

FUNNEL: Automatic Mining of Spatially Coevolving Epidemics

Yasuko Matsubara, Yasushi Sakurai (Kumamoto University)

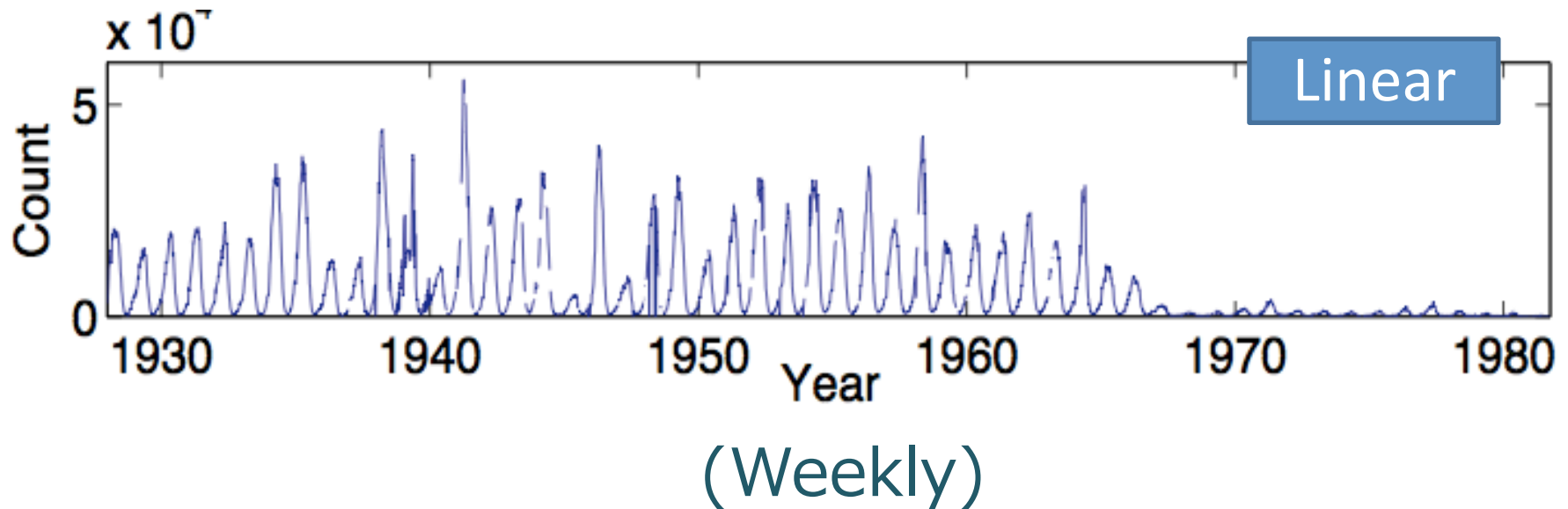
Willem G. van Panhuis (University of Pittsburgh)

Christos Faloutsos (CMU)



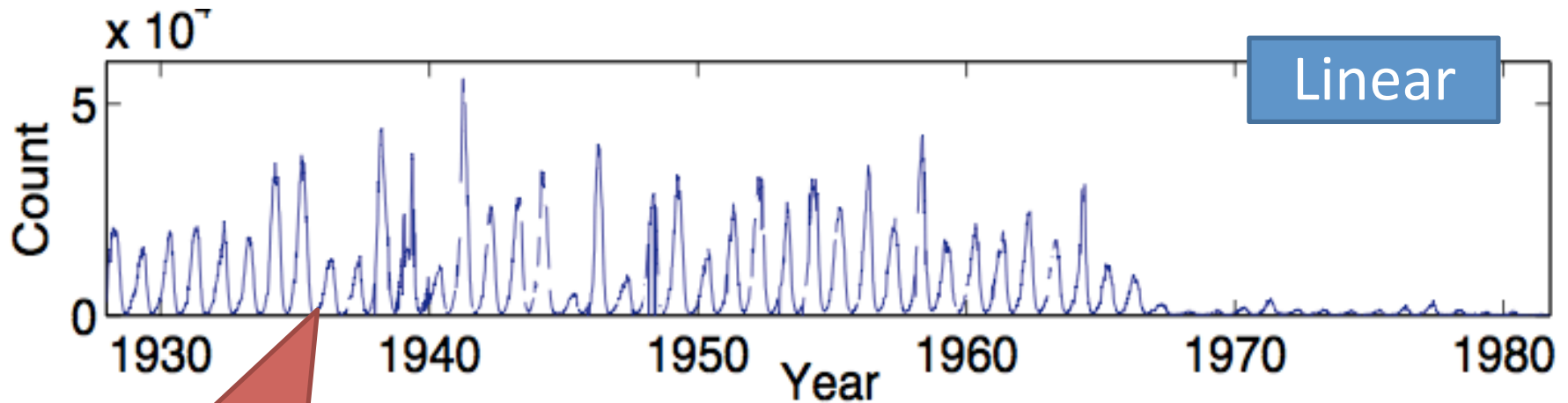
Motivation

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



Motivation

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

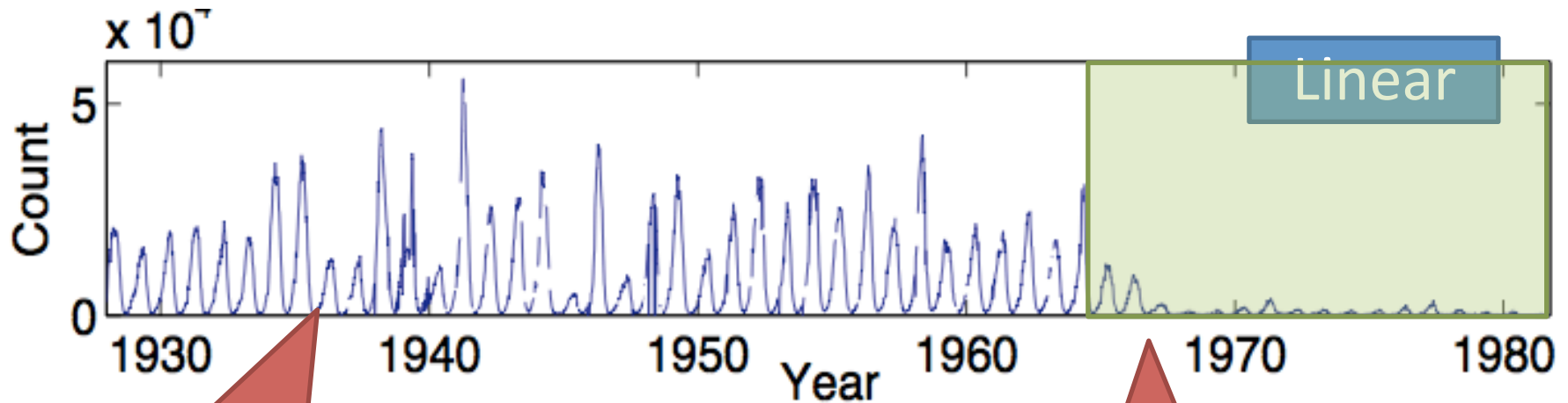


Yearly
periodicity

(Weekly)

Motivation

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



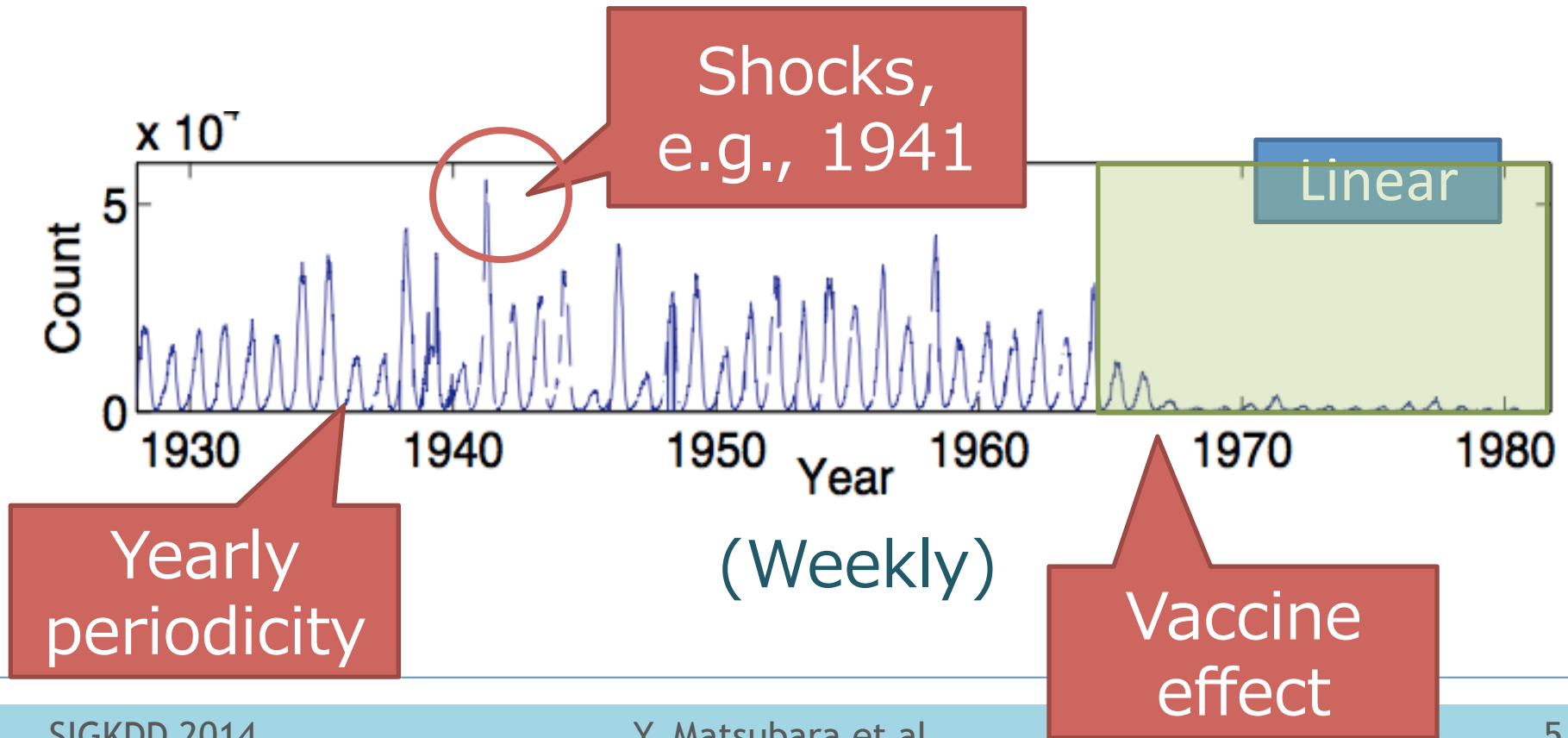
Yearly
periodicity

(Weekly)

Vaccine
effect

Motivation

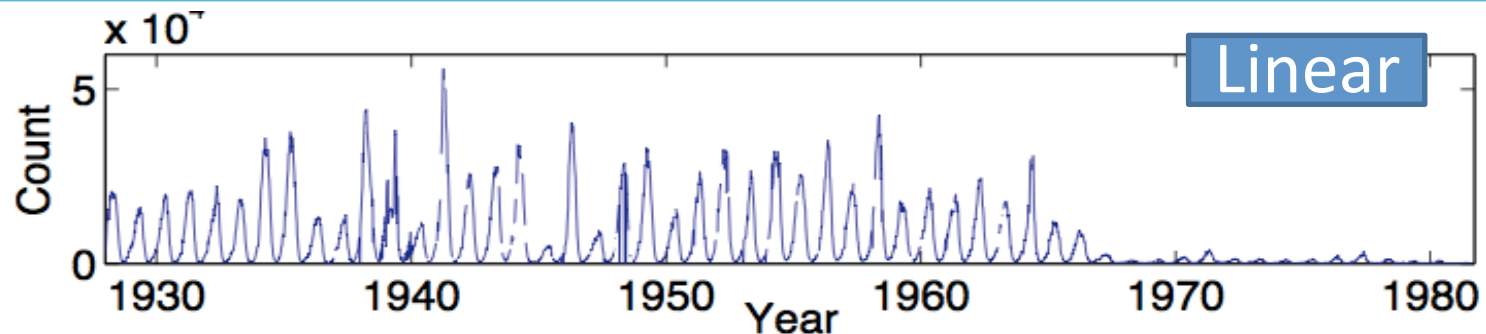
Given: Large set of epidemiological data
e.g., Measles cases in the U.S.



Motivation

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

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

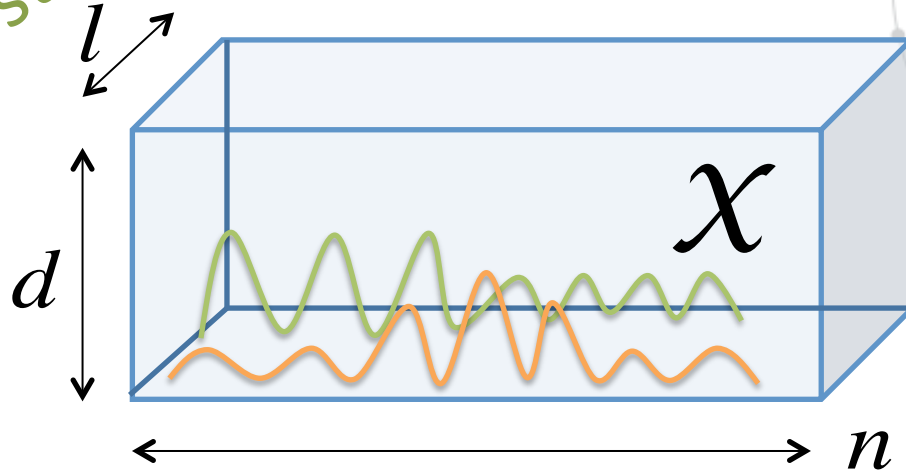


Data description

Project Tycho: infectious diseases in the U.S.

50 states

56 diseases



1888

Time (weekly) (> 125 years)



Data description

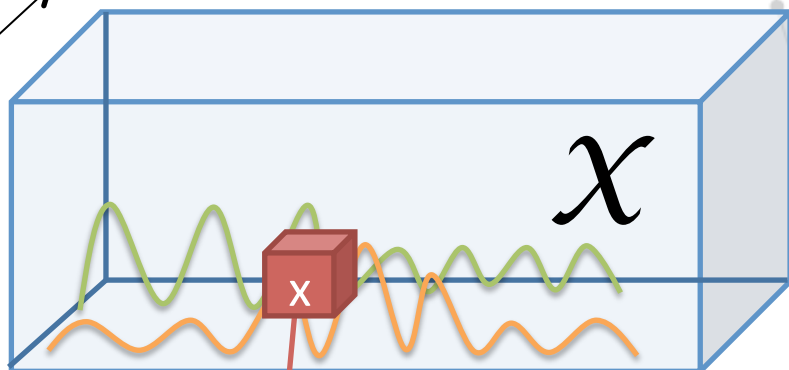
Project Tycho: infectious diseases in the U.S.



50 states

l

d



56 diseases

1888

Time (weekly) (> 125 years)

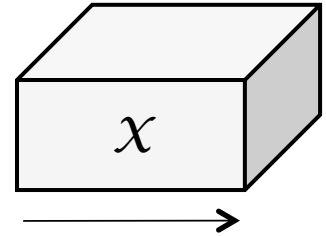
Element x : # of cases

e.g., 'measles', 'NY', 'April 1-7, 1931', '4000'

Problem definition

Given:

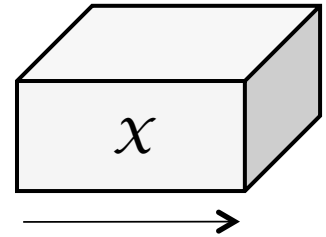
Tensor \mathcal{X} (disease x state x time)



Problem definition

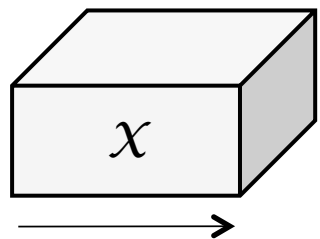
Given:

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

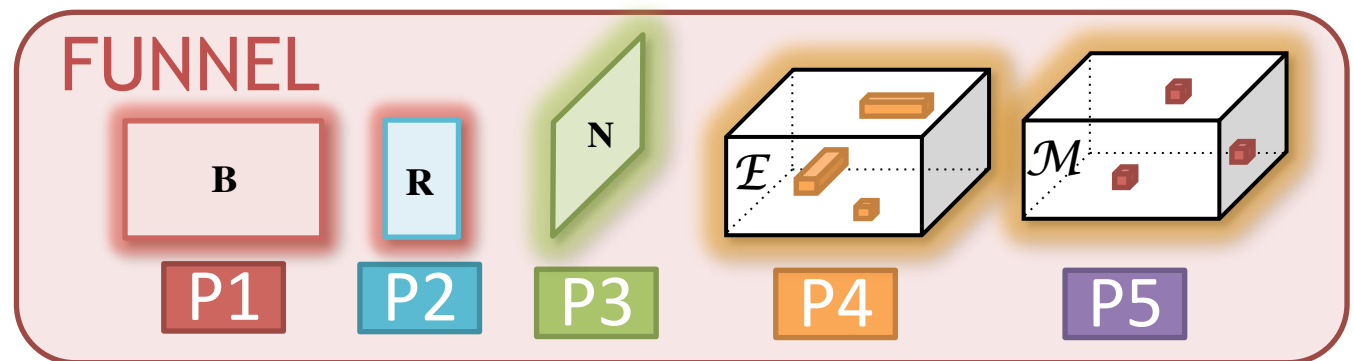


Find:

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



=



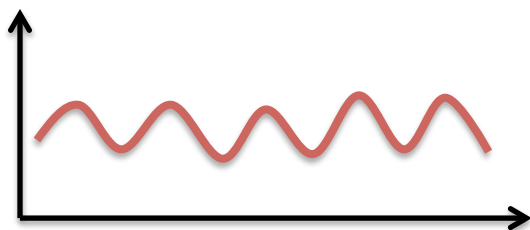
Problem definition

G

Te

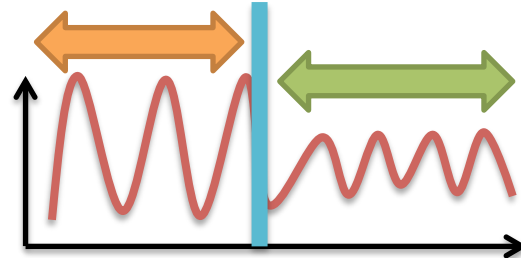
Fi

Seasonality

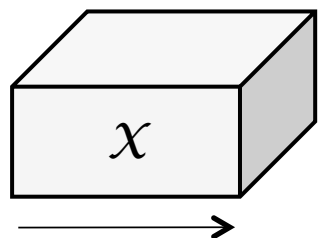


state

Discontinuities

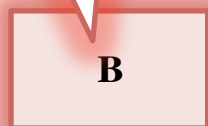


Compact description of \mathcal{X} , “*Automatically*”



=

FUNNEL



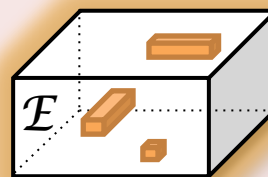
P1



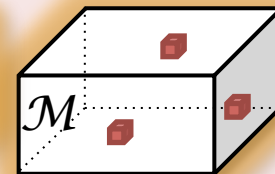
P2



P3



P4



P5

Problem definition

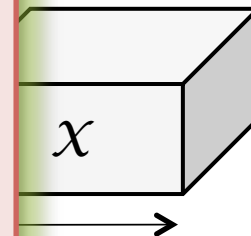
Given:
Tensor

NO magic numbers !

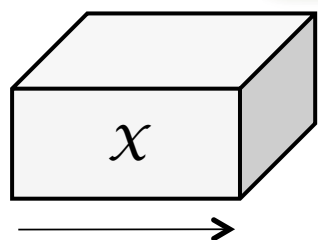


Find:
Compact

Parameter-free!

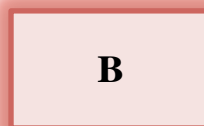


ically



=

FUNNEL



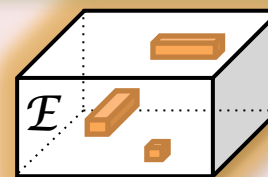
P1



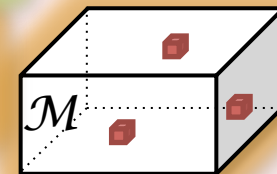
P2



P3



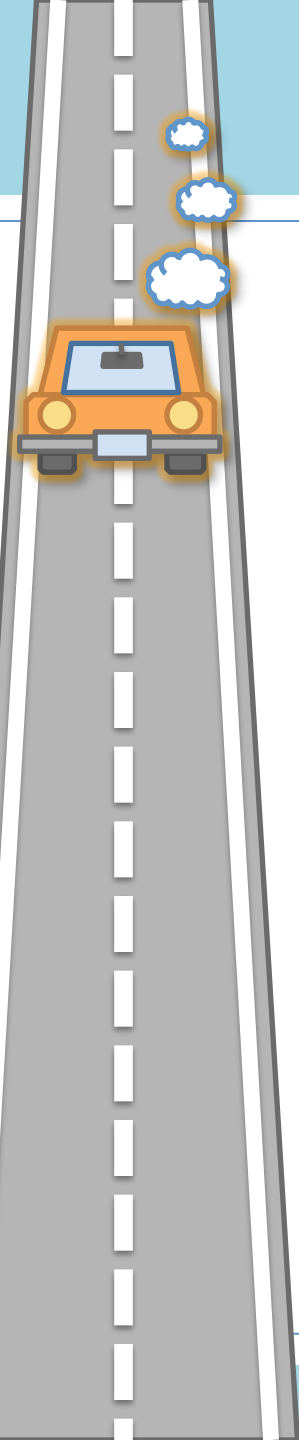
P4



P5

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

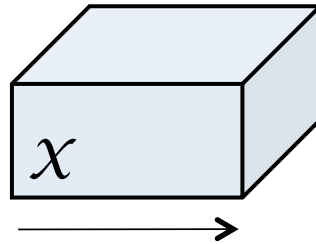
Q1

Q2

Q3

Q4

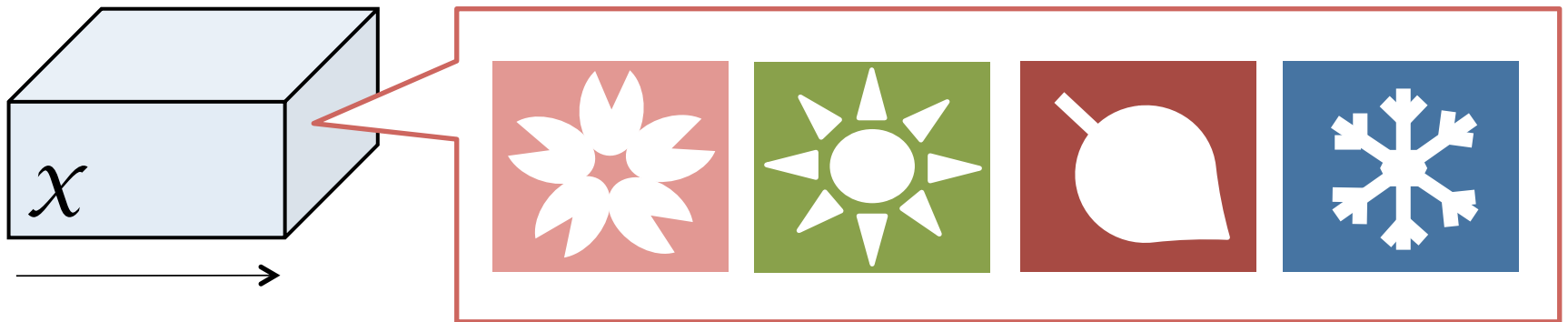
Q5



Q1

Are there any periodicities?

If yes, when is the peak season?



