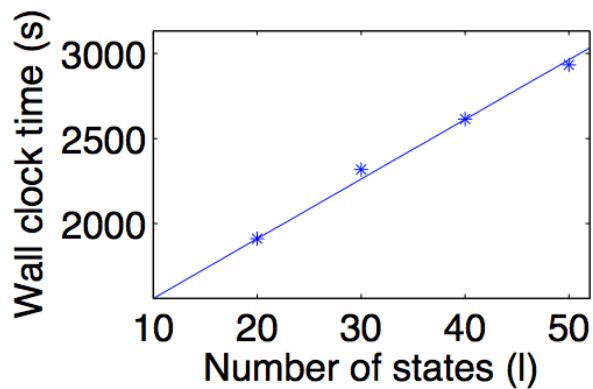
$$\{x_{ij}(t)\}_{i,j,t}^{d,l,n}$$

Q3. Scalability

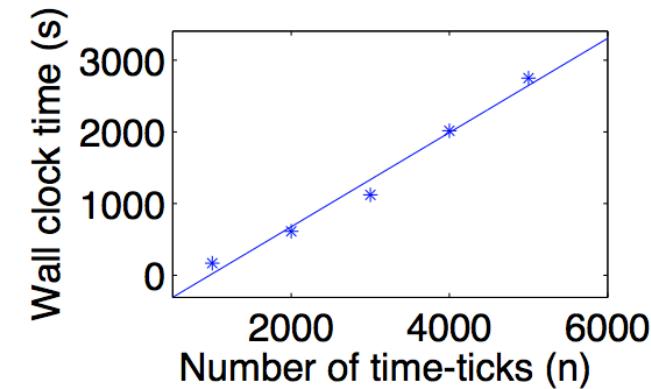
Wall clock time vs. diseases , states , Time



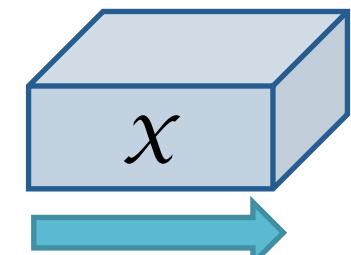
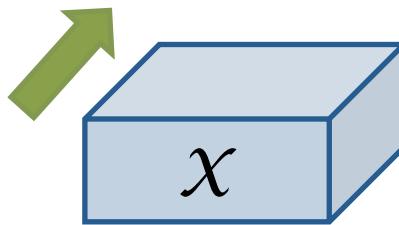
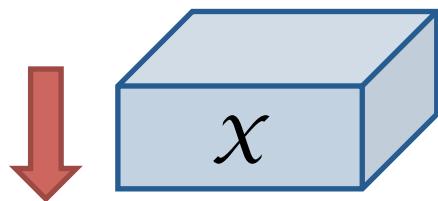
(a) Diseases (d)



(b) States (l)



(c) Duration (n)



FunnelFit is linear w.r.t. data size : $O(dln)$