Answers

Q1

Q2

Q3

Q4

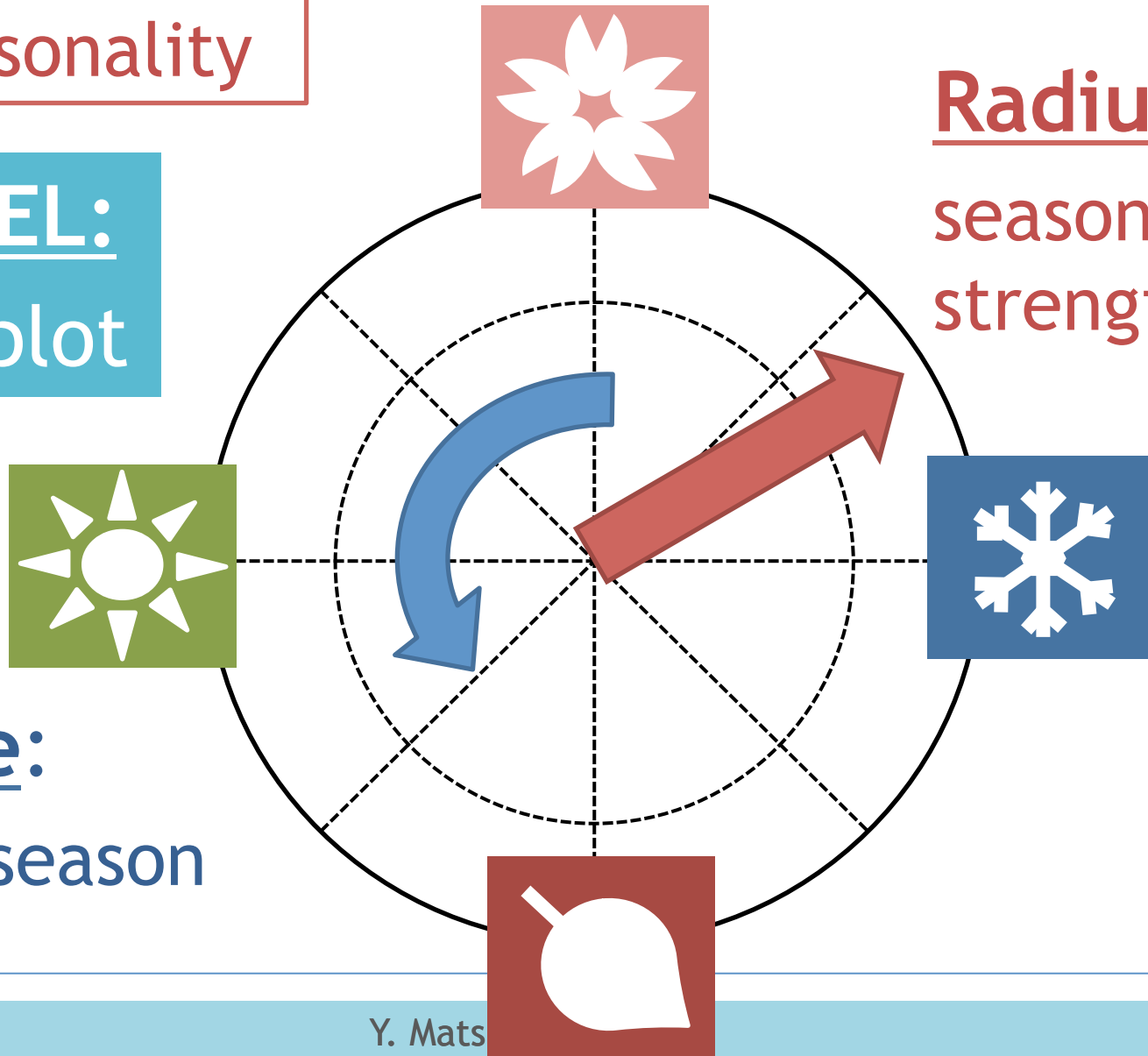
Q5

P1 Seasonality

FUNNEL:
Polar plot

Radius:
seasonality
strength

Angle:
peak season



Answers

Q1

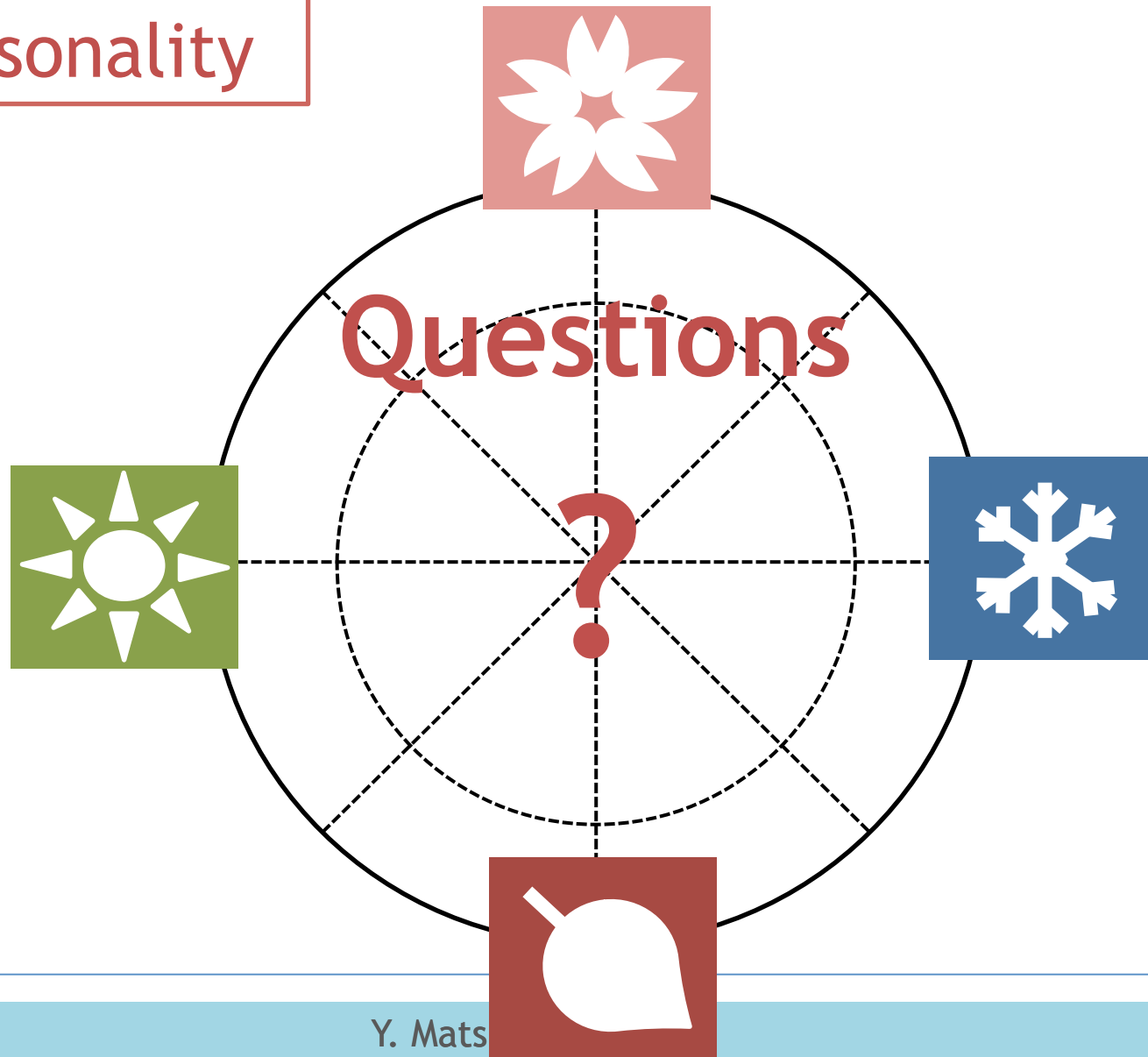
Q2

Q3

Q4

Q5

P1 Seasonality



Answers

Q1

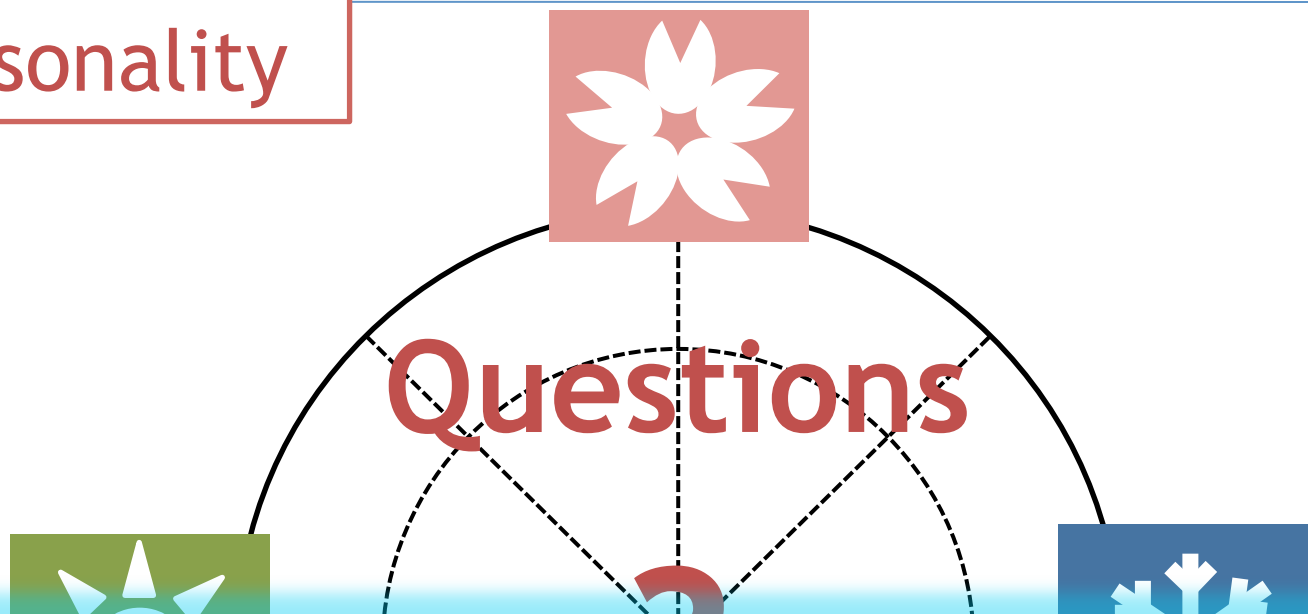
Q2

Q3

Q4

Q5

P1 Seasonality



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

Answers

Q1

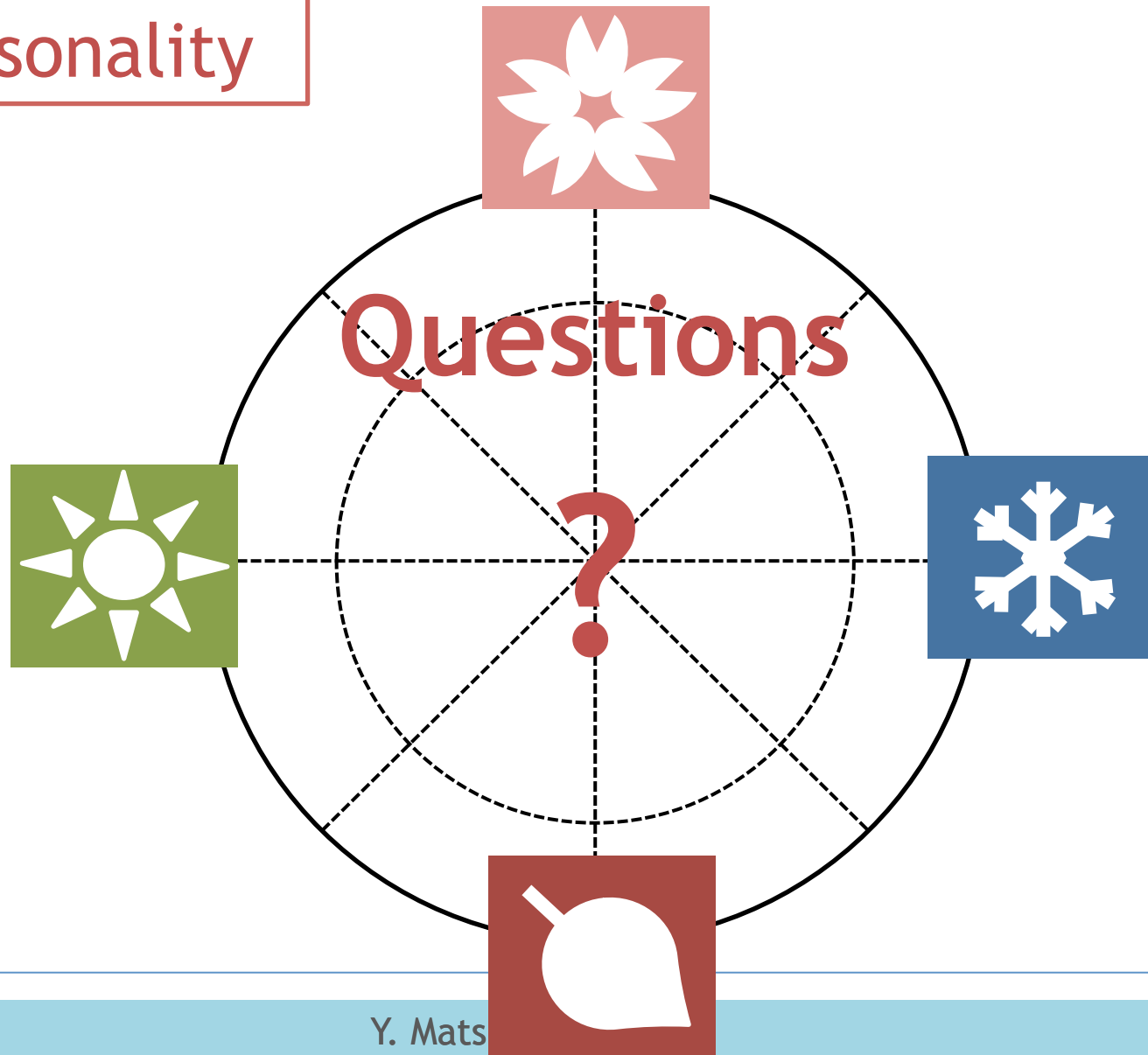
Q2

Q3

Q4

Q5

P1 Seasonality



Answers

Q1

Q2

Q3

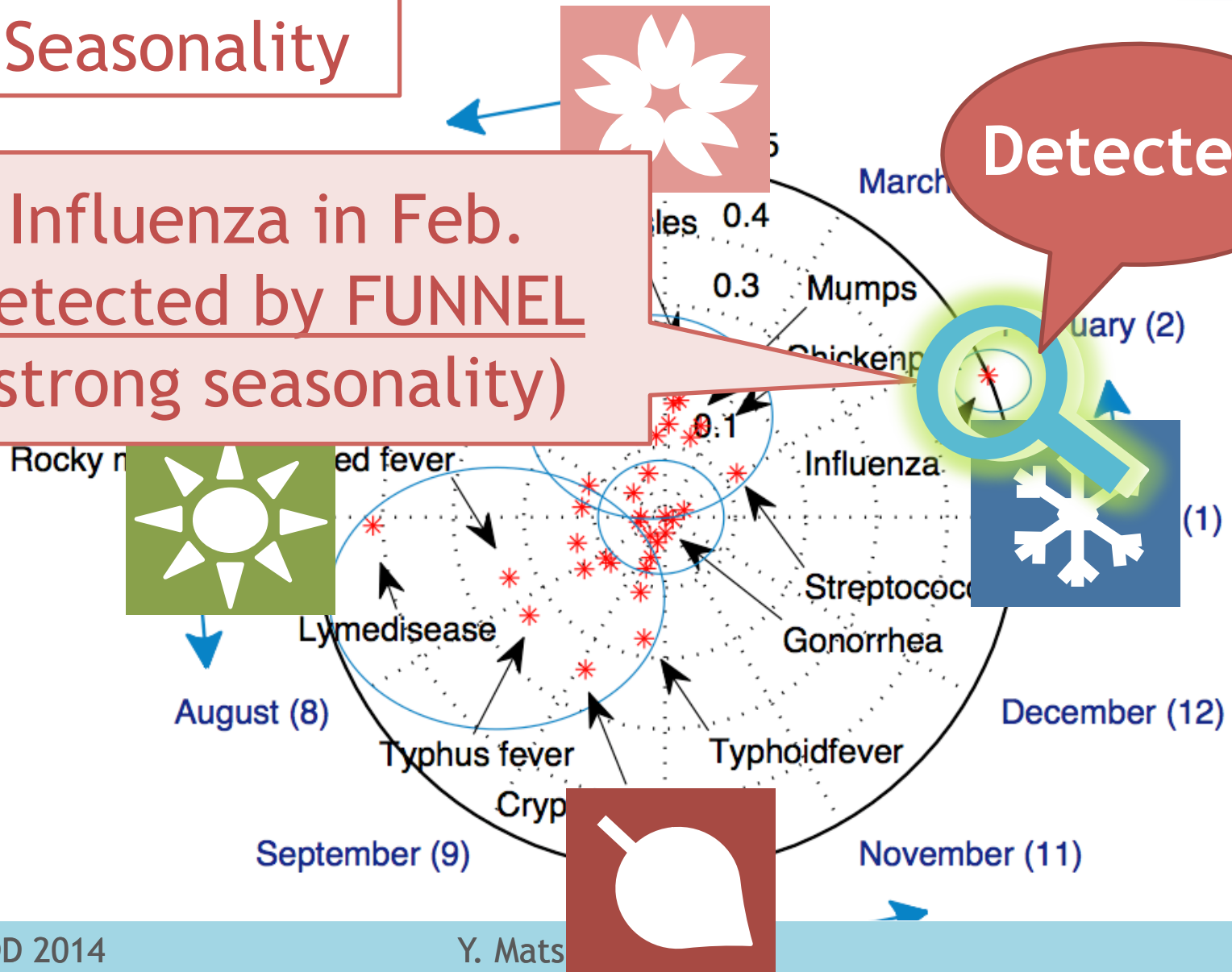
Q4

Q5

P1 Seasonality

Influenza in Feb.
Detected by FUNNEL
(strong seasonality)

Detected!



Answers

Q1

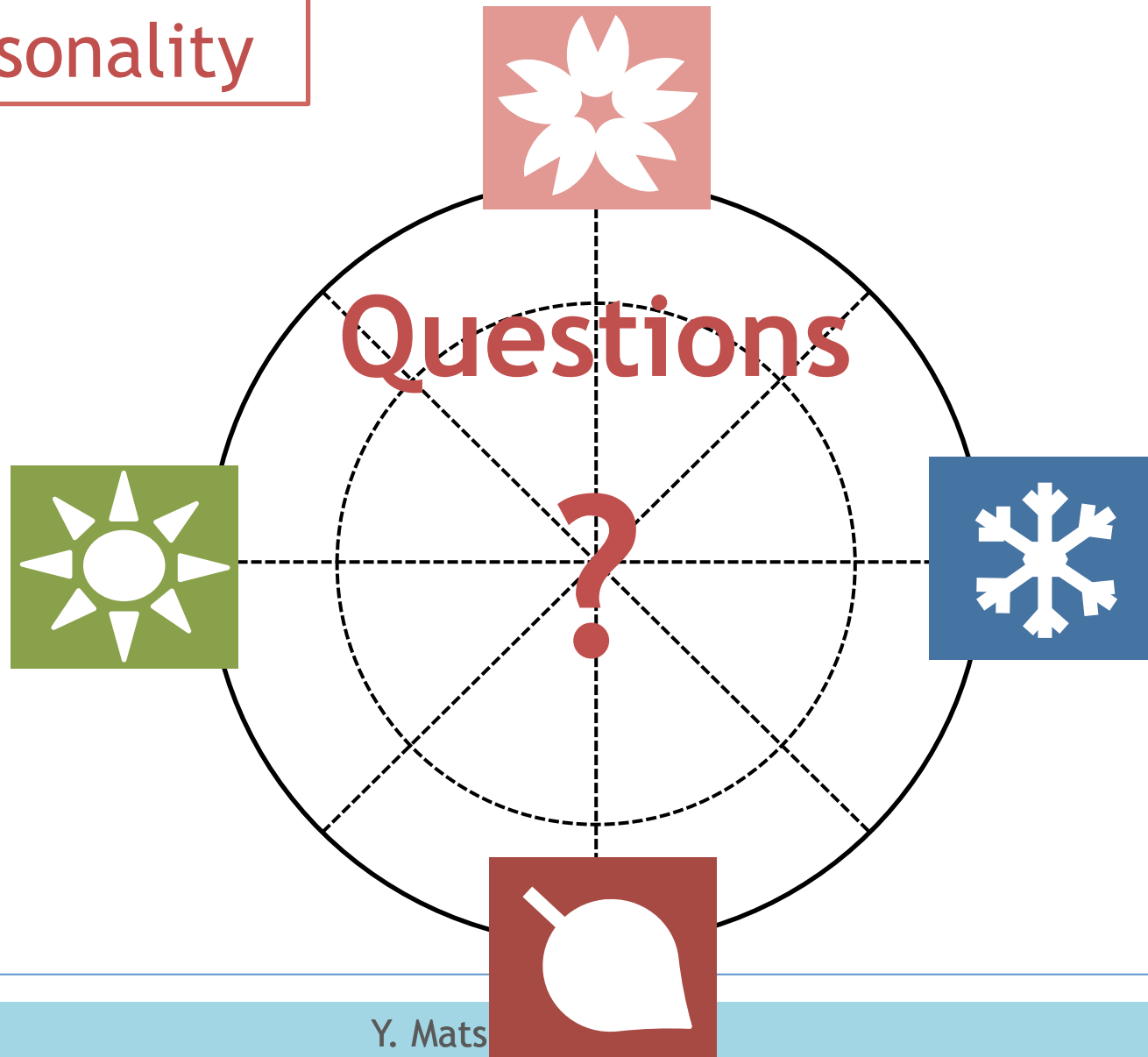
Q2

Q3

Q4

Q5

P1 Seasonality



Answers

Q1

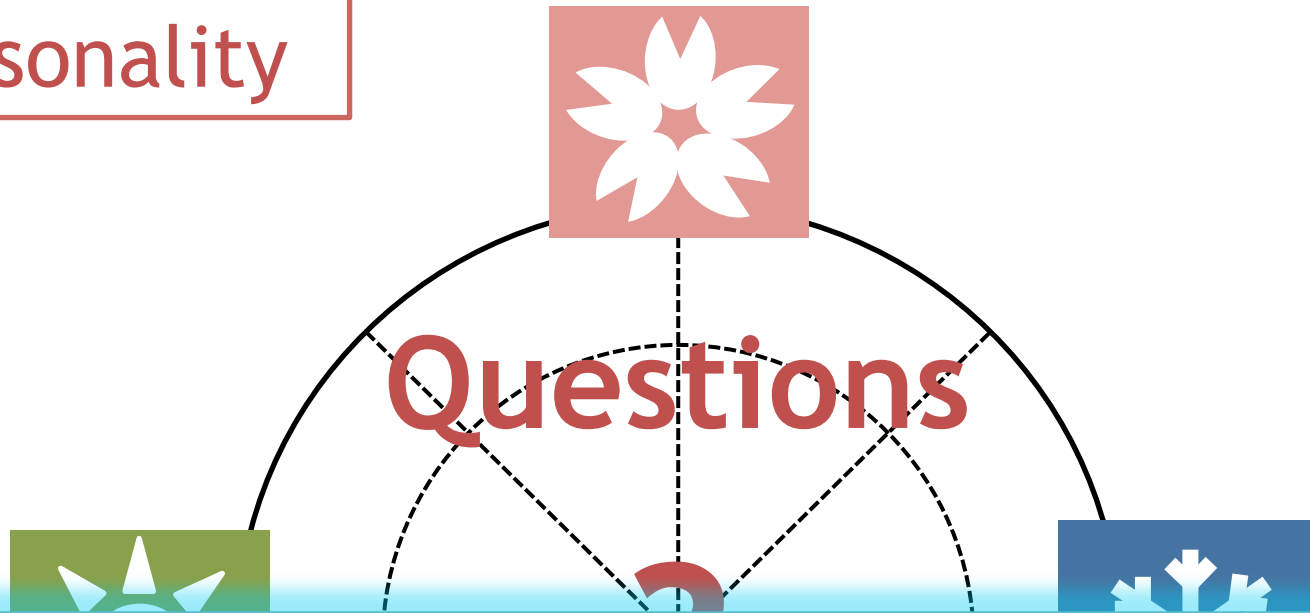
Q2

Q3

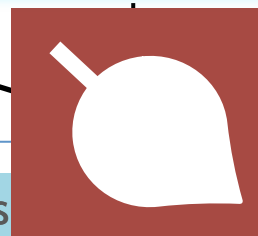
Q4

Q5

P1 Seasonality



Q: How about measles ?



Answers

Q1

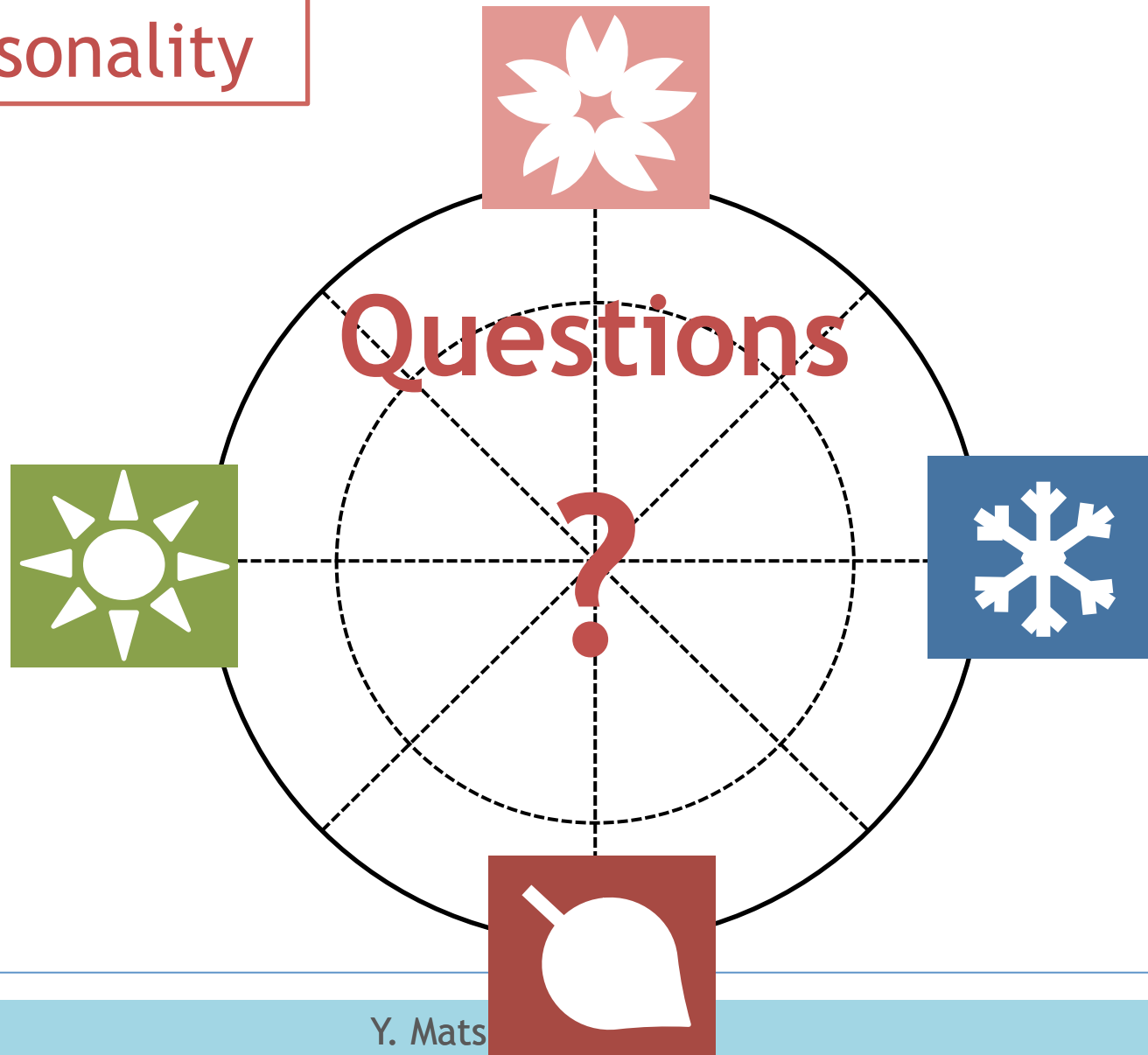
Q2

Q3

Q4

Q5

P1 Seasonality



Answers

Q1

Q2

Q3

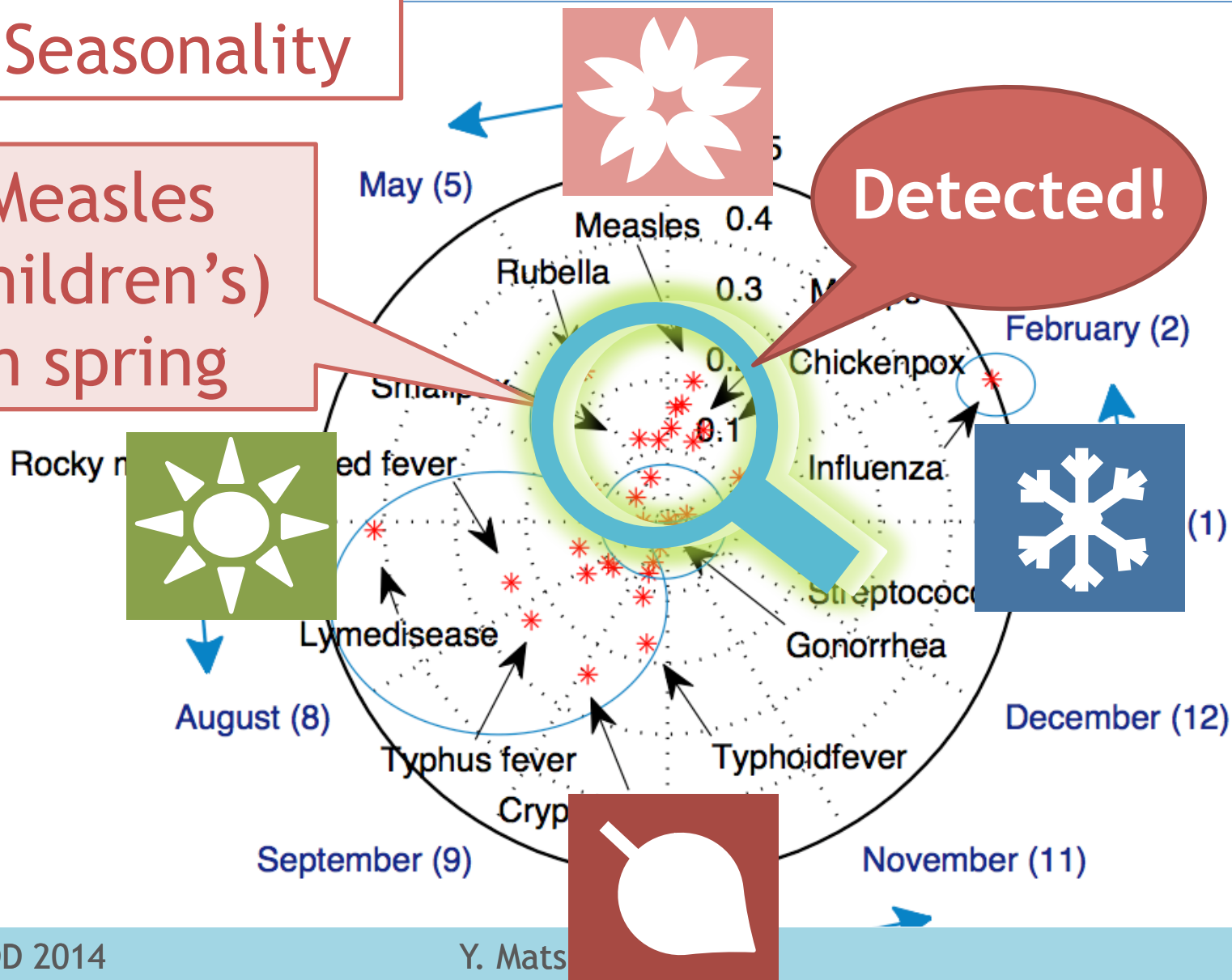
Q4

Q5

P1 Seasonality

Measles
(children's)
in spring

Detected!



Answers

Q1

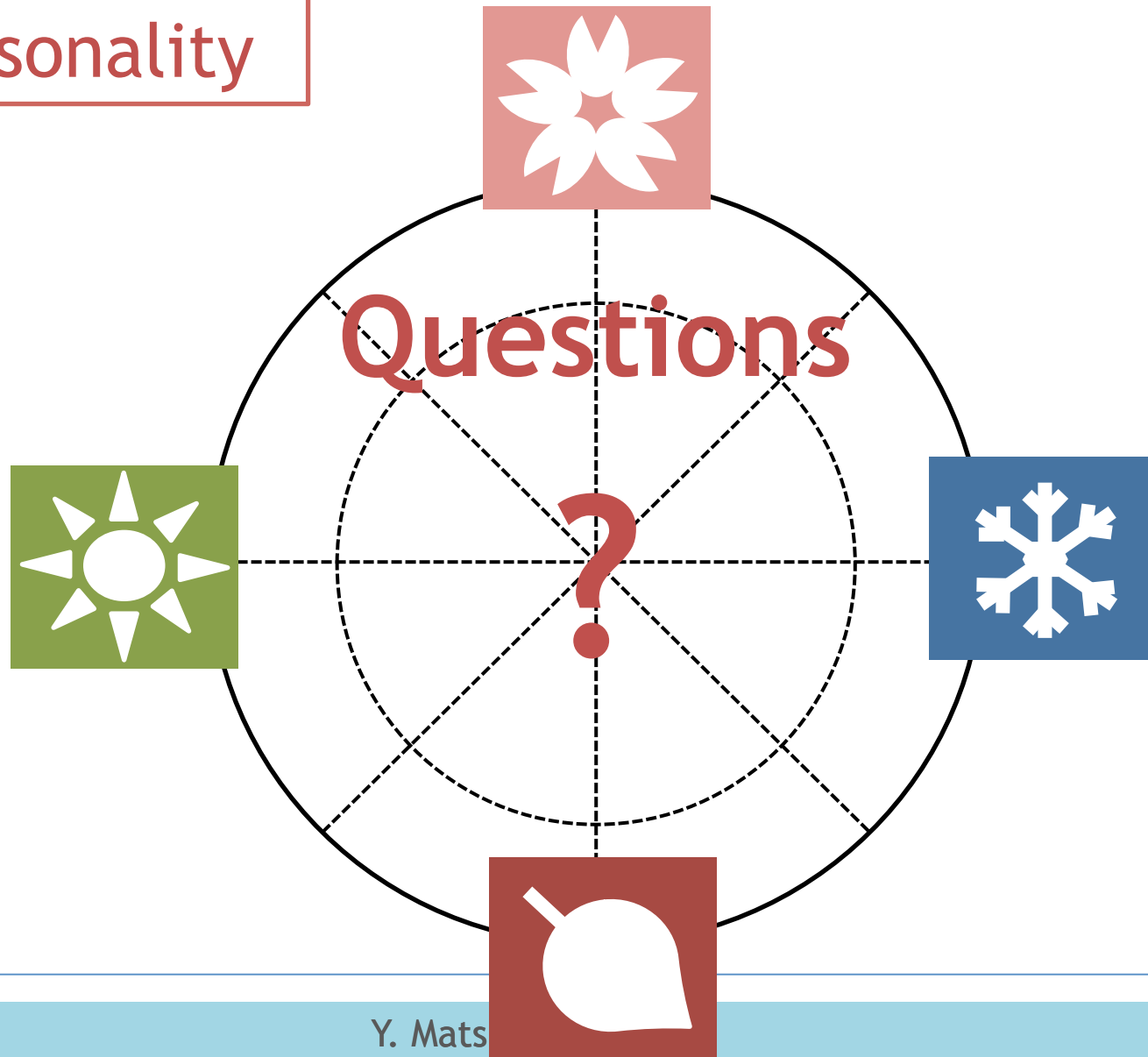
Q2

Q3

Q4

Q5

P1 Seasonality



Answers

Q1

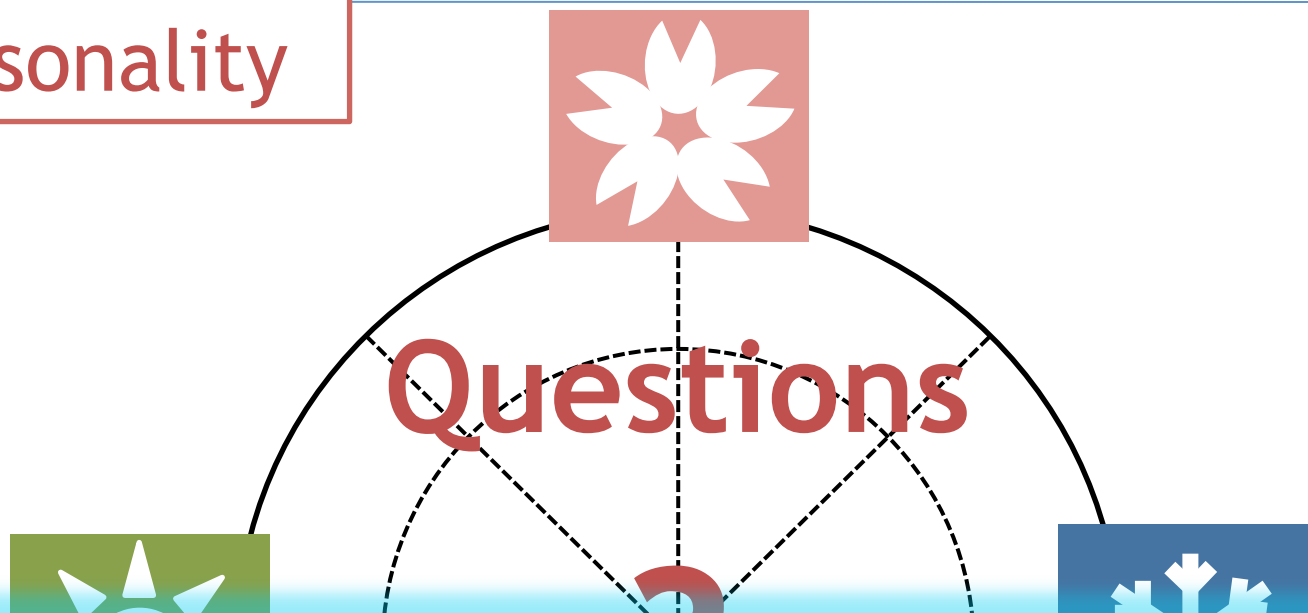
Q2

Q3

Q4

Q5

P1 Seasonality



Q: Which disease peaks in summer?

Answers

Q1

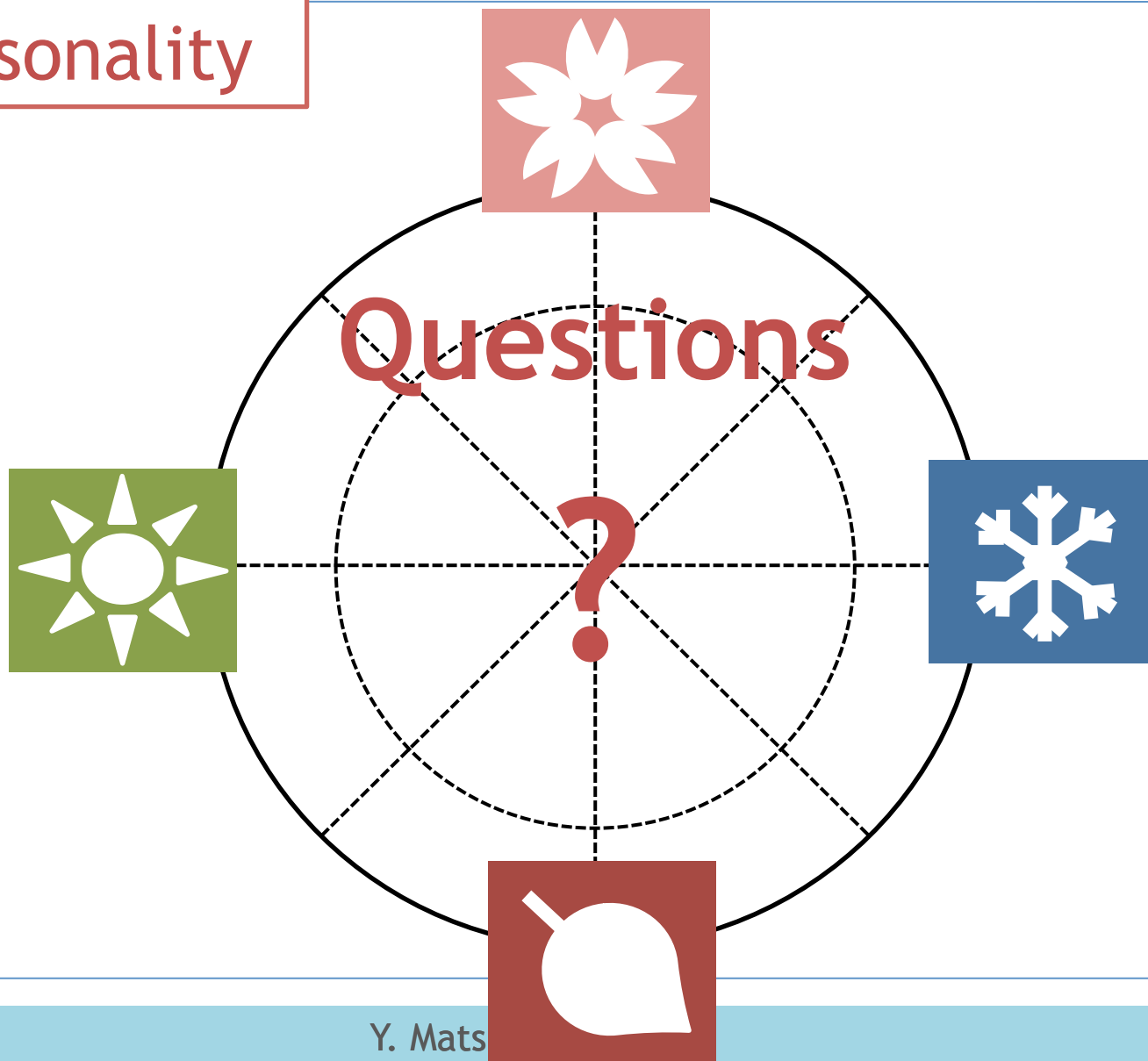
Q2

Q3

Q4

Q5

P1 Seasonality



Answers

Q1

Q2

Q3

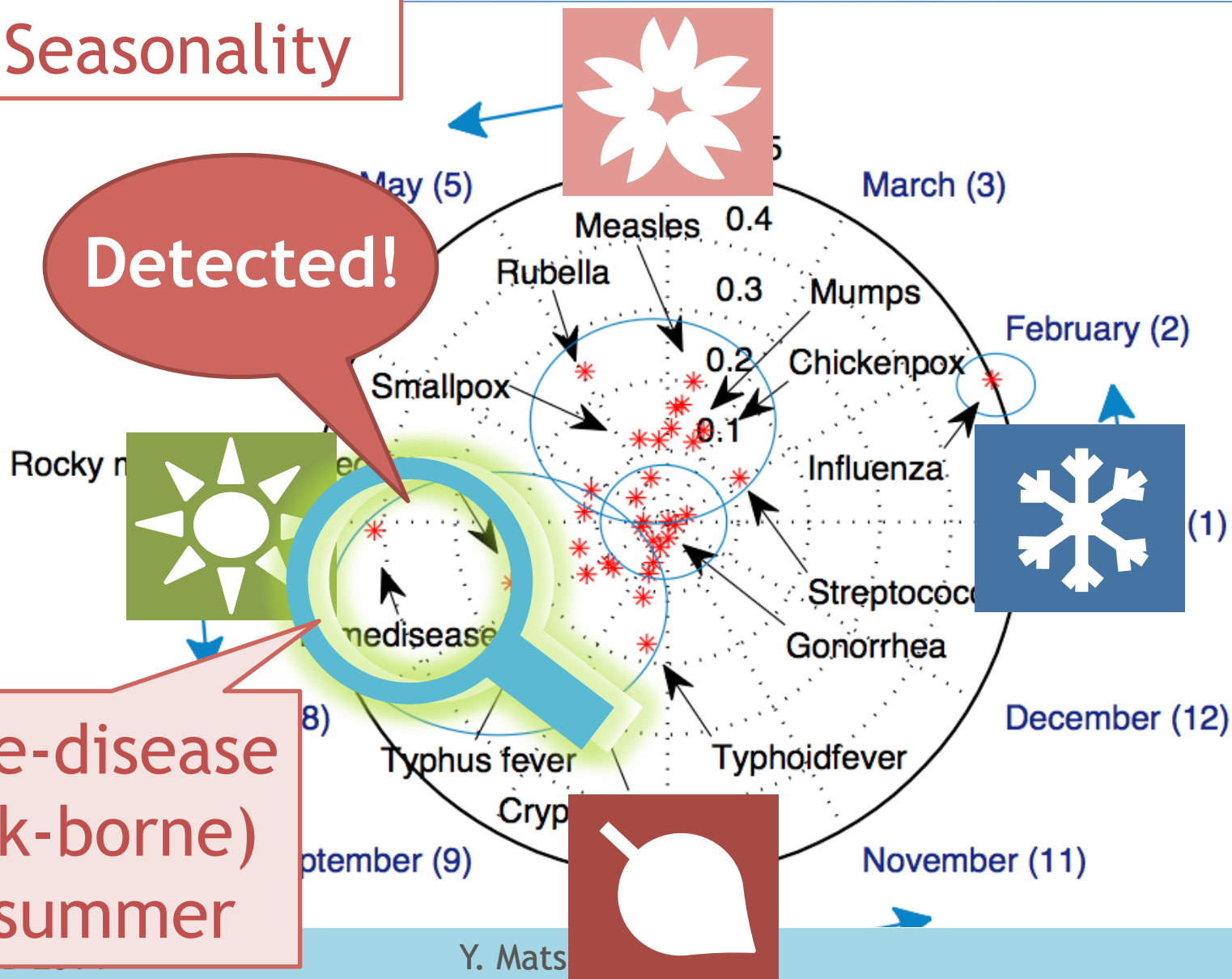
Q4

Q5

P1 Seasonality

Detected!

Lyme-disease
(tick-borne)
in summer



Answers

Q1

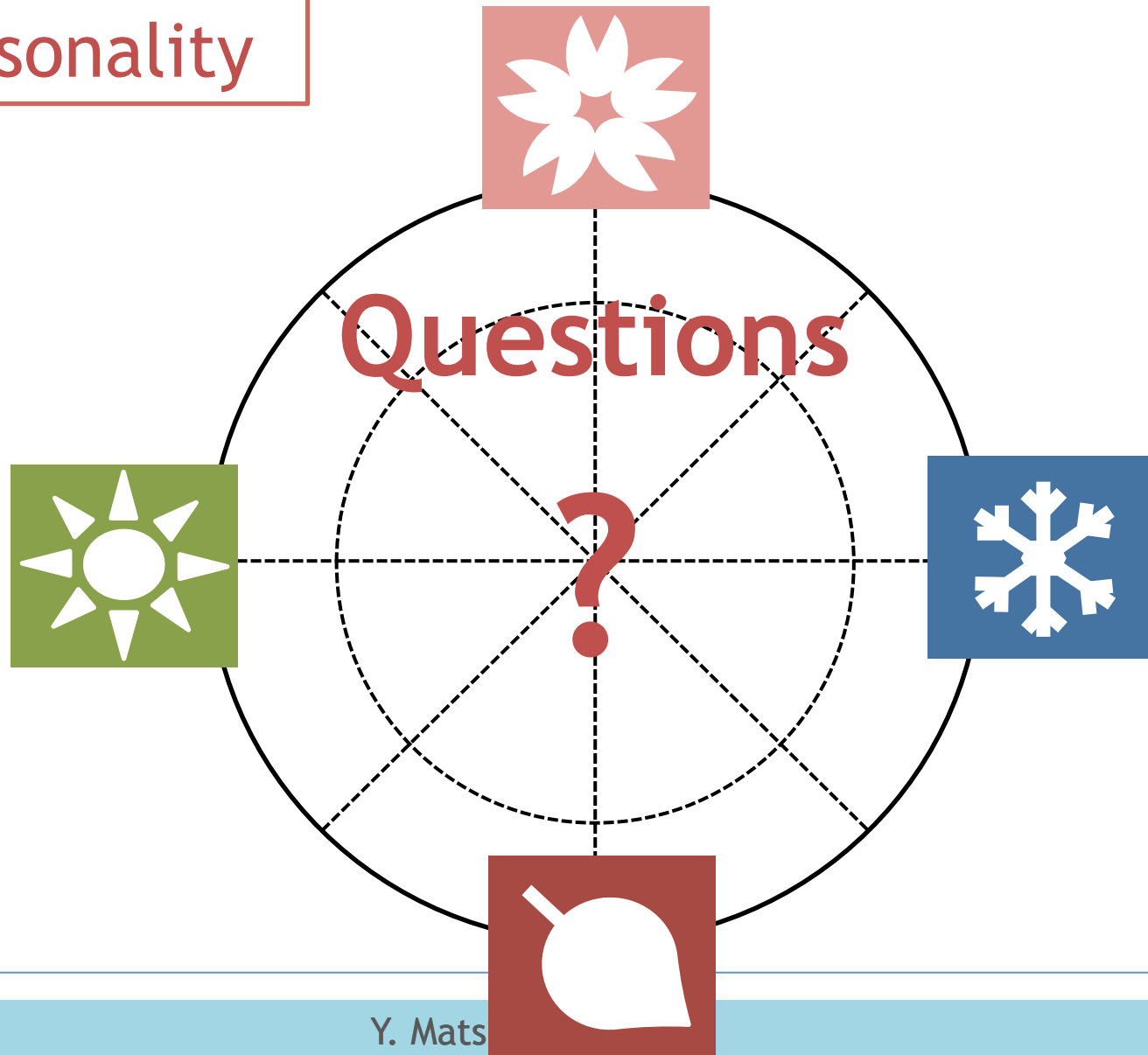
Q2

Q3

Q4

Q5

P1 Seasonality



Answers

Q1

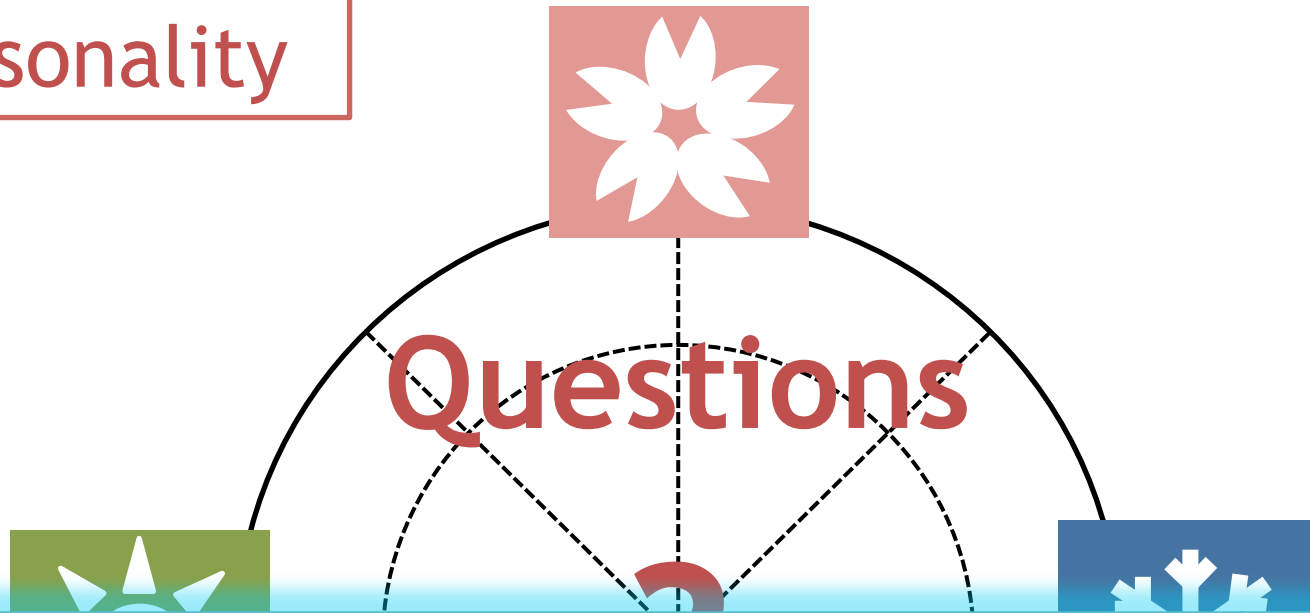
Q2

Q3

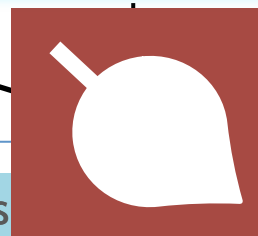
Q4

Q5

P1 Seasonality



Q: Which disease has no periodicity?



Answers

Q1

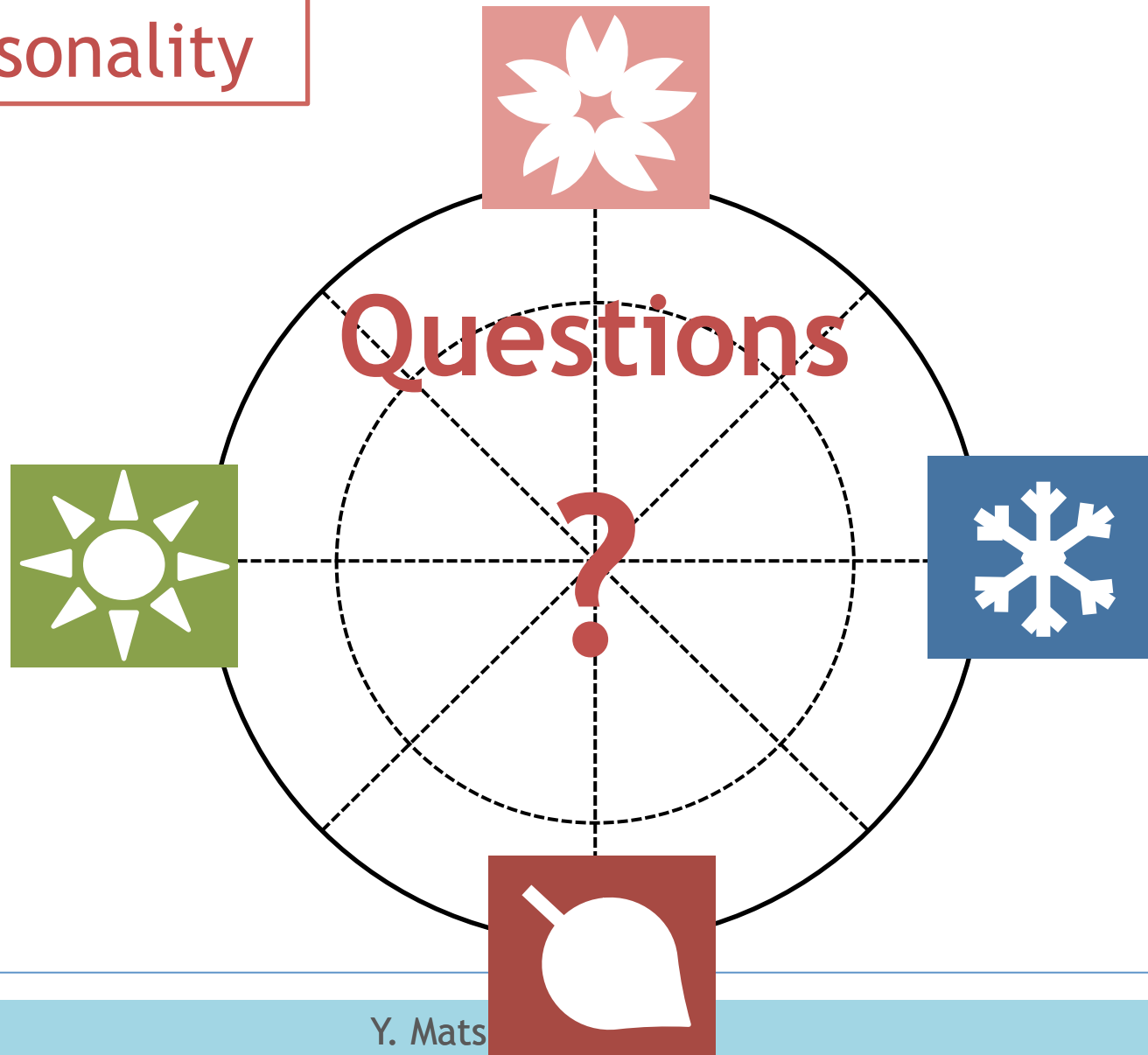
Q2

Q3

Q4

Q5

P1 Seasonality



Answers

Q1

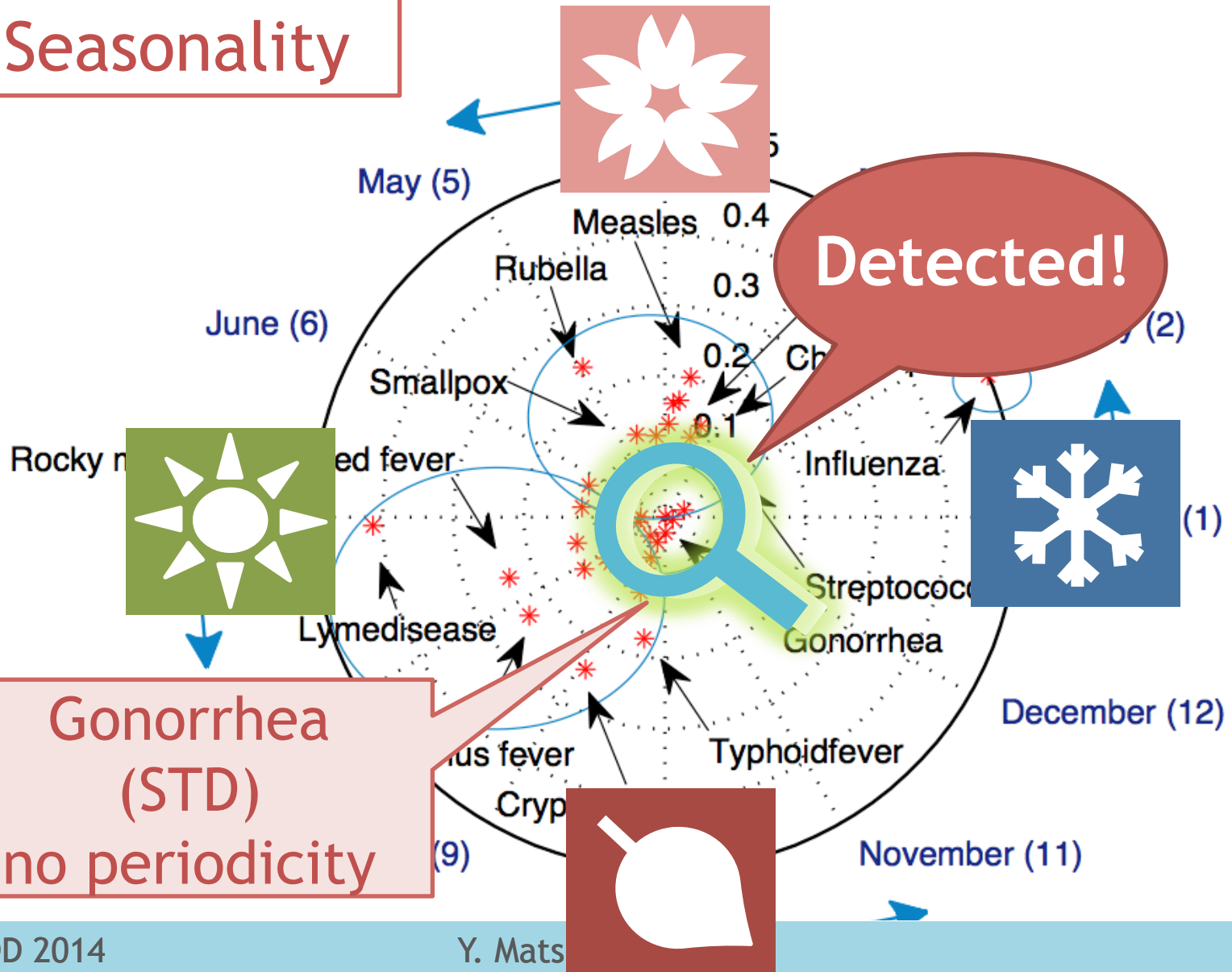
Q2

Q3

Q4

Q5

P1 Seasonality



Questions

Q1

Q2

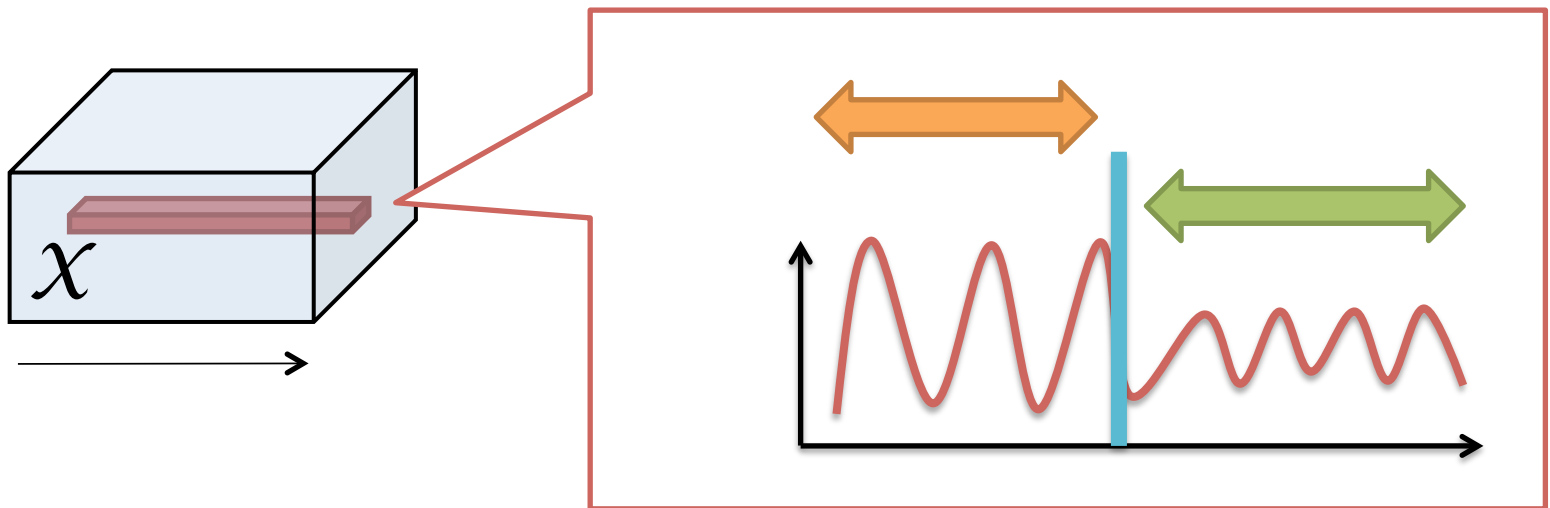
Q3

Q4

Q5

Q2

Can we see any discontinuities?



Answers

Q1

Q2

Q3

Q4

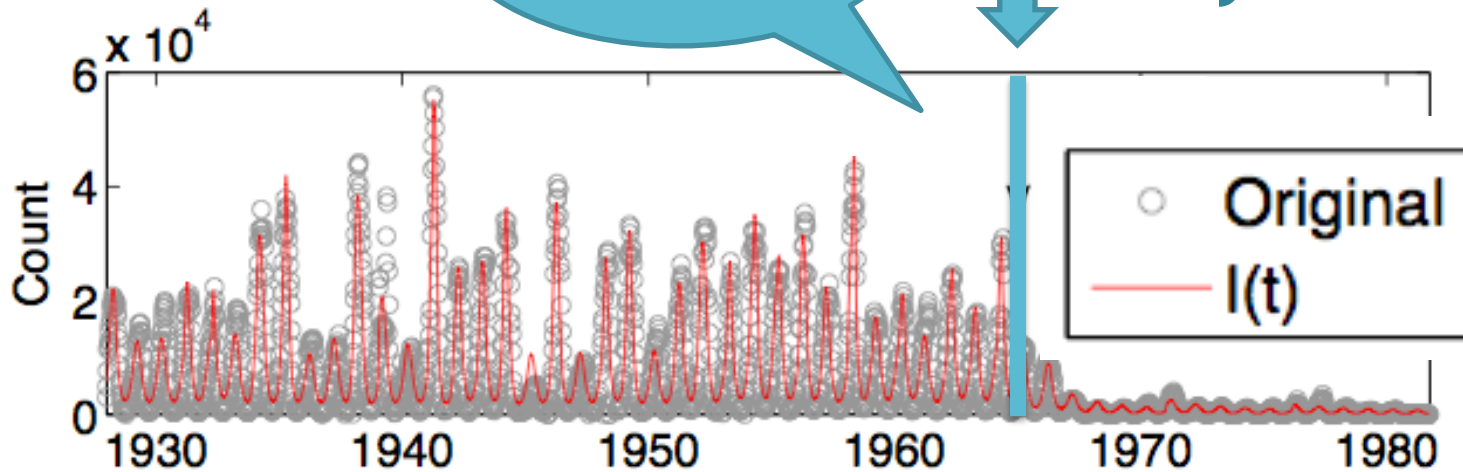
Q5

P2 Disease reduction effect

Measles

Detected!

1965: Detected by FUNNEL



1963:
Vaccine licensure

Questions

Q1

Q2

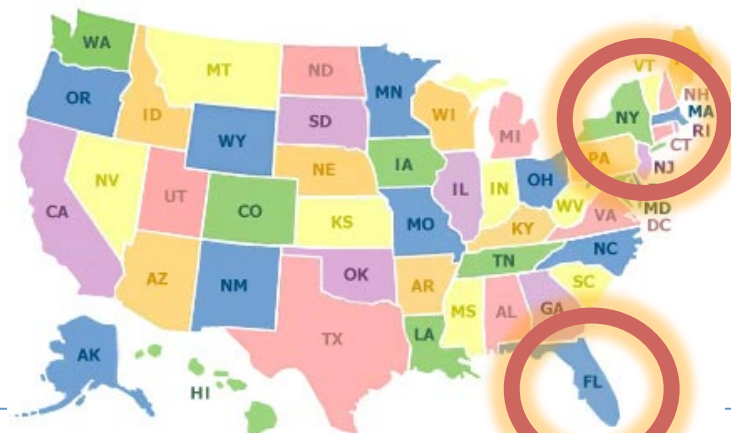
Q3

Q4

Q5

Q3

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



Answers

Q1

Q2

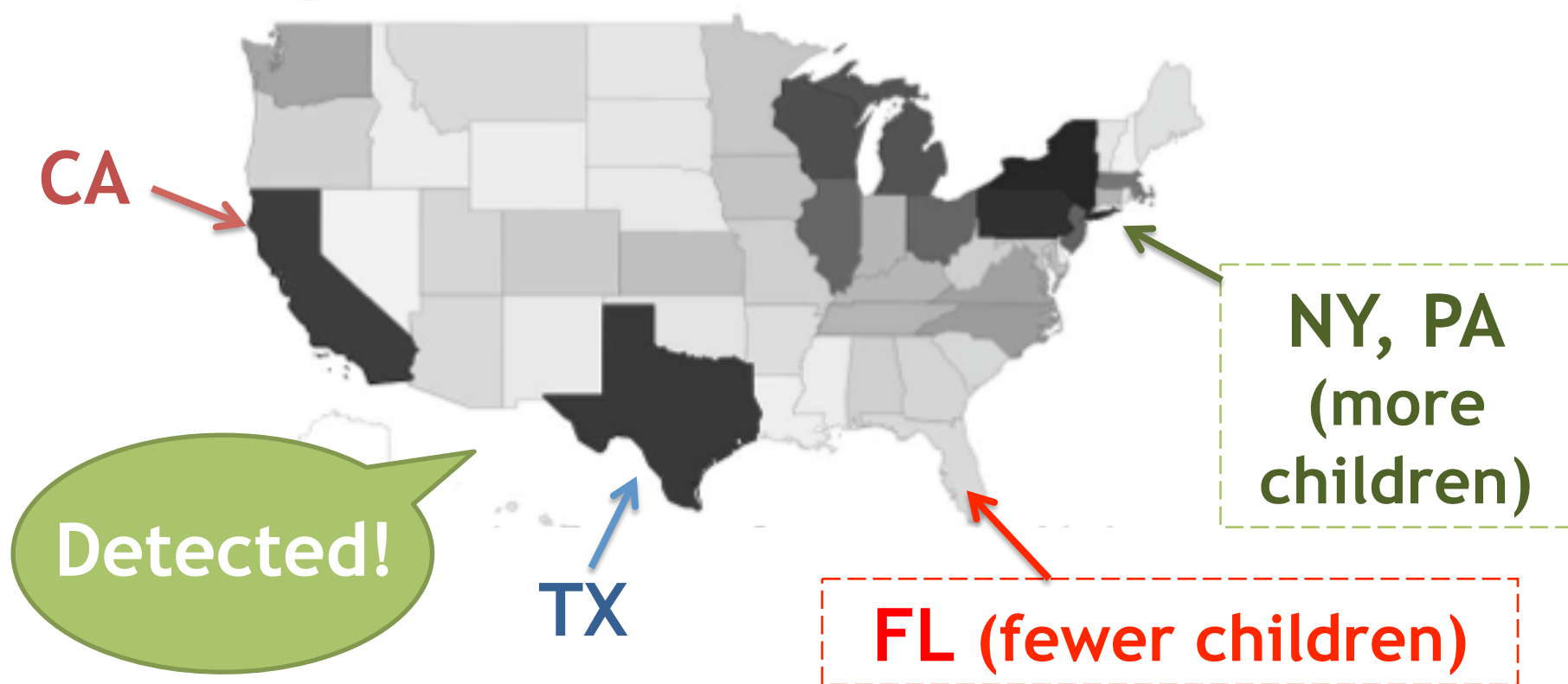
Q3

Q4

Q5

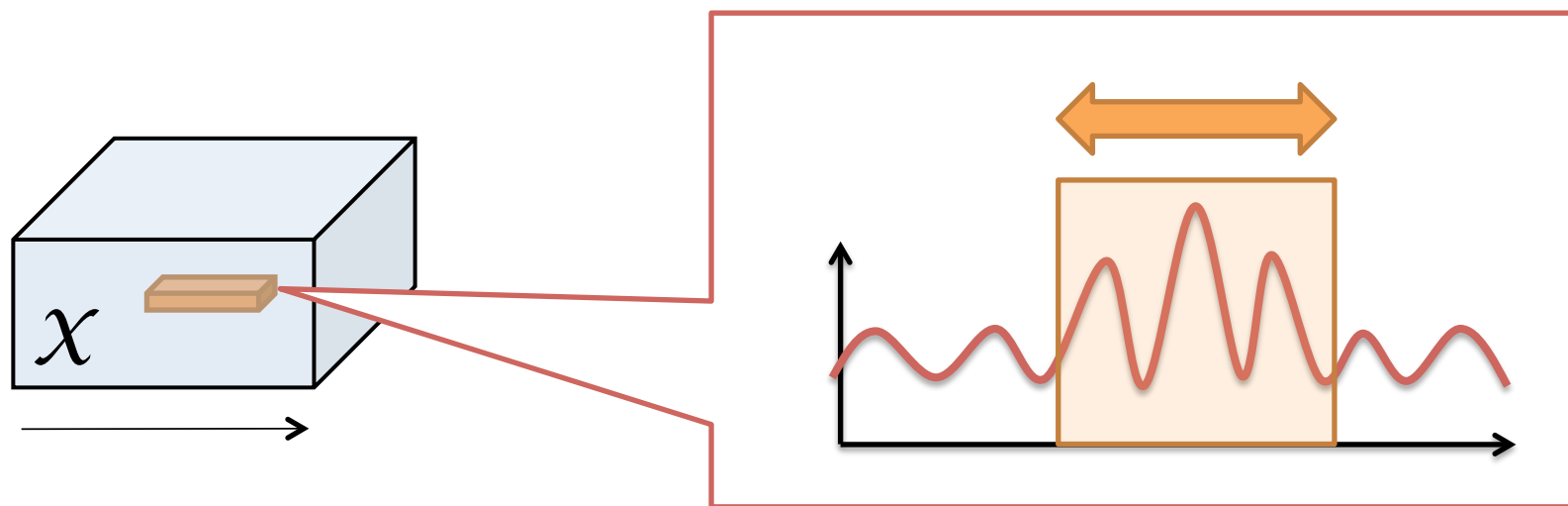
P3 area sensitivity

FUNNEL's guess of susceptibles (measles)



Q4

Are there any external shock events, like wars?



Answers

Q1

Q2

Q3

Q4

Q5

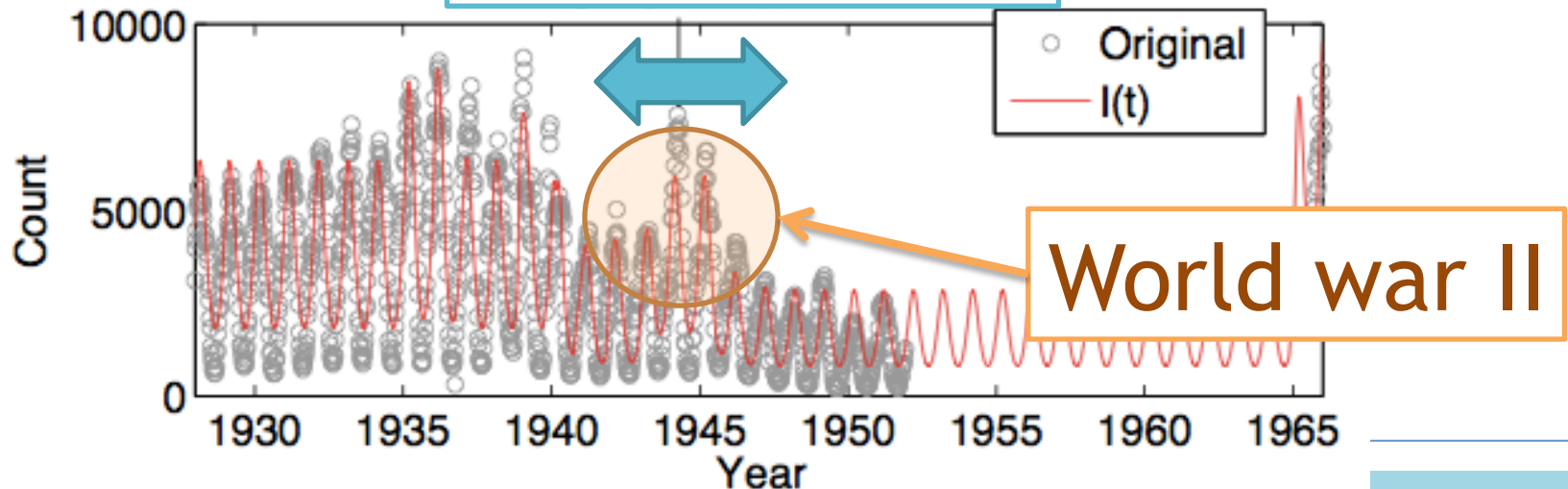
P4 external shock events

Funnel can detect external shocks
“**fully-automatically**” !

Scarlet fever

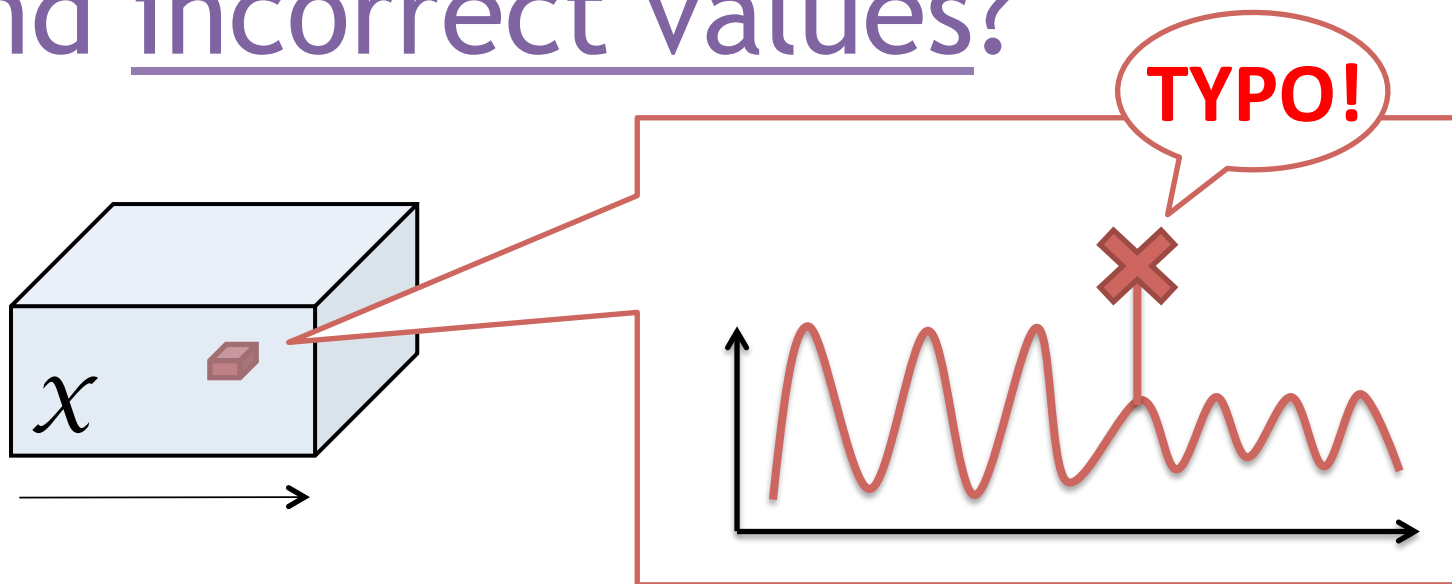
Detected by
FUNNEL

Detected!



Q5

How can we remove mistakes and incorrect values?



Answers

Q1

Q2

Q3

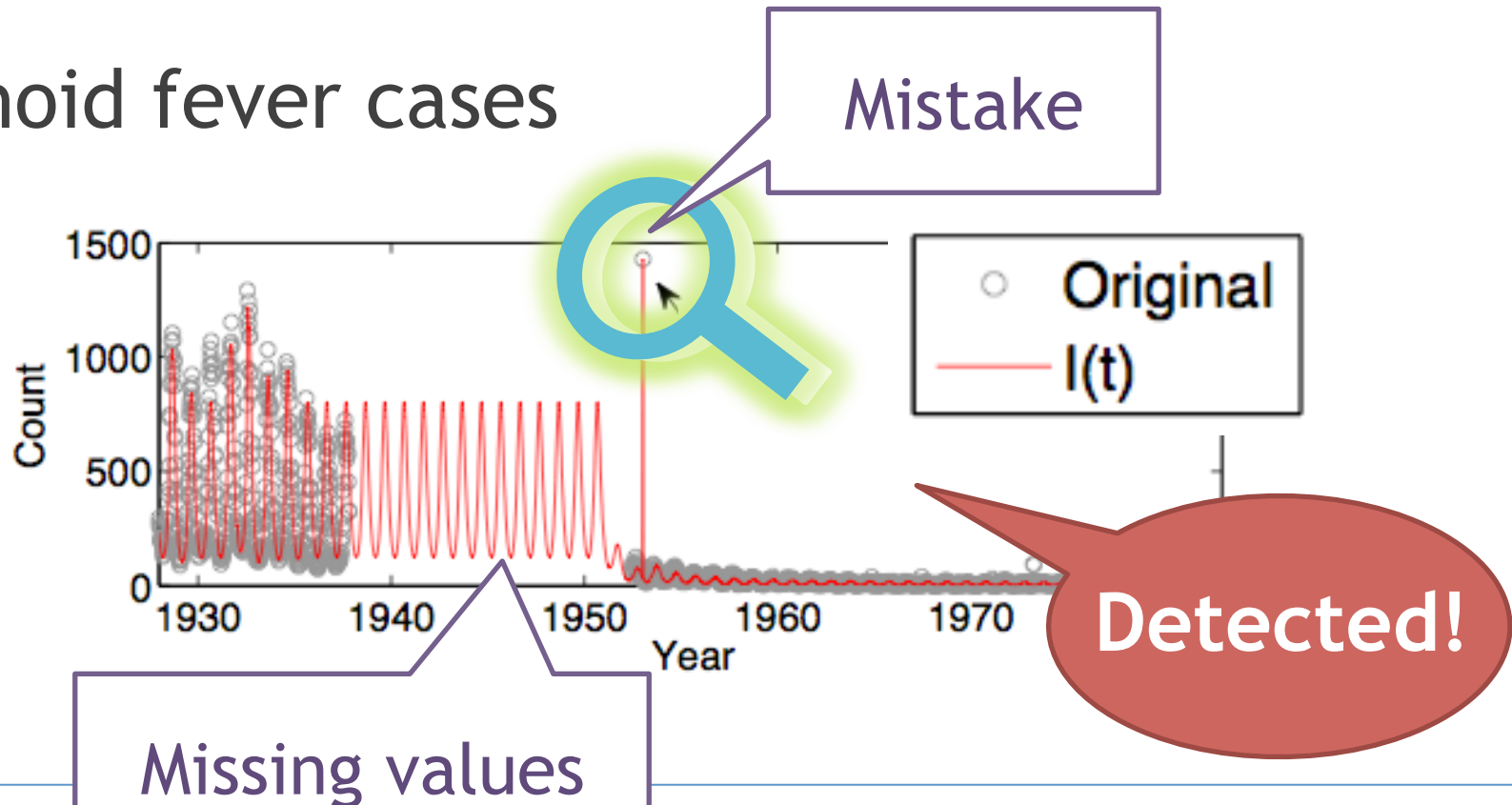
Q4

Q5

P5 mistakes

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

Typhoid fever cases



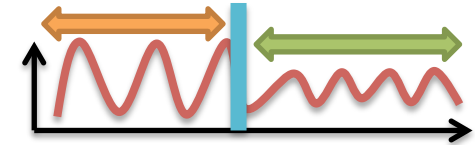
Modeling power of FUNNEL

Our model can capture 5 properties

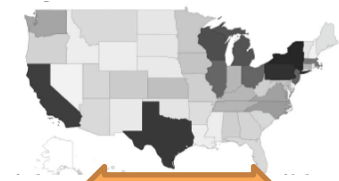
P1 Seasonality



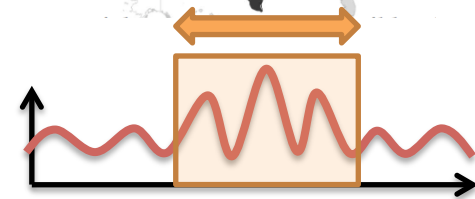
P2 Disease reductions



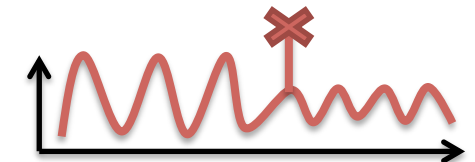
P3 Area sensitivity



P4 External events

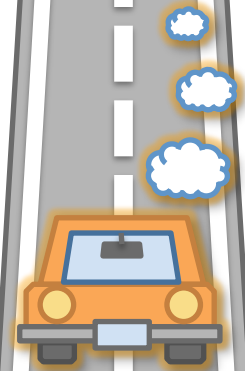


P5 Mistakes



Roadmap

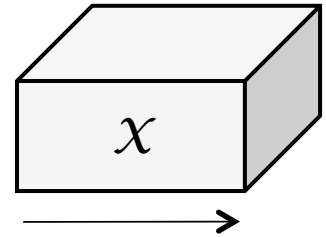
- ✓ Motivation
- ✓ Modeling power of FUNNEL
 - Overview - main ideas
 - Proposed model - idea #1
 - Algorithm - idea #2
 - Experiments
 - Discussion
 - Conclusions



Problem definition

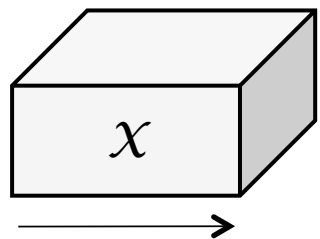
Given:

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

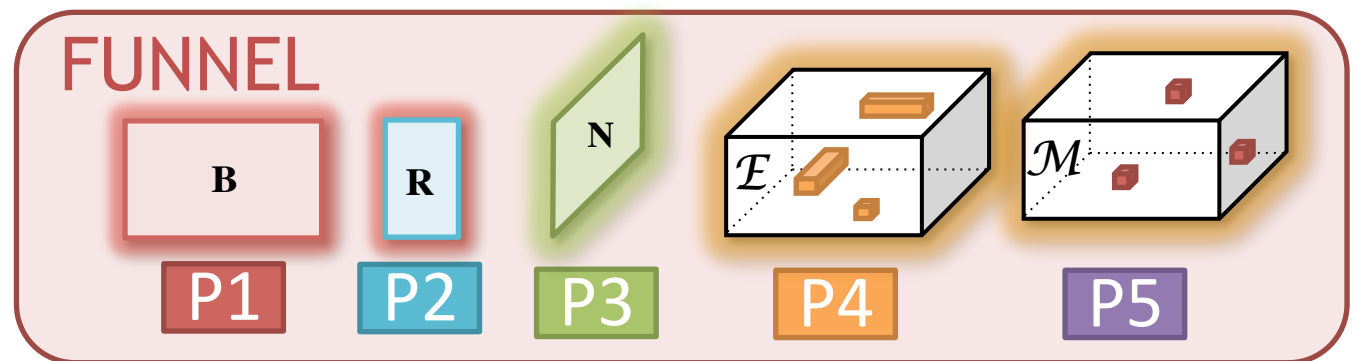


Find:

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

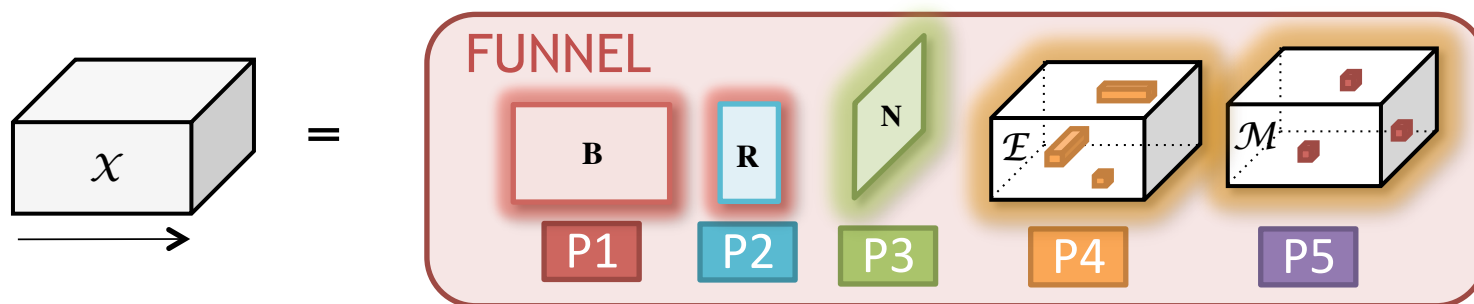


=



Two main ideas

Idea #1: Grey-box model



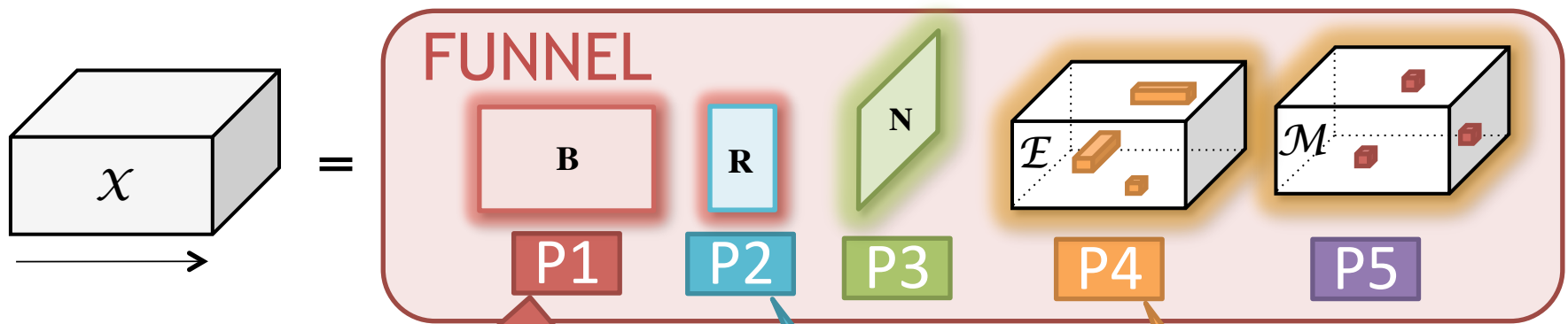
Idea #2: MDL for fitting

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



Two main ideas

Idea #1: Grey-box model - domain knowledge



(SIRS+) : 6 parameters

$$S(t+1) = S(t) - \beta(t)\epsilon(t)S(t)I(t)$$

$$I(t+1) = I(t) + \beta(t)\epsilon(t)S(t)I(t)$$

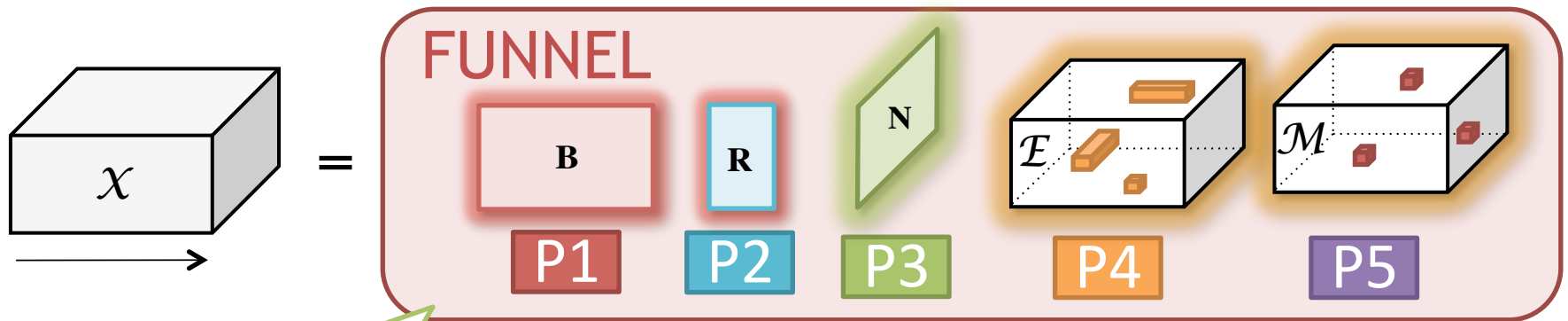
$$V(t+1) = V(t) + \delta I(t) - \gamma V(t) +$$

Vaccine

Shocks

Two main ideas

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



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

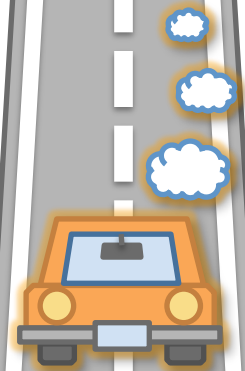
NO magic numbers



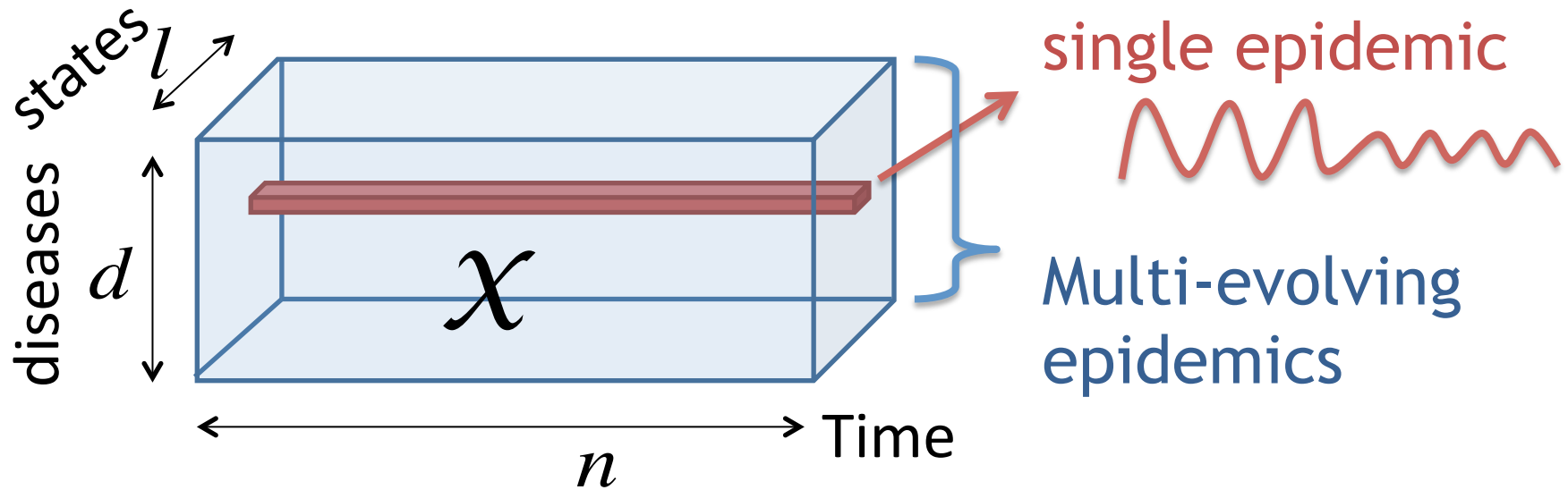
Parameter-free!

Roadmap

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

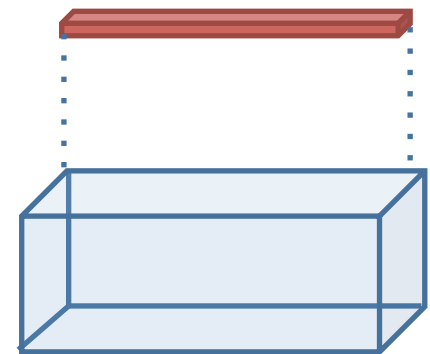


Proposed model: FUNNEL

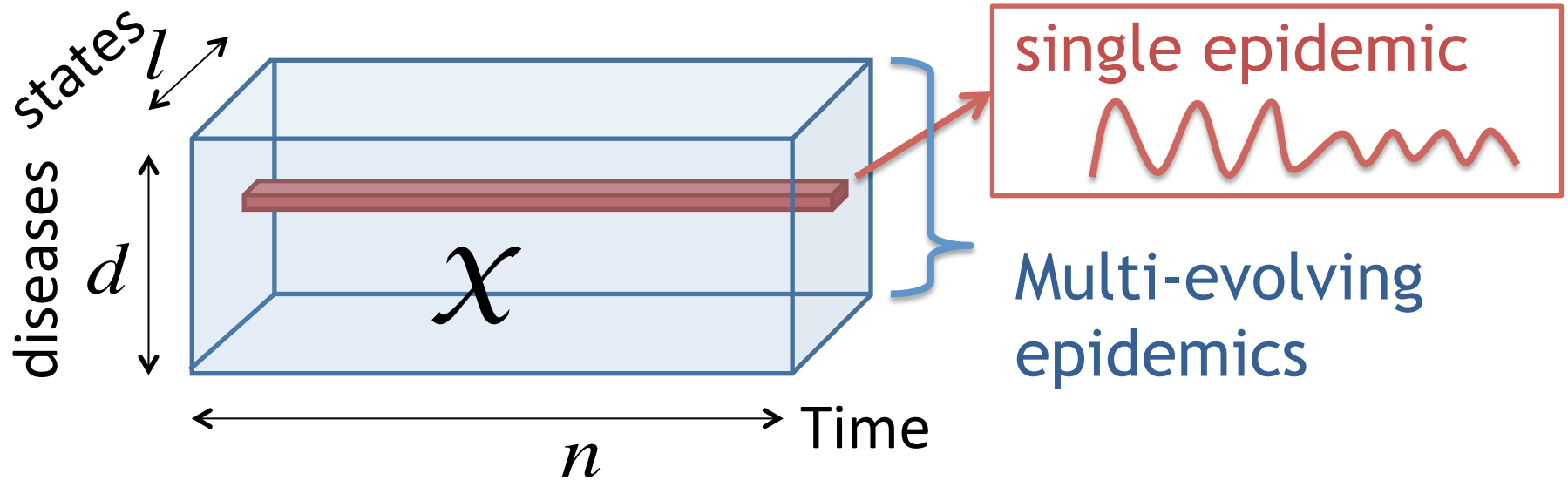


(a) FUNNEL-single

(b) FUNNEL-full

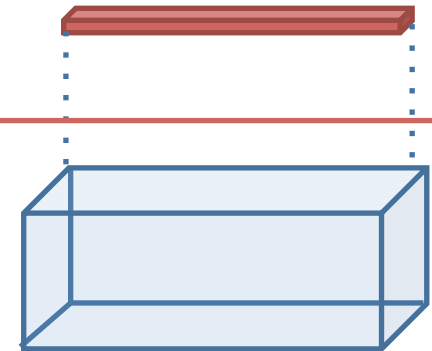


Proposed model: FUNNEL



(a) FUNNEL-single

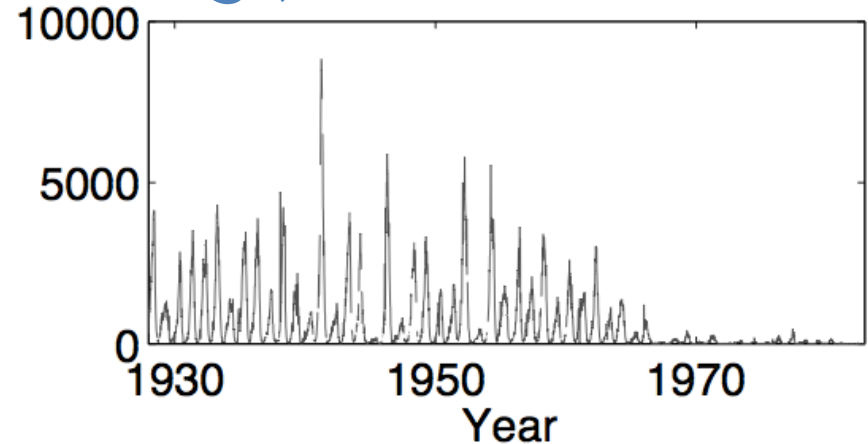
(b) FUNNEL-full



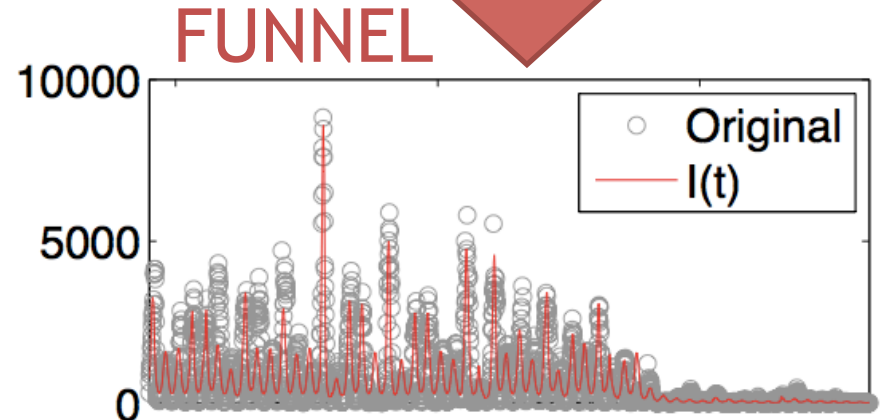
FUNNEL - with a single epidemic

Given:
“single” epidemic
sequence

e.g., measles in NY



Find:
nonlinear equation,
model parameters



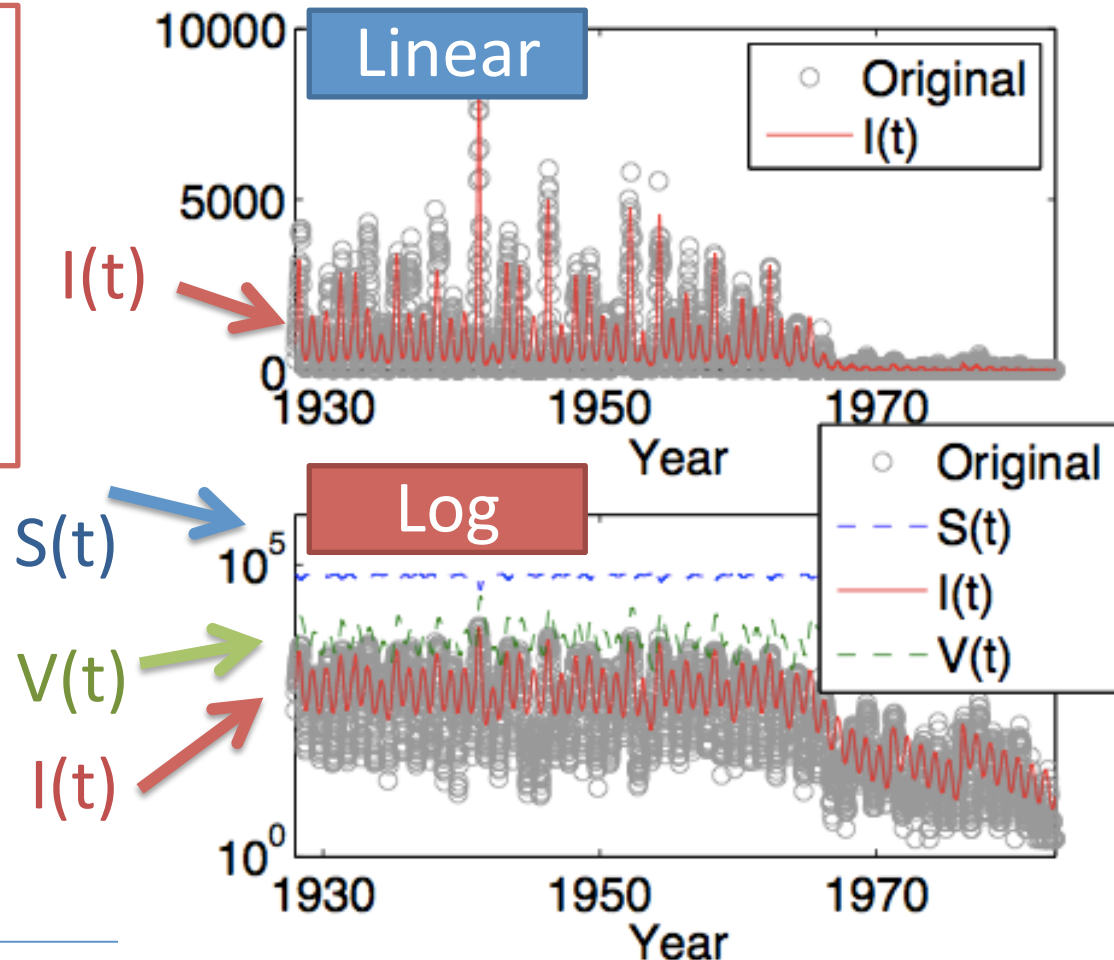
FUNNEL - with a single epidemic

Details

With a single epidemic: Funnel-RE

People of 3 classes

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



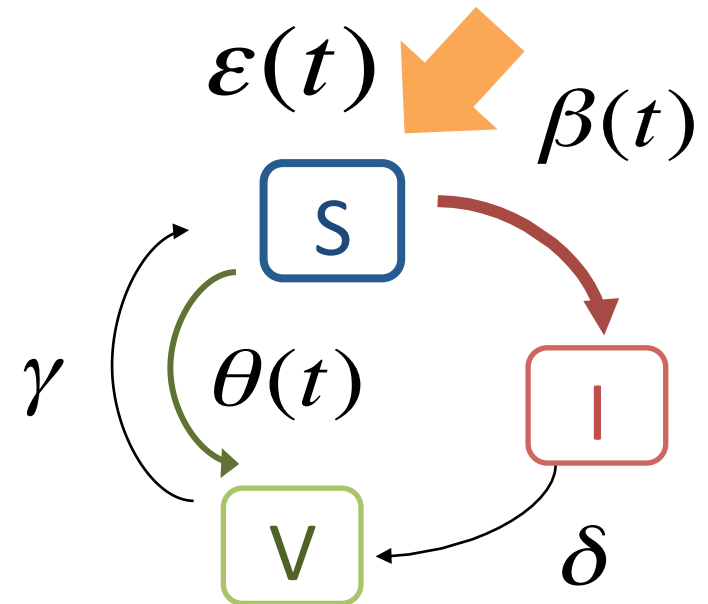
FUNNEL - with a single epidemic

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

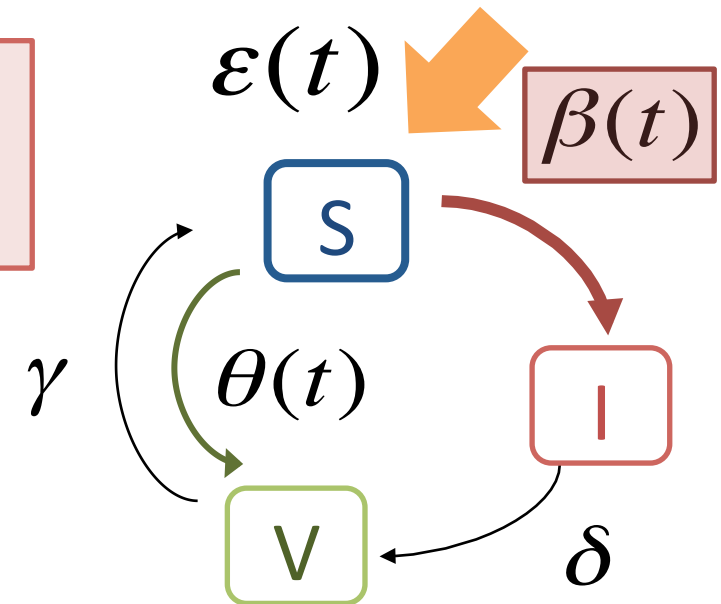
With a single epidemic: Funnel-RE

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

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

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

$P_p = 52$



FUNNEL - with a single epidemic

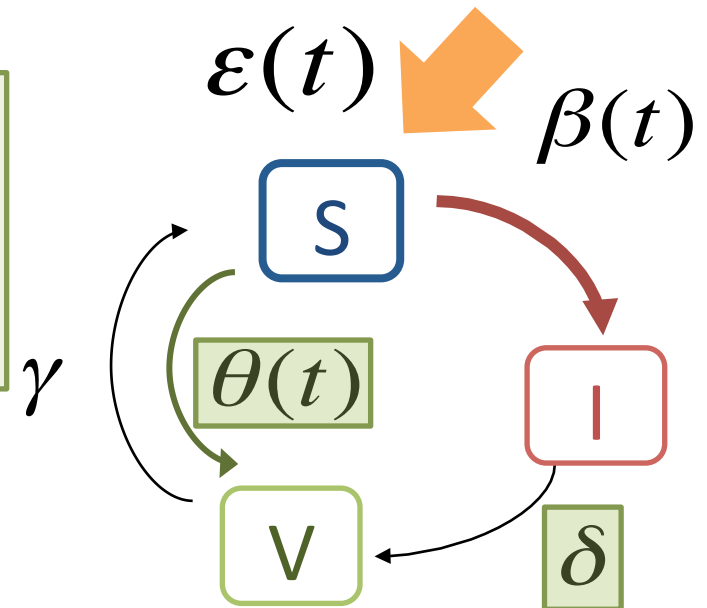
Details

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

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



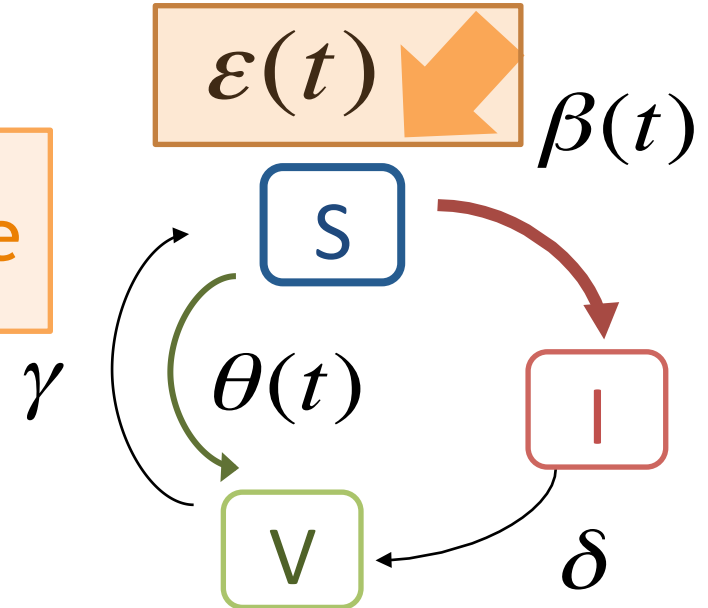
FUNNEL - with a single epidemic

Details

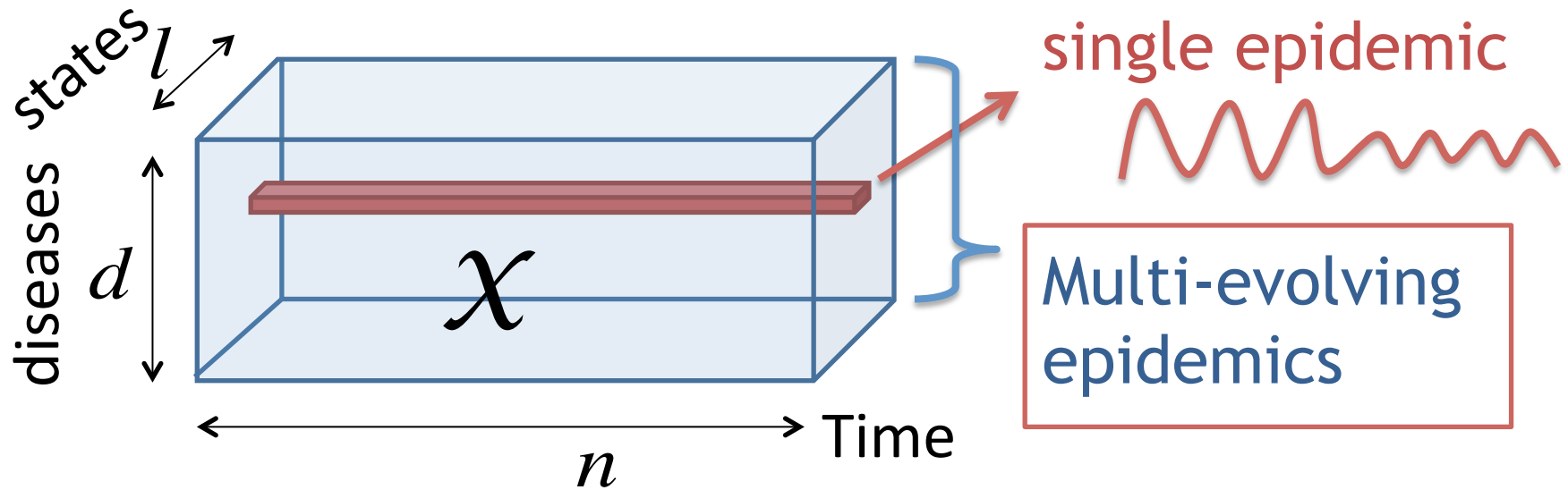
With a single epidemic: Funnel-RE

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

$\epsilon(t)$: temporal susceptible rate

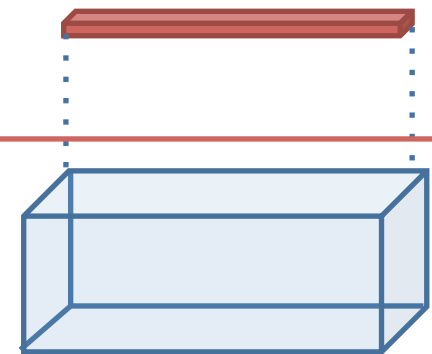


Proposed model: FUNNEL

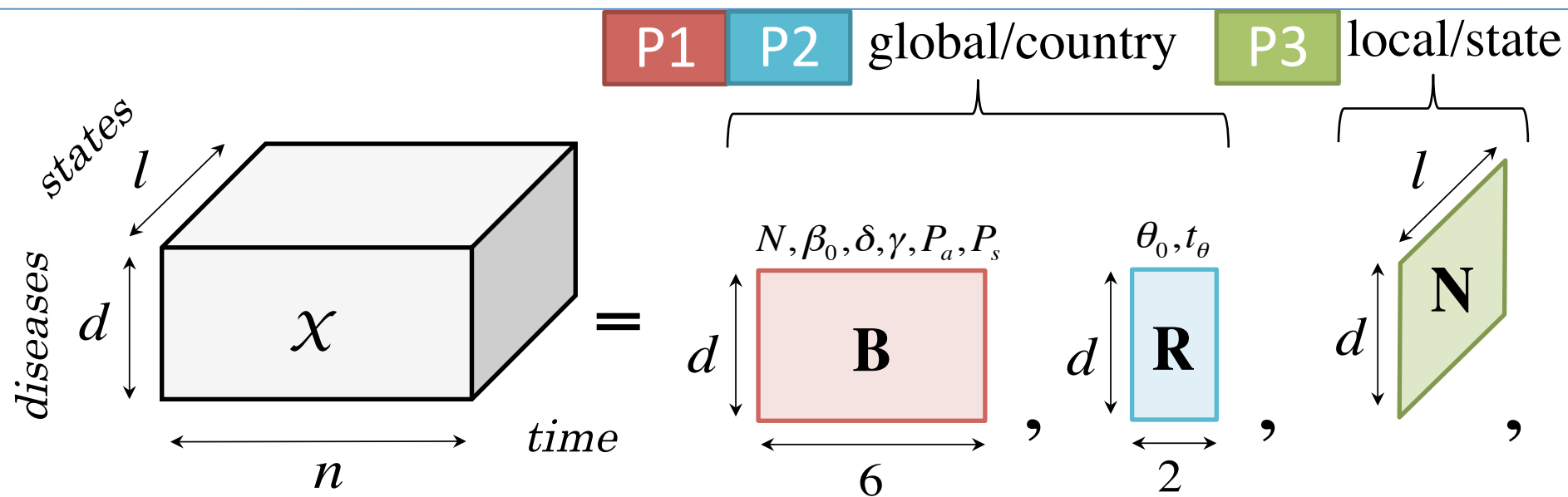


(a) FUNNEL-single

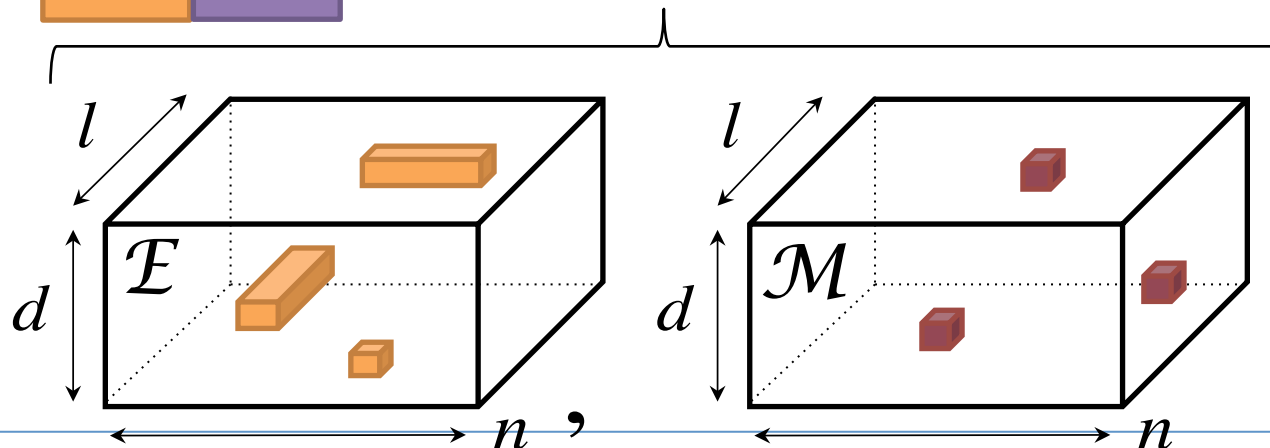
(b) FUNNEL-full



Proposed model: FUNNEL-full

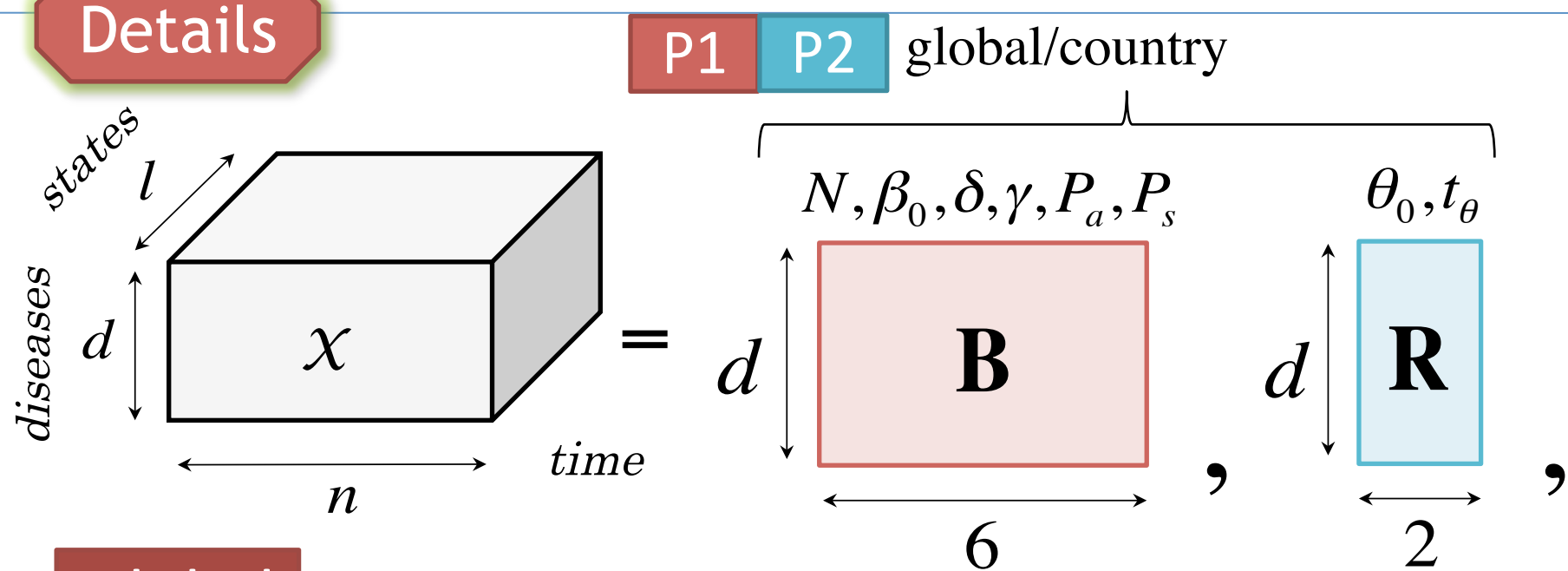


$P4$ $P5$ extra - \mathcal{E} : shocks & \mathcal{M} : mistakes



Proposed model: FUNNEL-full

Details



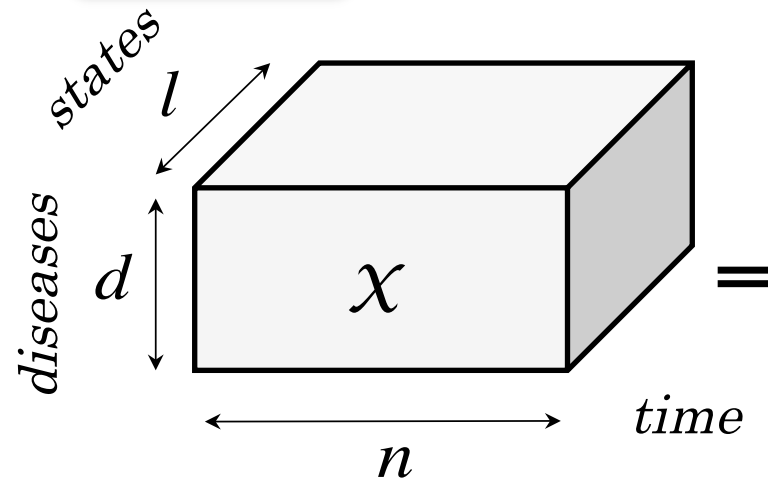
Global

P1 Base matrix \mathbf{B} ($d \times 6$)

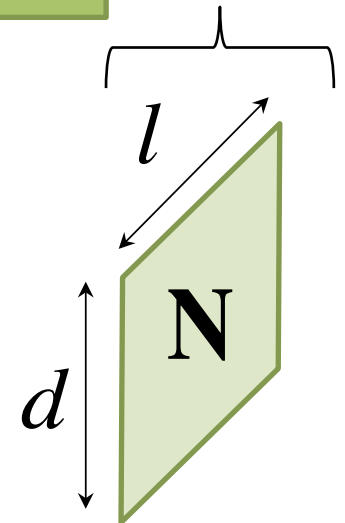
P2 Disease reduction matrix \mathbf{R} ($d \times 2$)

Proposed model: FUNNEL-full

Details



P3 local/state



Local

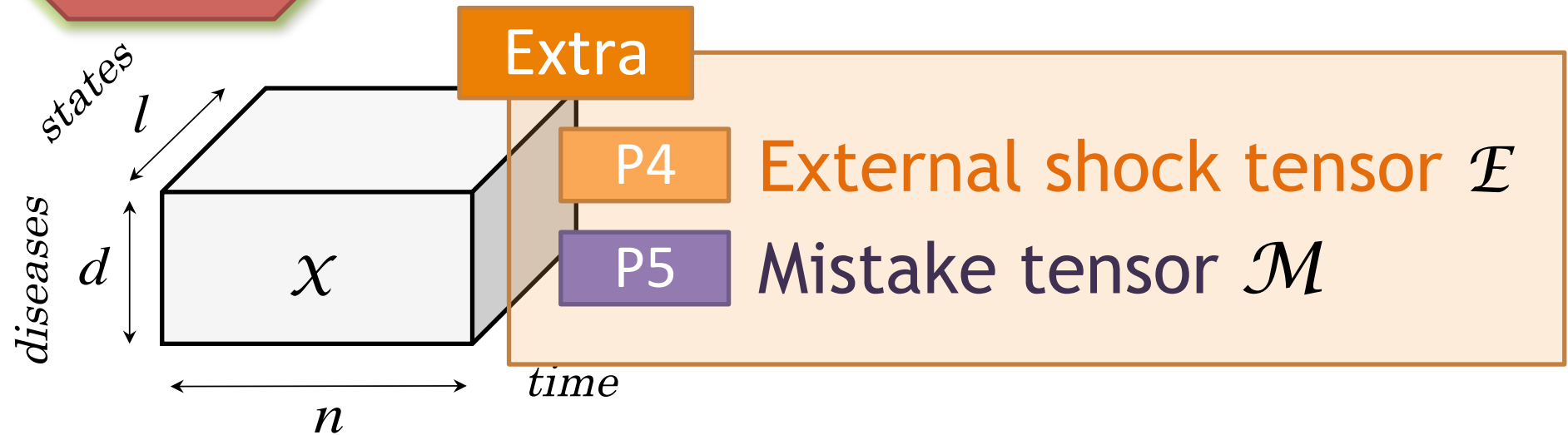
P3

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

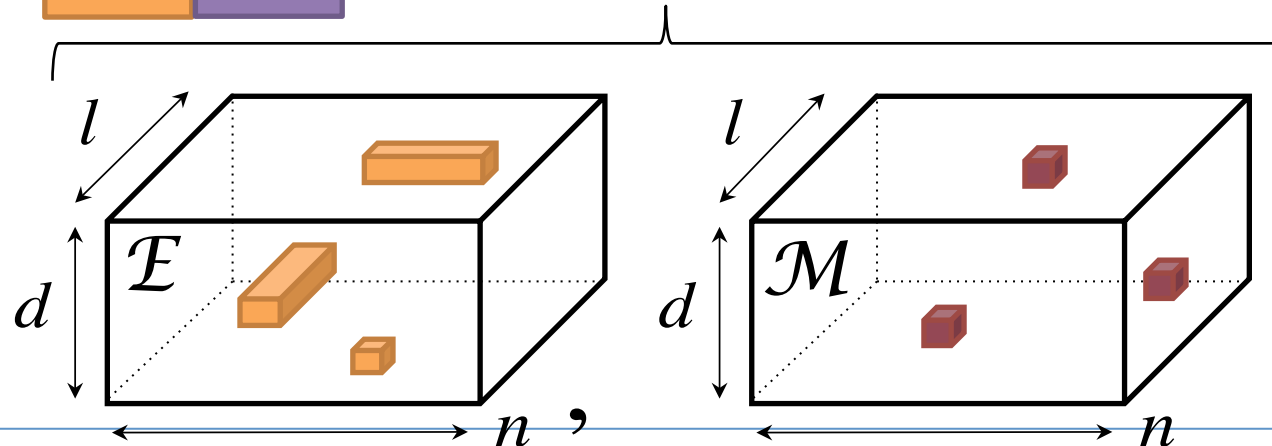
$\mathbf{N} = \{N_{ij}\}_{i,j=1}^{d,l}$: potential population of disease i in state j

Proposed model: FUNNEL-full

Details

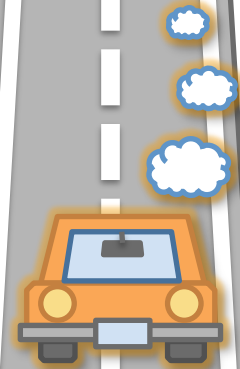


\mathcal{P}_4 \mathcal{P}_5 extra - \mathcal{E} : shocks & \mathcal{M} : mistakes



Roadmap

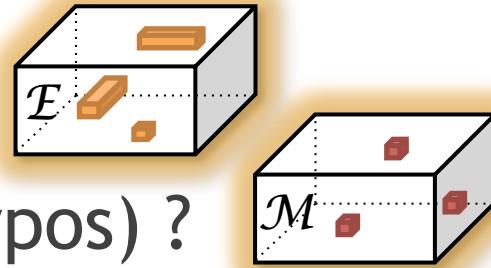
- ✓ Motivation
- ✓ Modeling power of FUNNEL
- ✓ Overview - main ideas
- ✓ Proposed model - idea #1
 - Algorithm - idea #2
 - Experiments
 - Discussion
 - Conclusions



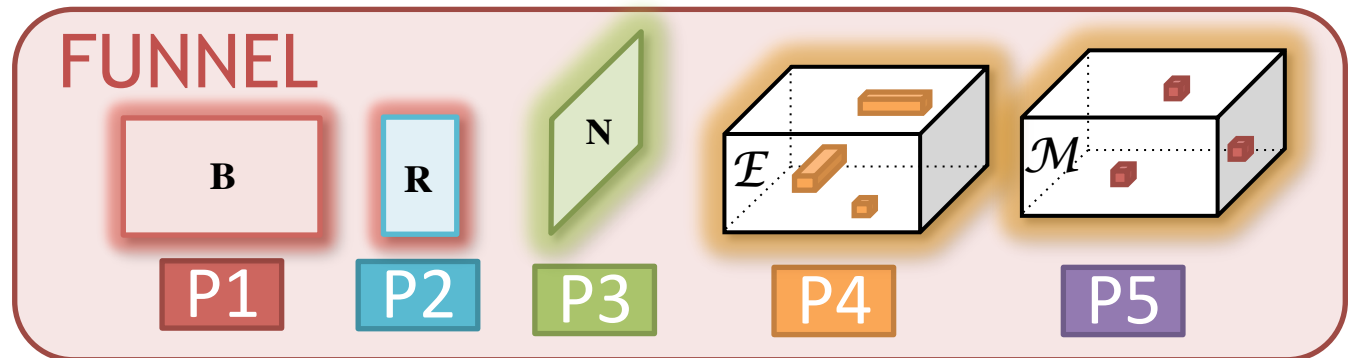
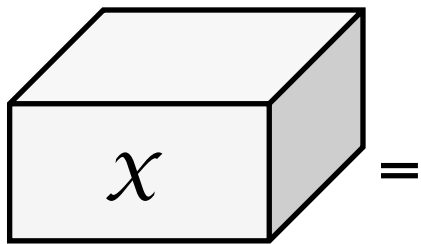
Challenges

Q1. How to automatically

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



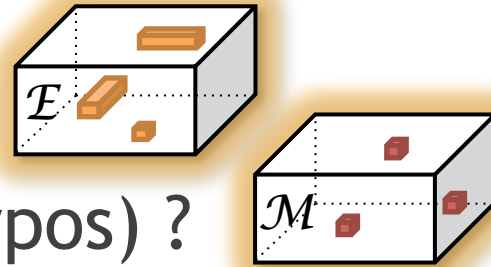
Q2. How to efficiently estimate model parameters ?



Challenges

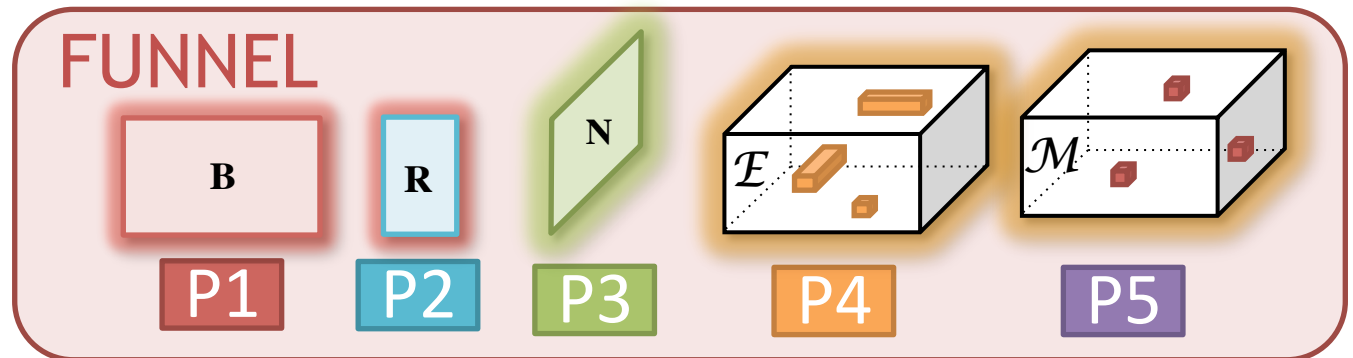
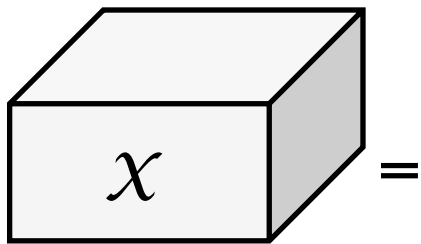
Q1. How to automatically

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



Idea (1) : Model description cost

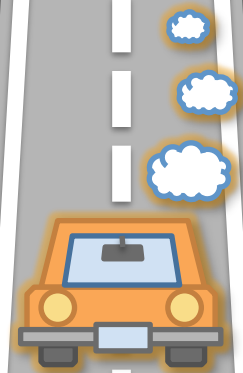
Q2. How to efficiently estimate model parameters ?



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

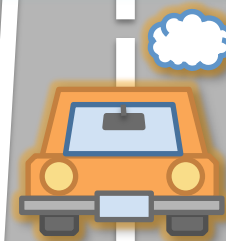
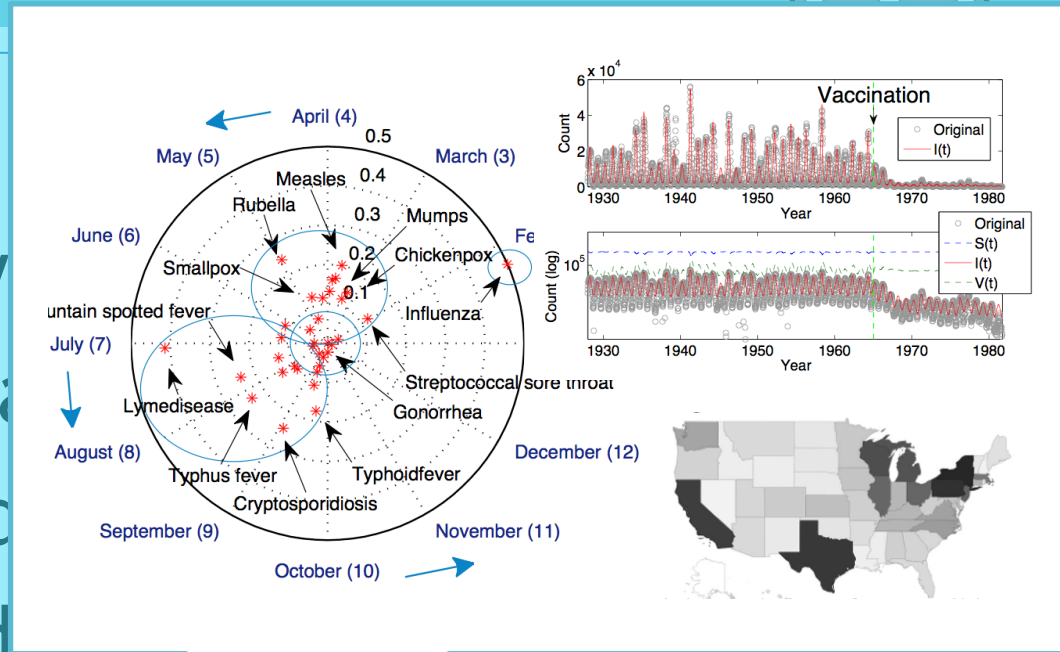
Roadmap

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



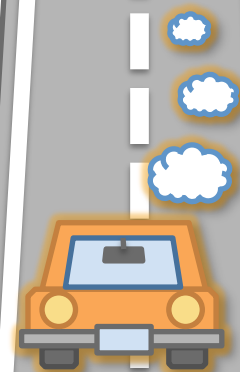
Roadmap

- ✓ Motivation
- ✓ Modeling power
- ✓ Overview - ma
- ✓ Proposed mod
- ✓ Algorithm - id
- ✓ Experiments
 - Discussion
 - Conclusions



Roadmap

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

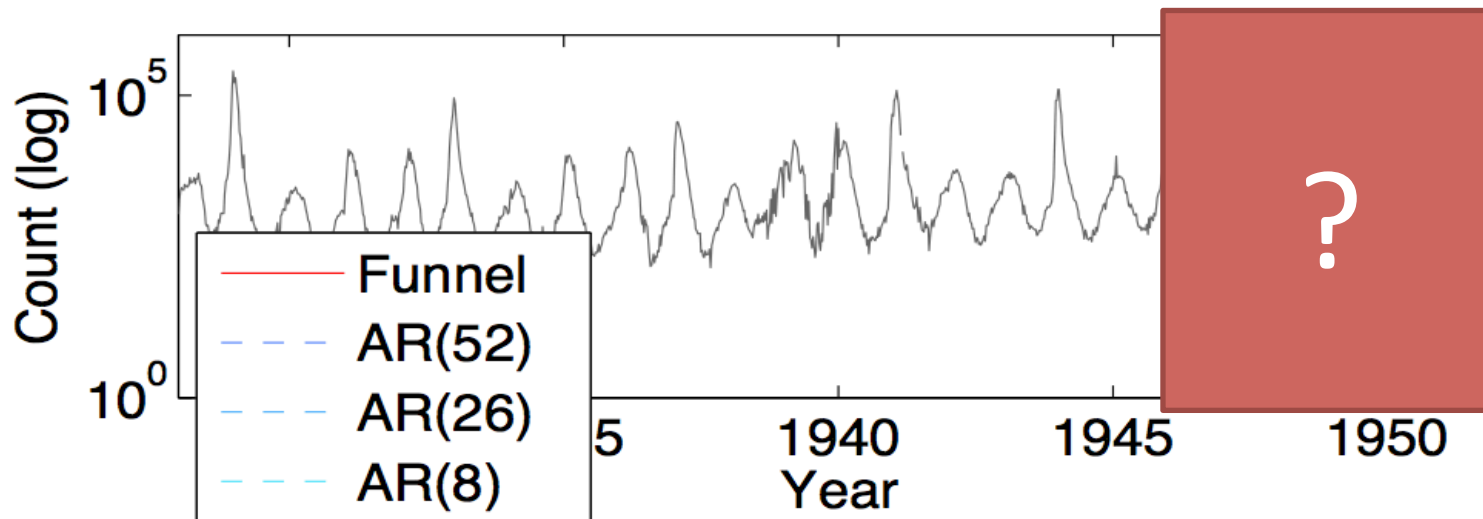


(a) FUNNEL at work - forecasting

Forecasting future epidemics

Train:
2/3 sequences

Forecast:
1/3 following years



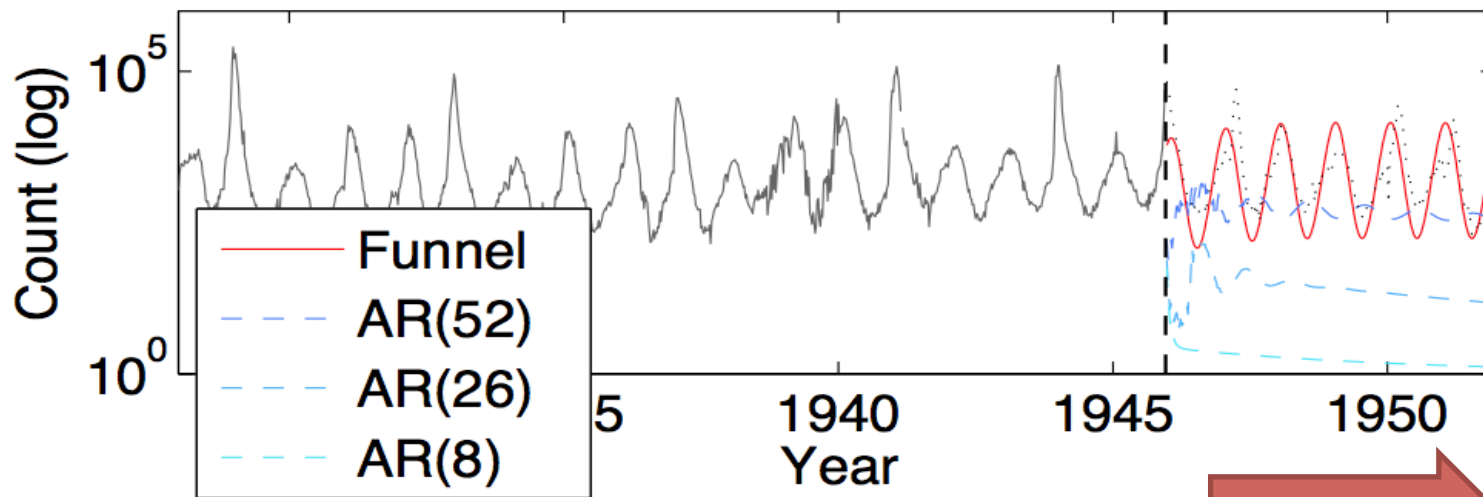
(a) Influenza

(a) FUNNEL at work - forecasting

Forecasting future epidemics

Train:
2/3 sequences

Forecast:
1/3 following years



(a) Influenza

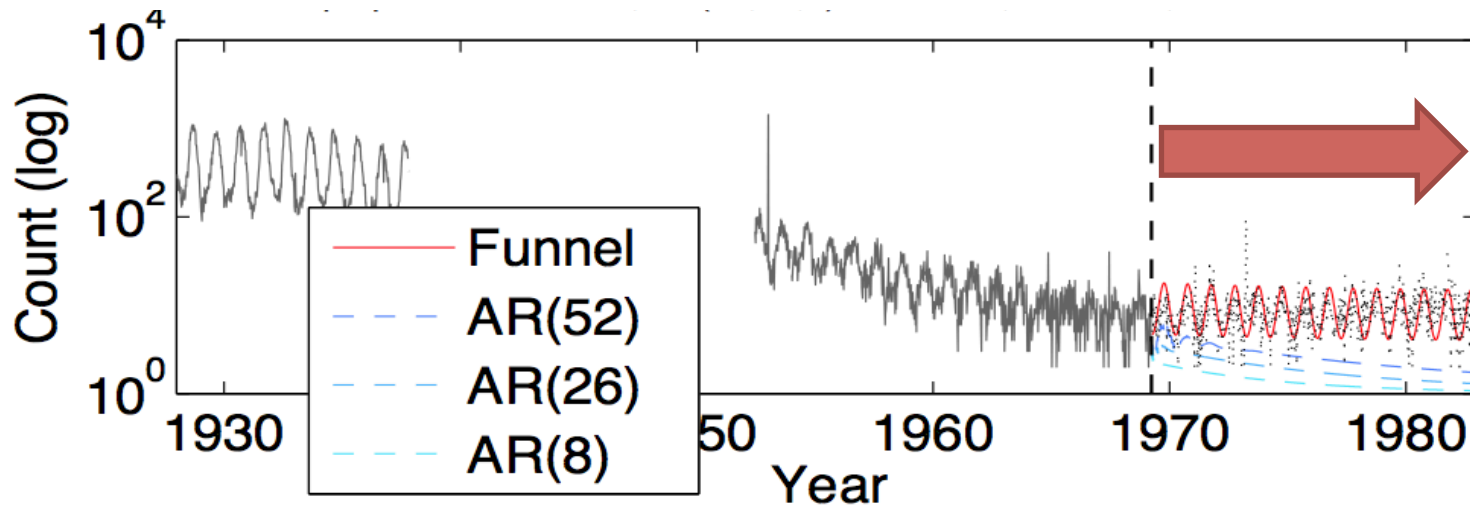
Funnel can capture future epidemics (AR: fail)

(a) FUNNEL at work - forecasting

Forecasting future epidemics

Train:
2/3 sequences

Forecast:
1/3 following years

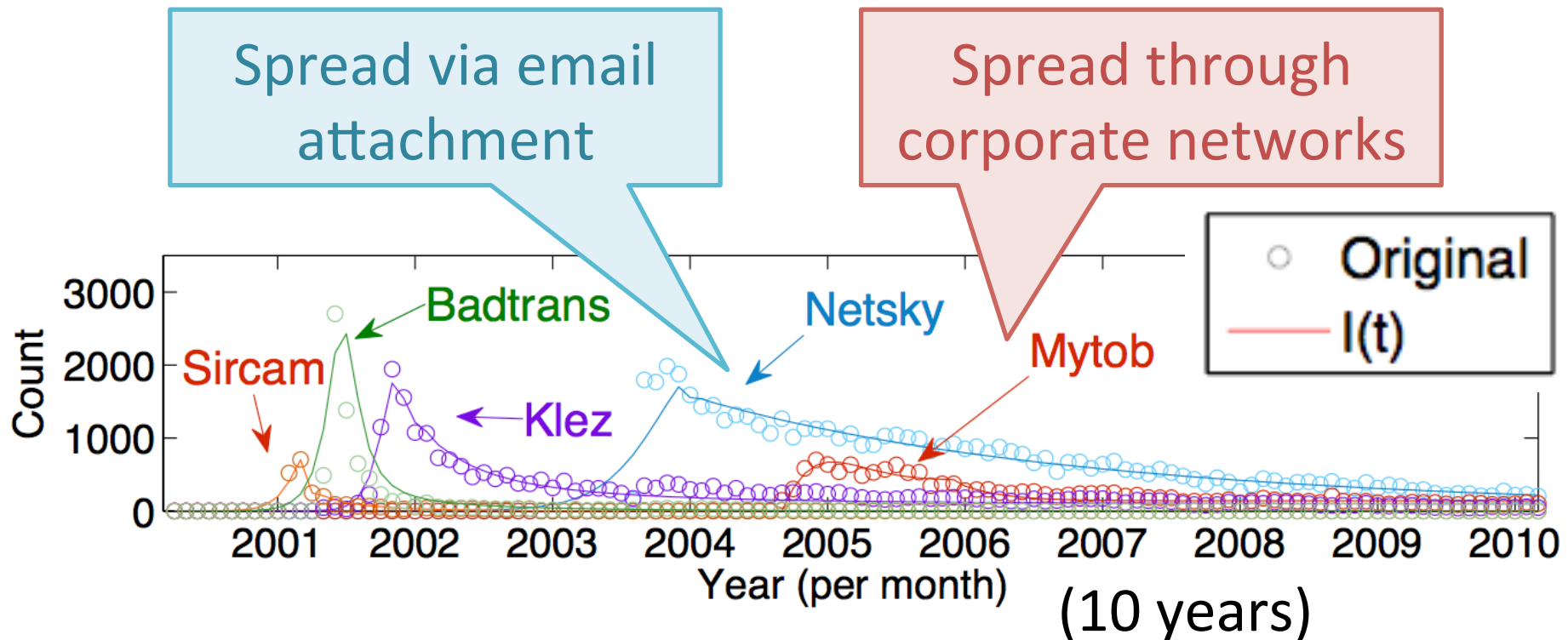


(c) Typhoid fever

Funnel can capture future epidemics (AR: fail)

(b) Generality of FUNNEL

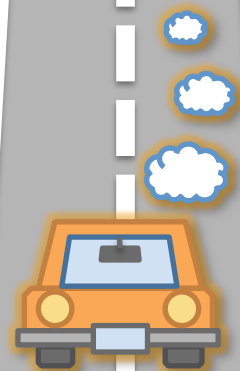
Epidemics on computer networks



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**

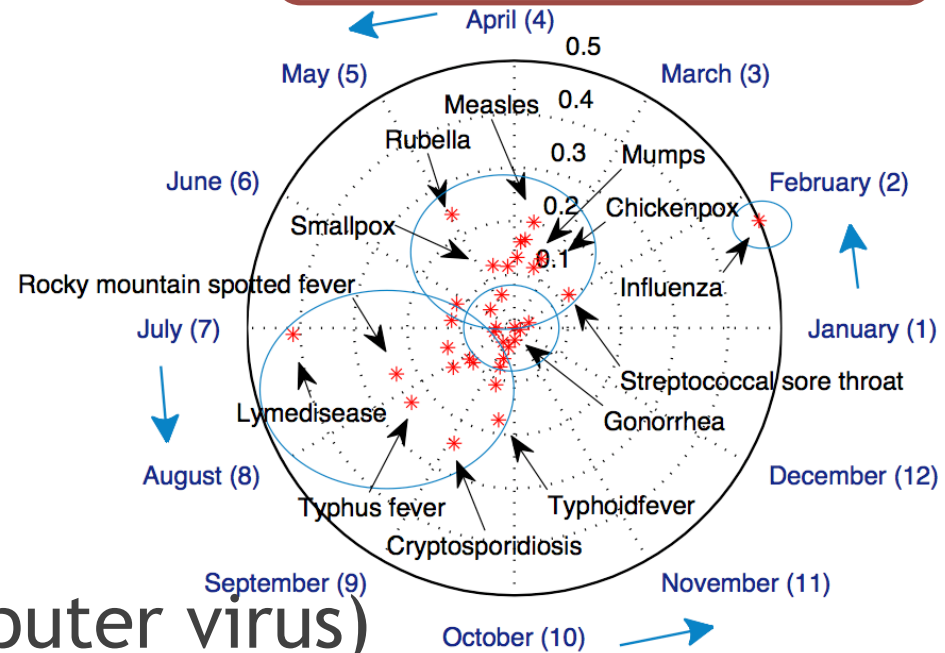
No training set

✓ **Scalable**

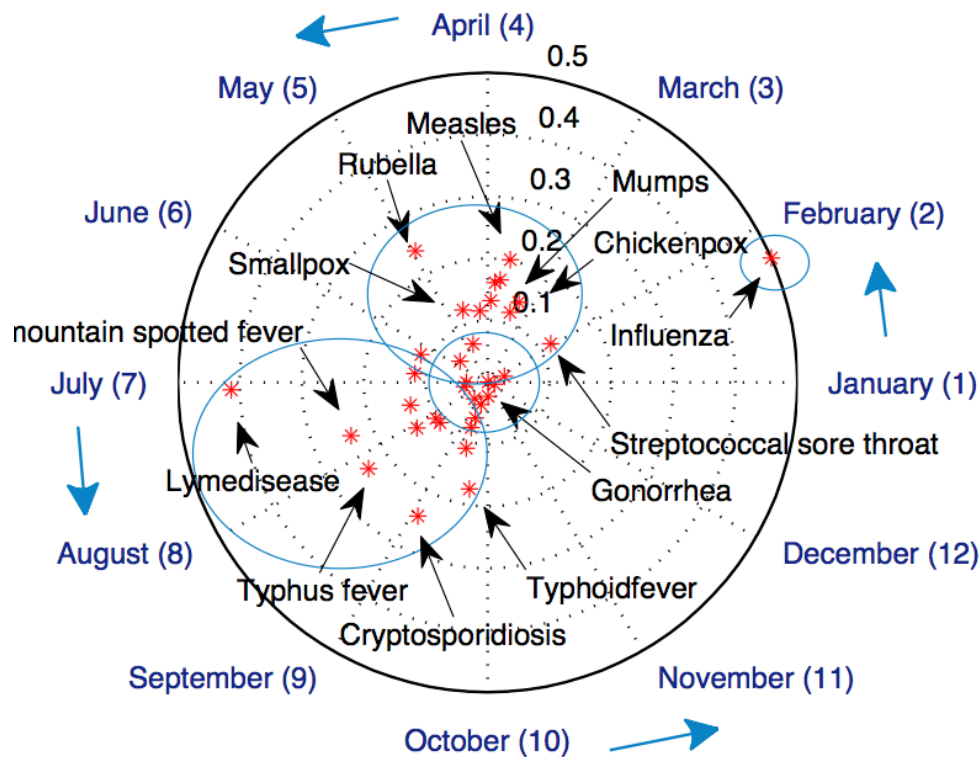
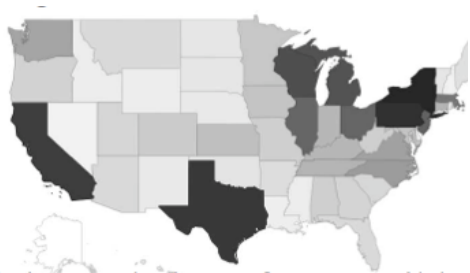
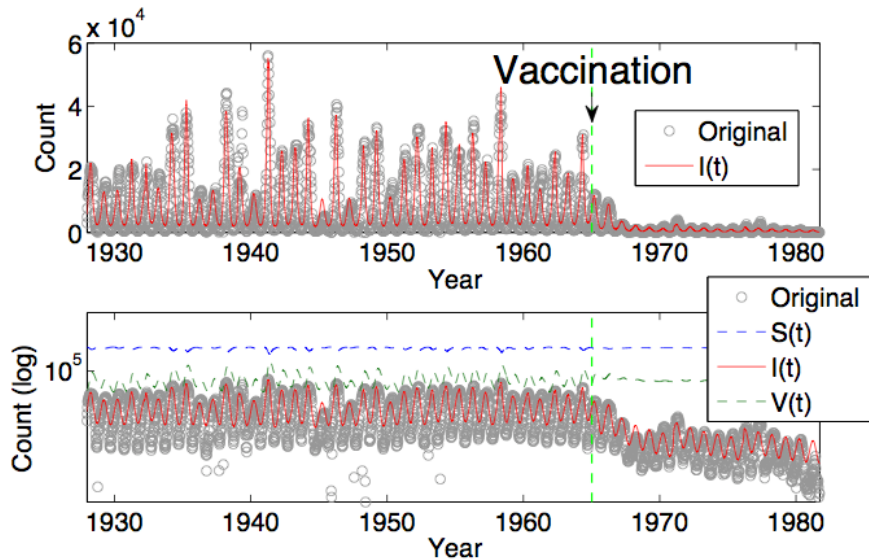
It scales linearly

✓ **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)

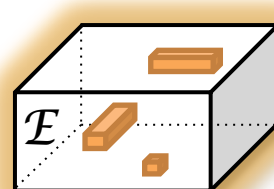


Proposed model: FUNNEL-full

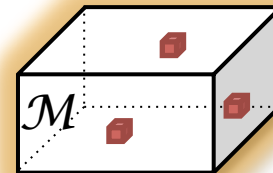
Details

What's the difference??

External shock vs. Mistake

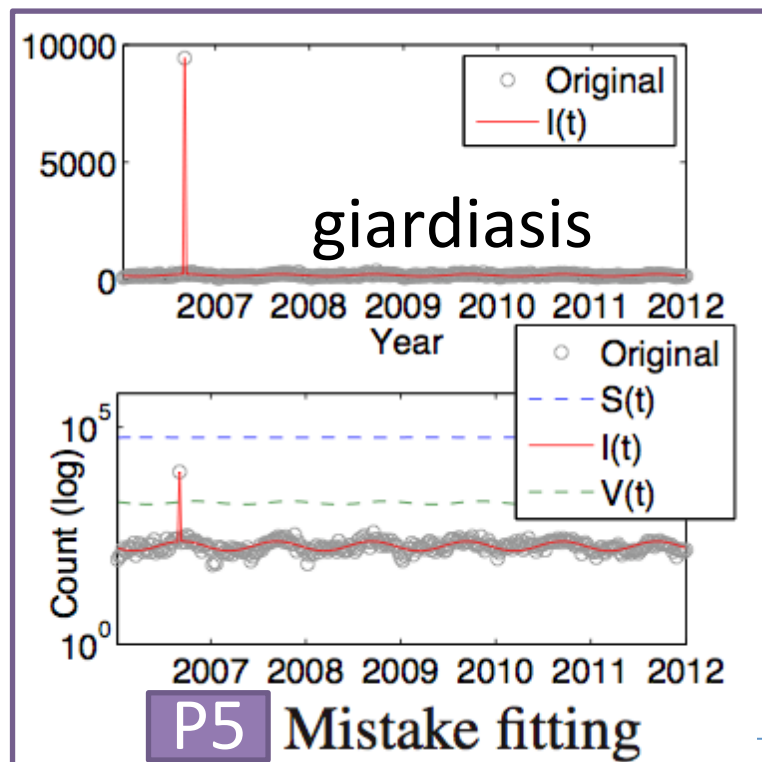
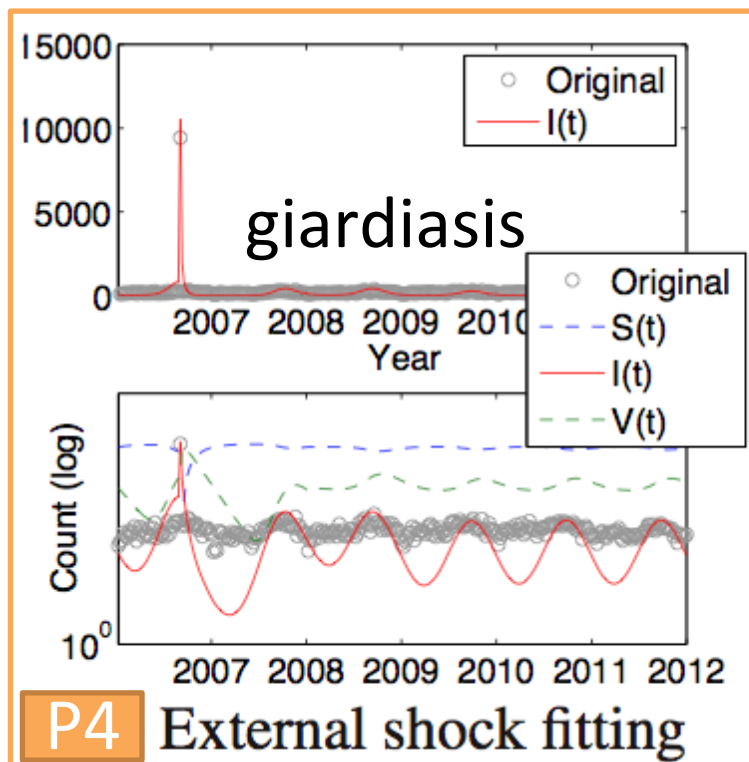


vs.



P4

P5

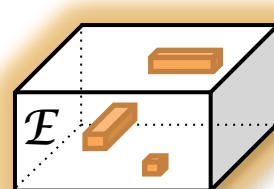


Proposed model: FUNNEL-full

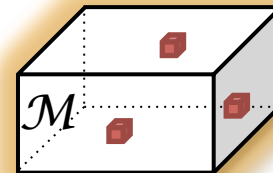
Details

What's the difference??

External shock vs. Mistake

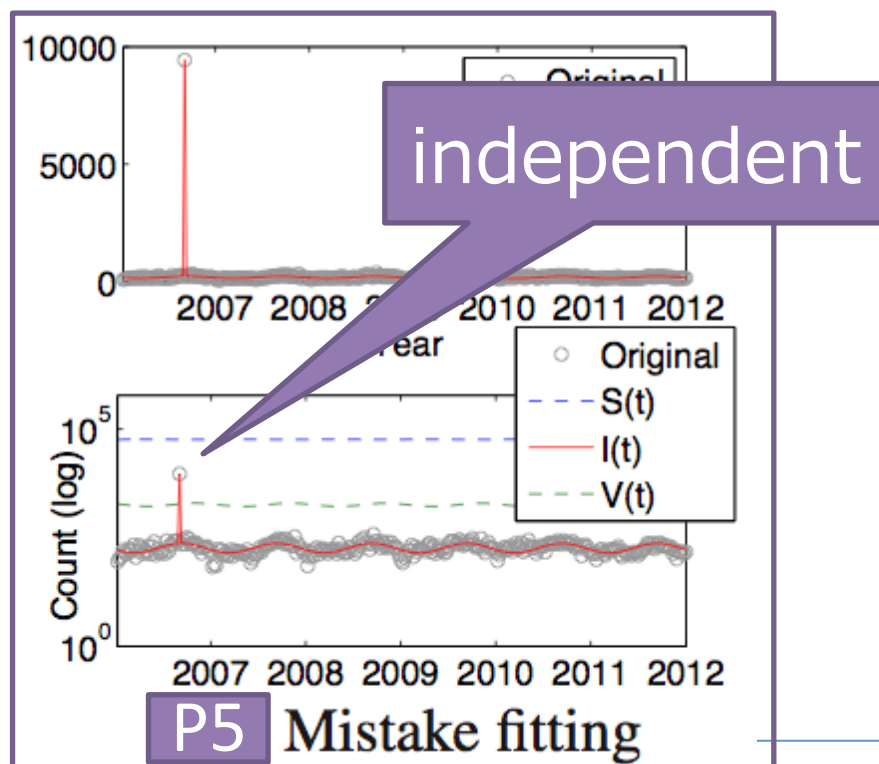
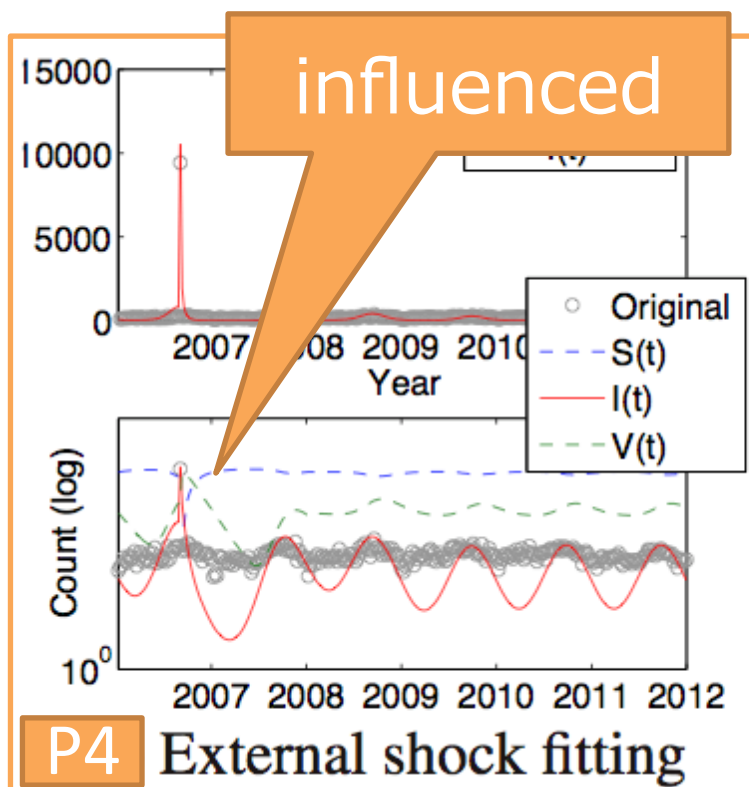


vs.



P4

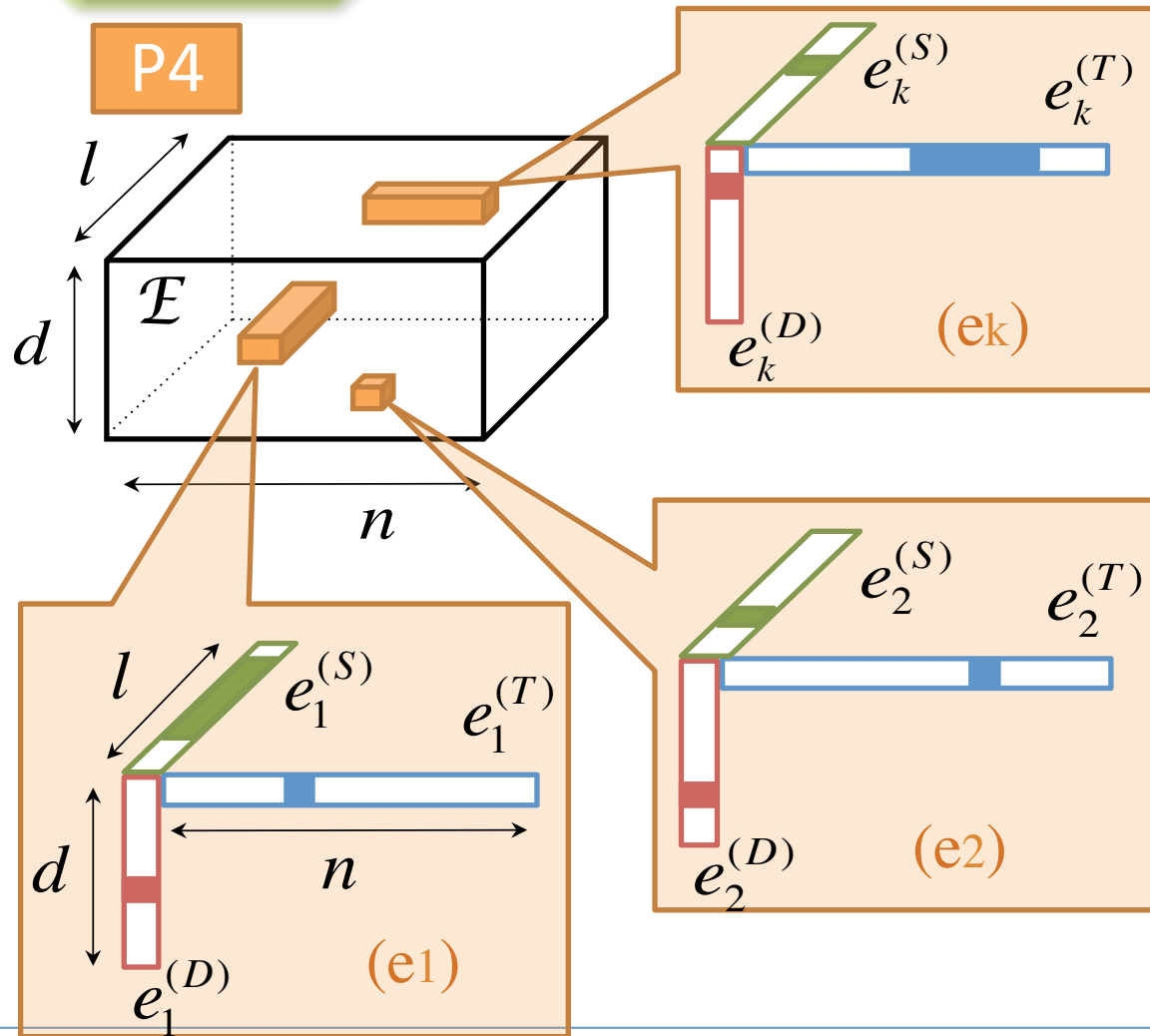
P5



Proposed model: FUNNEL-full

Details

P4



$$= \begin{matrix} \begin{matrix} l & \longleftrightarrow & k \\ \text{E}^{(S)} \end{matrix} \\ \begin{matrix} t_\mu, t_\sigma, \epsilon_0 \\ \text{E}^{(T)} \\ k \end{matrix} \\ \begin{matrix} e^{(D)} \\ \text{E}^{(D)} \\ k \end{matrix} \begin{matrix} \updownarrow 1 \\ \leftarrow 3 \end{matrix} \end{matrix}$$

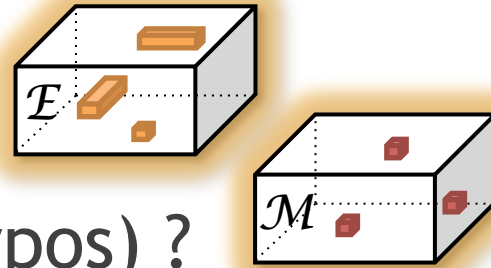
$$\mathcal{F} = \{ \mathbf{E}^{(D)}, \mathbf{E}^{(T)}, \mathbf{E}^{(S)} \}$$

Disease matrix Time matrix State matrix

Idea (1): Model description cost

Q1. How should we

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



Idea (1) : Model description cost

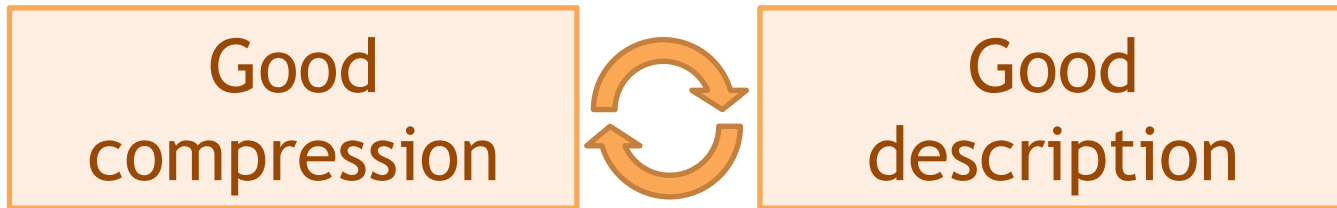
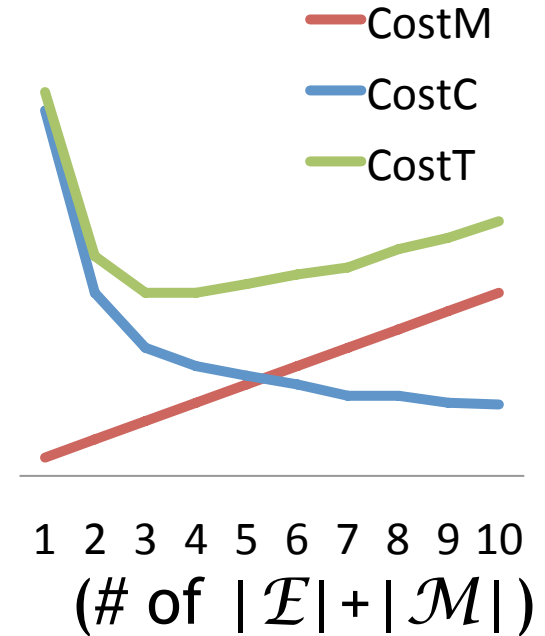
- Minimize coding cost
- find “optimal” # of externals/mistakes
- “automatically”

Idea (1): Model description cost

Idea: Minimize encoding cost!

$$\min \left(\boxed{\text{Cost}_M(\mathcal{F})} + \boxed{\text{Cost}_c(\mathcal{X} | \mathcal{F})} \right)$$

Model cost Coding cost



Idea (1): Model description cost

Details

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

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

$$\begin{aligned} \text{Cost}_T(\mathcal{X}; \mathcal{F}) &= \log^*(d) + \log^*(l) + \log^*(n) \\ &+ \text{Cost}_M(\mathbf{B}) + \text{Cost}_M(\mathbf{R}) + \text{Cost}_M(\mathbf{N}) \\ &+ \text{Cost}_M(\mathcal{E}) + \text{Cost}_M(\mathcal{M}) + \text{Cost}_C(\mathcal{X}|\mathcal{F}) \end{aligned}$$

Idea (1): Model description cost

Details

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

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

Dimensions of \mathcal{X}

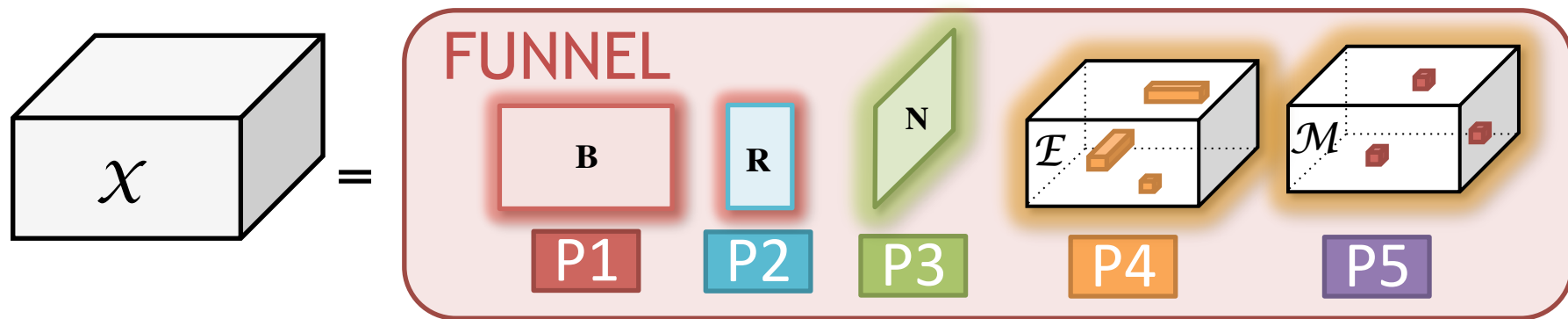
$$\begin{aligned} \text{Cost}_T(\mathcal{X}; \mathcal{F}) = & \log^*(d) + \log^*(l) + \log^*(n) \\ & + \text{Cost}_M(\mathbf{B}) + \text{Cost}_M(\mathbf{R}) + \text{Cost}_M(\mathbf{N}) \\ & + \text{Cost}_M(\mathcal{E}) + \text{Cost}_M(\mathcal{M}) + \text{Cost}_C(\mathcal{X}|\mathcal{F}) \end{aligned}$$

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** ?

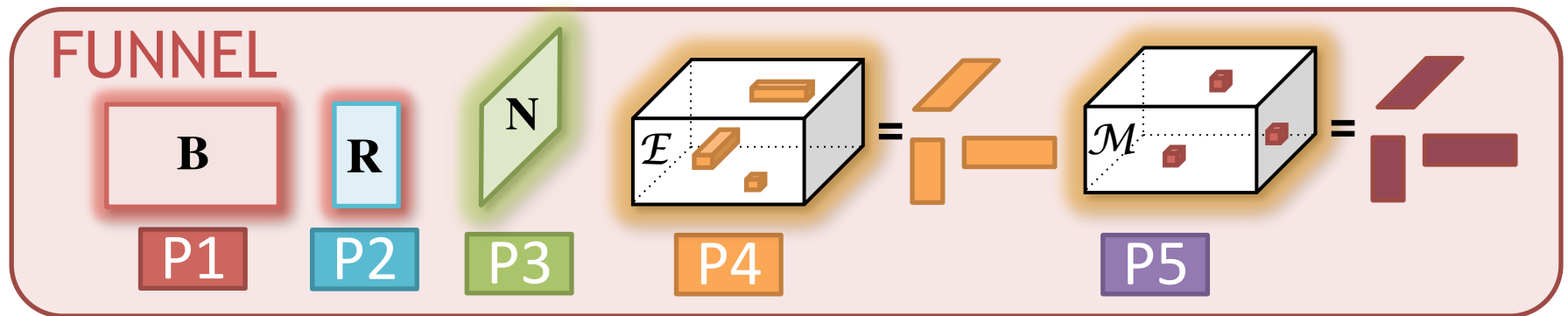


Idea (2): Multi-layer optimization

- 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**

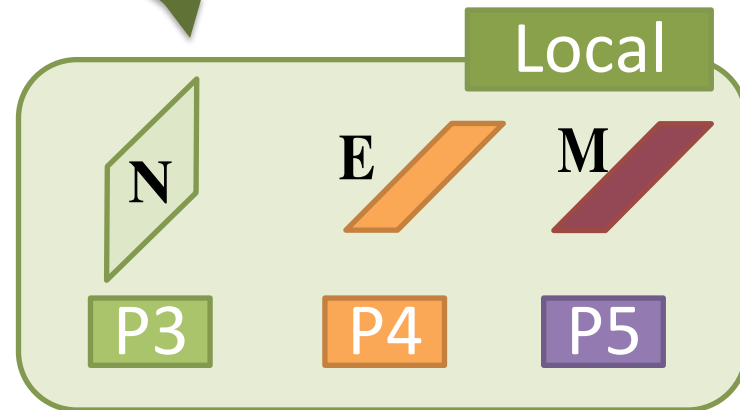
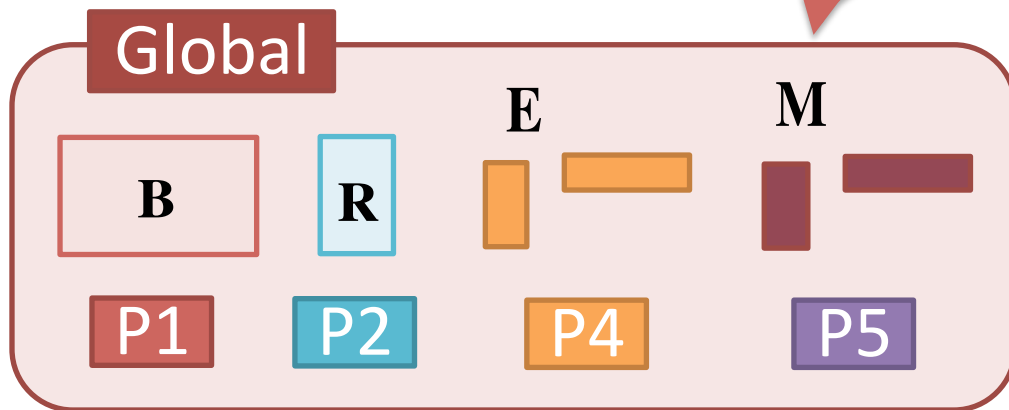
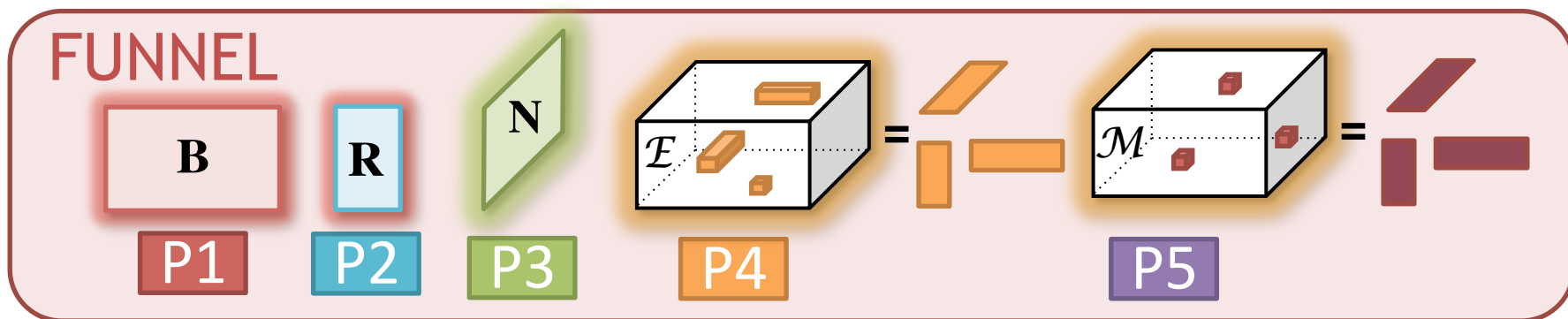


Global

Local

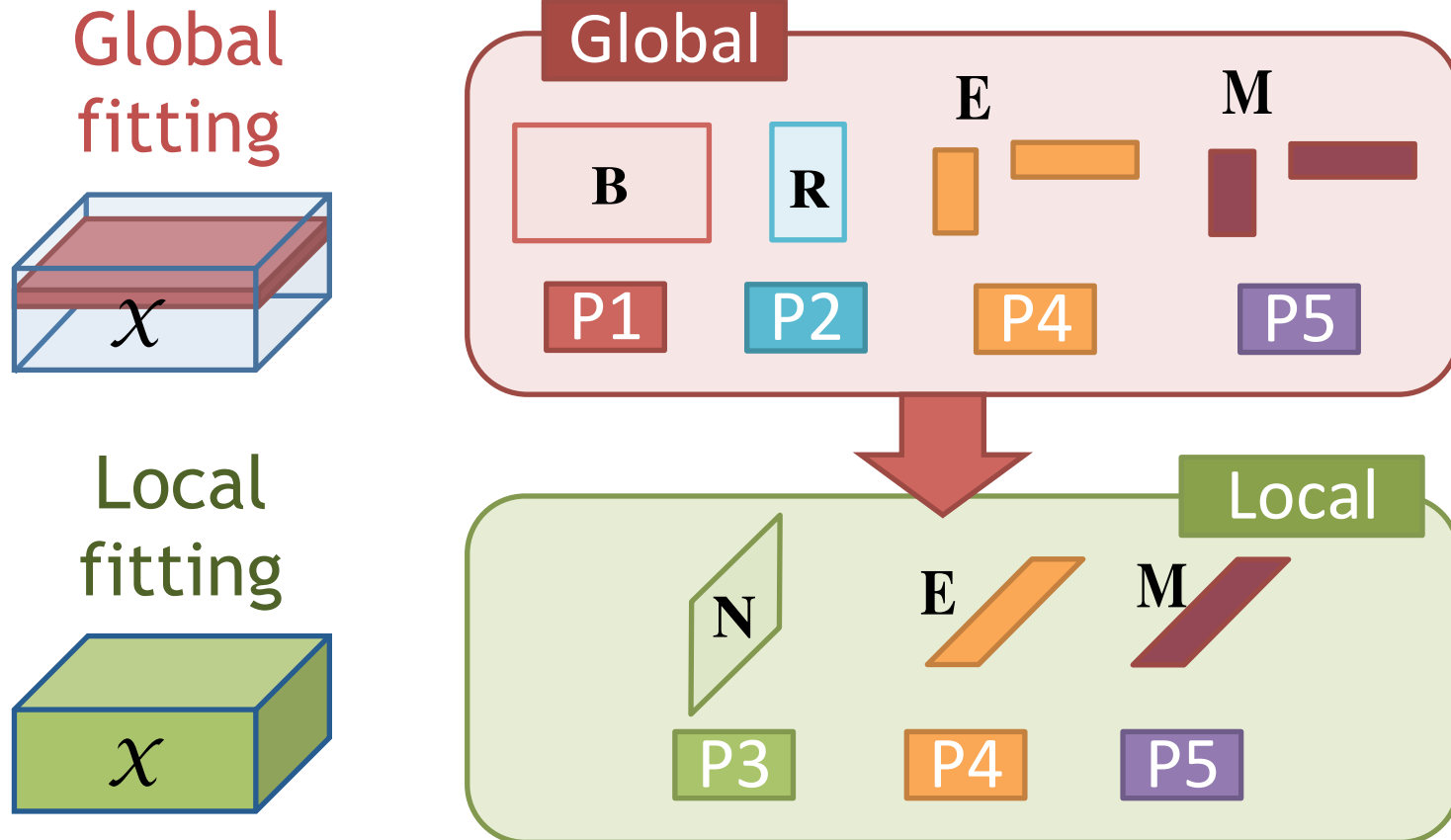
Idea (2): Multi-layer optimization

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

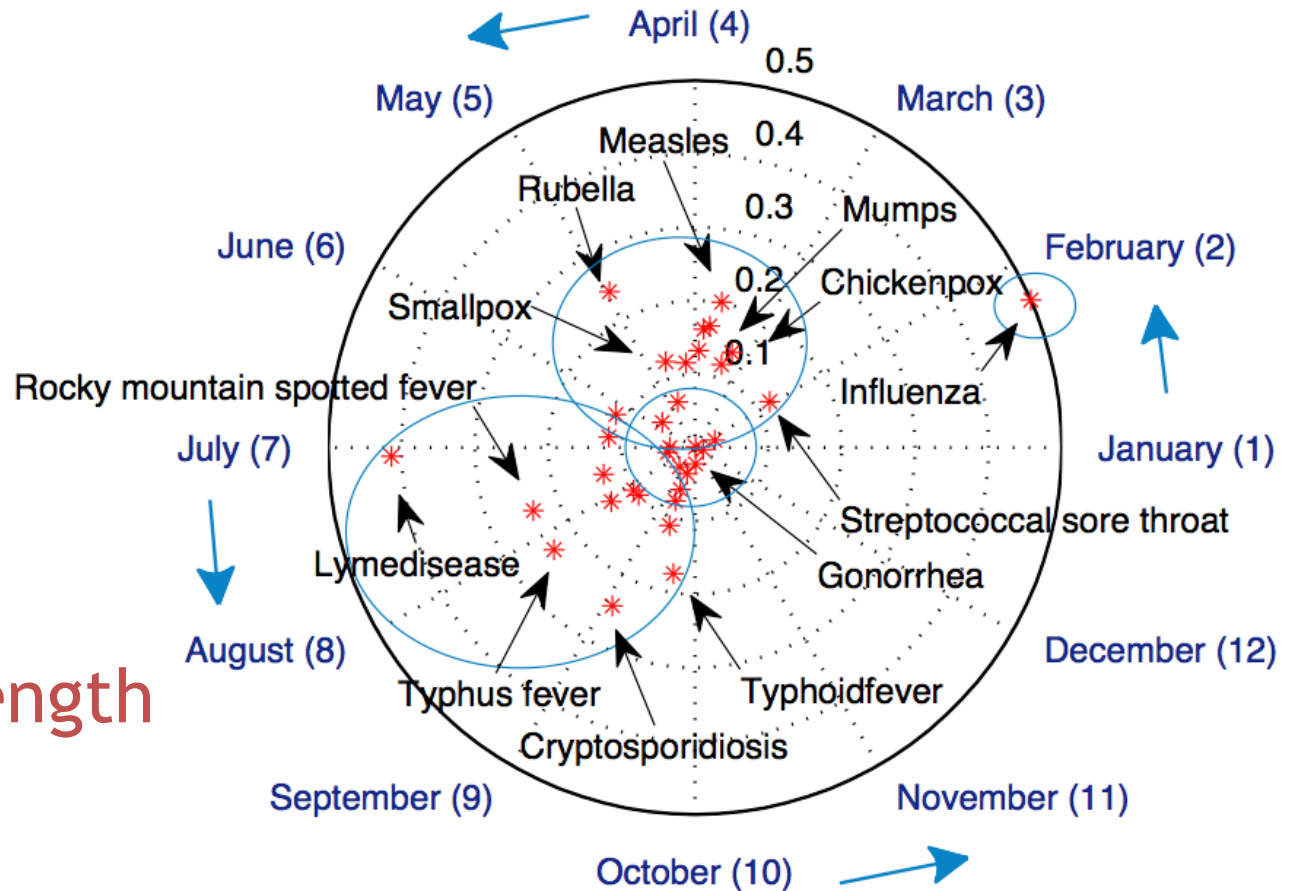
external shock events

P5

mistakes, incorrect values

Q1. Sense-making

P1 Disease seasonality

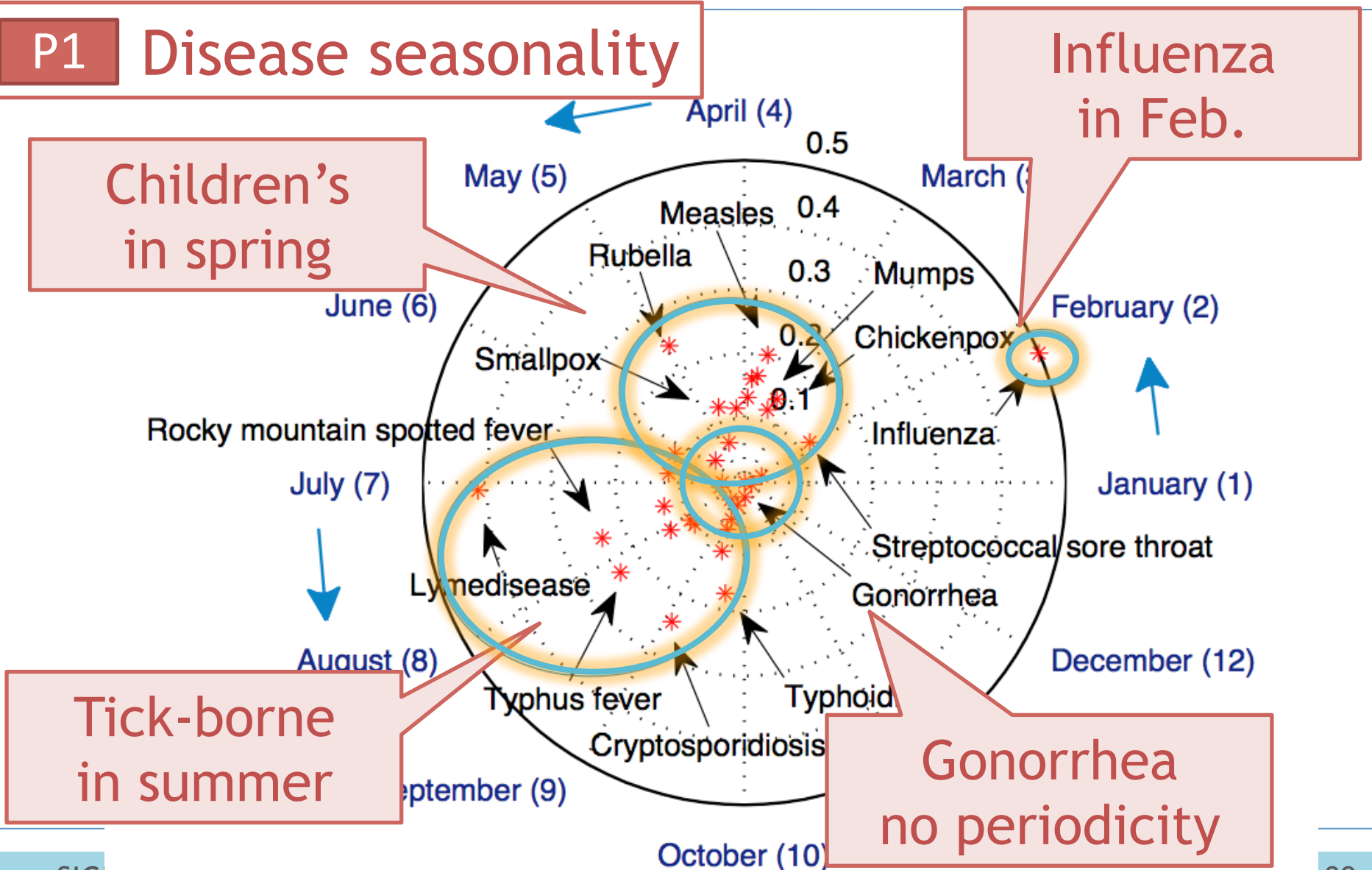


Radius:
seasonality strength

Angle:
peak season

Q1. Sense-making

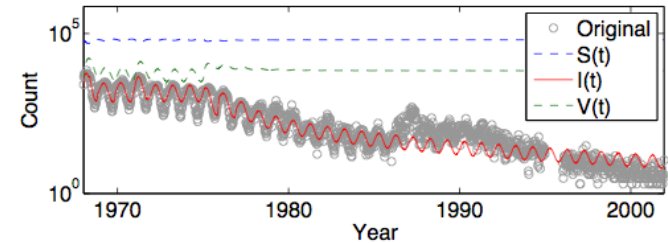
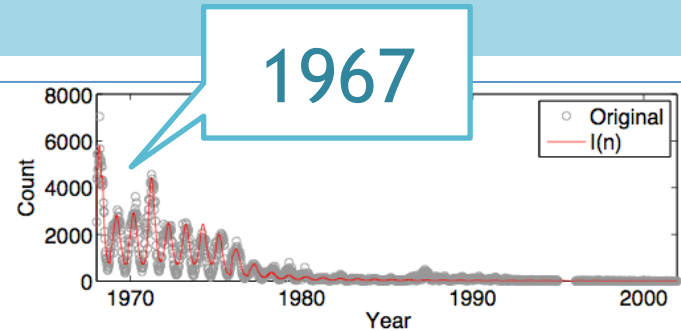
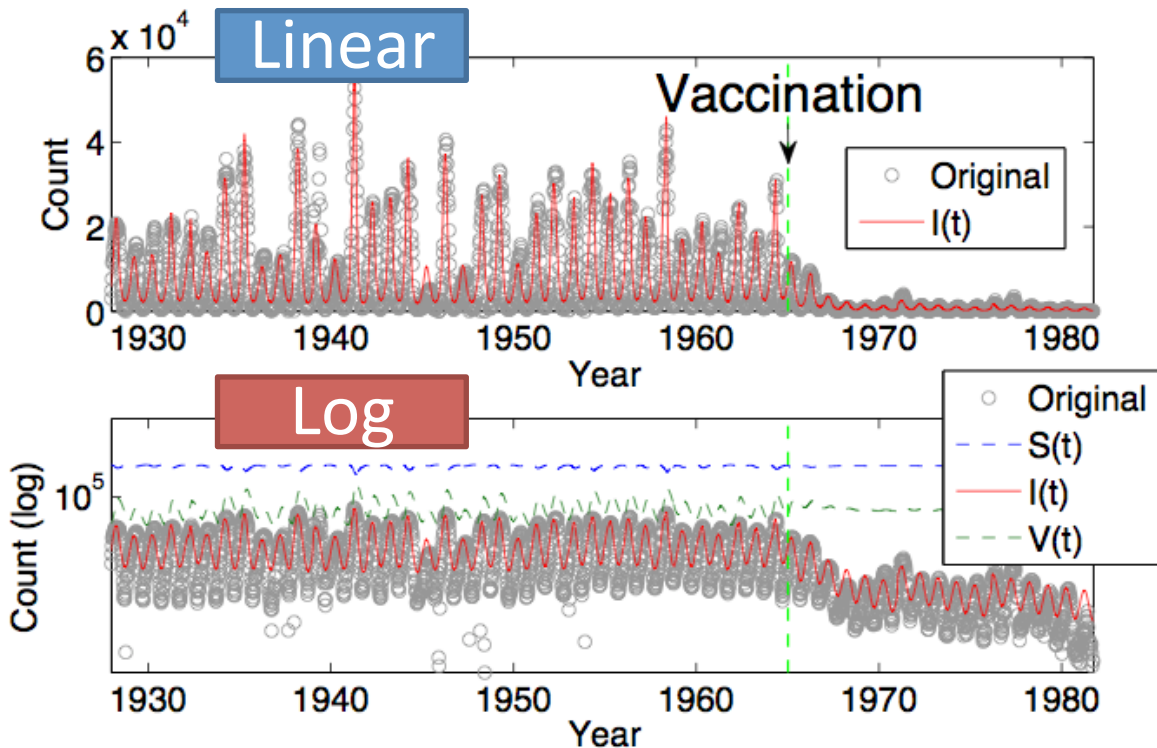
P1 Disease seasonality



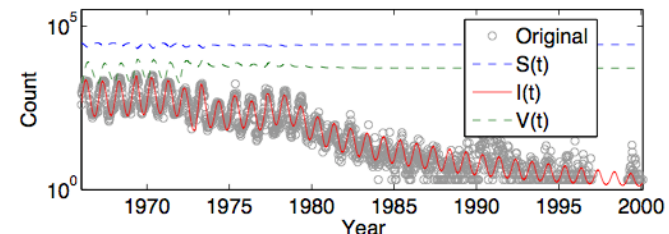
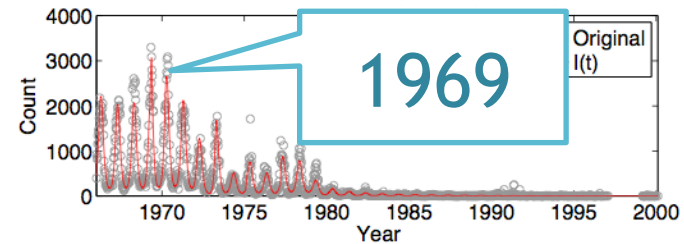
Q1. Sense-making

P2 Disease reduction effect

Measles (vaccine licensure: 1963)



(c) Mumps (P1), (P2), (P4)

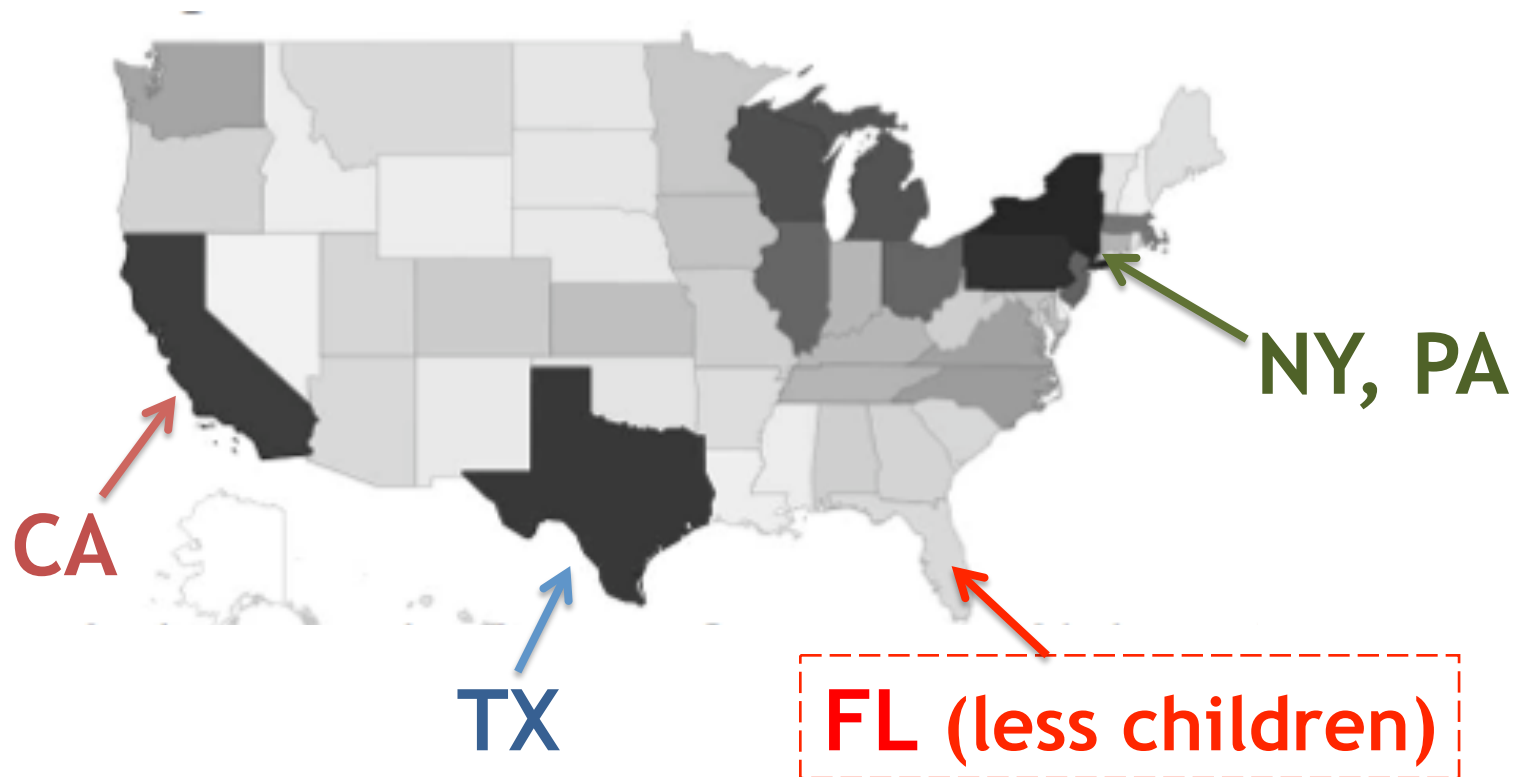


(d) Rubella (P1), (P2), (P4)

Q1. Sense-making

P3 area specificity and sensitivity

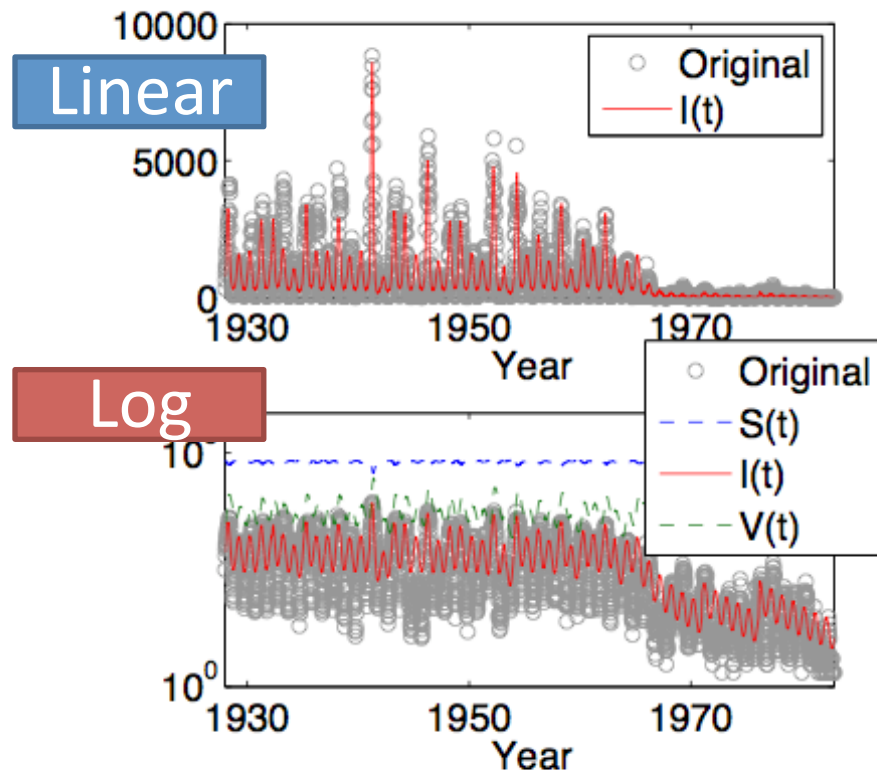
Potential population of susceptibles (measles)



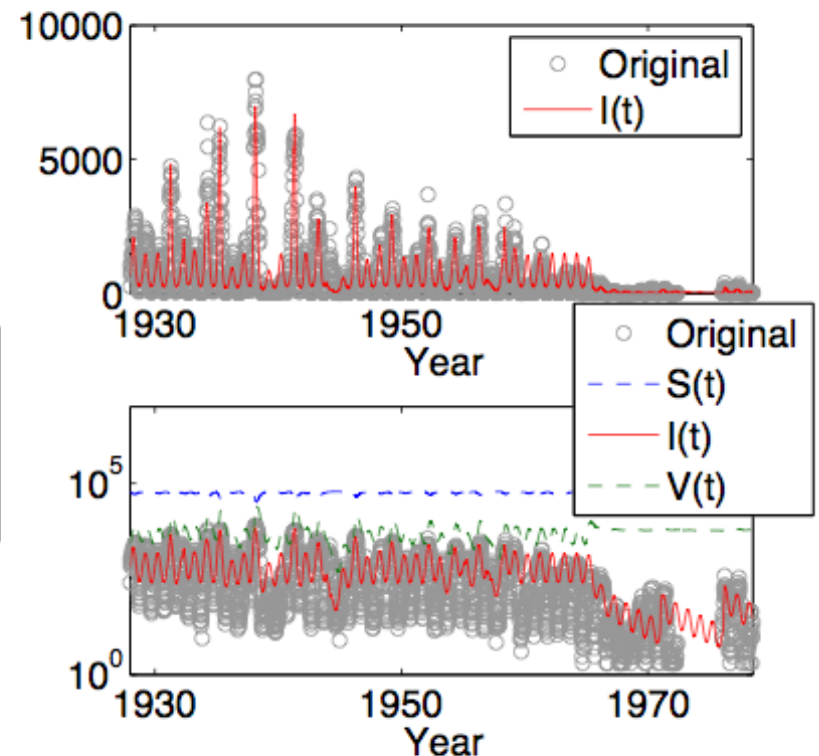
Q1. Sense-making

P3 area specificity and sensitivity

Measles in NY and PA



(a) New York State (NY)

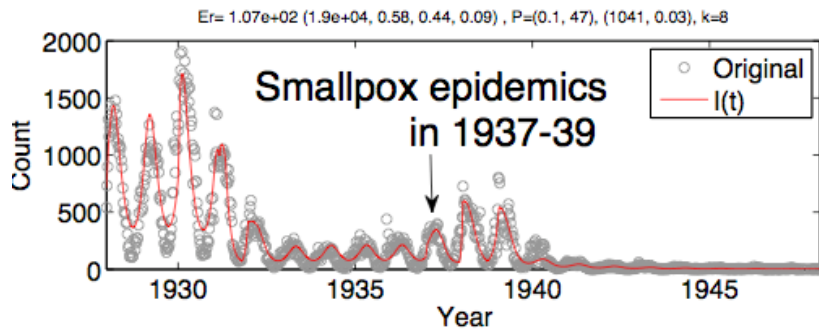


(b) Pennsylvania (PA)

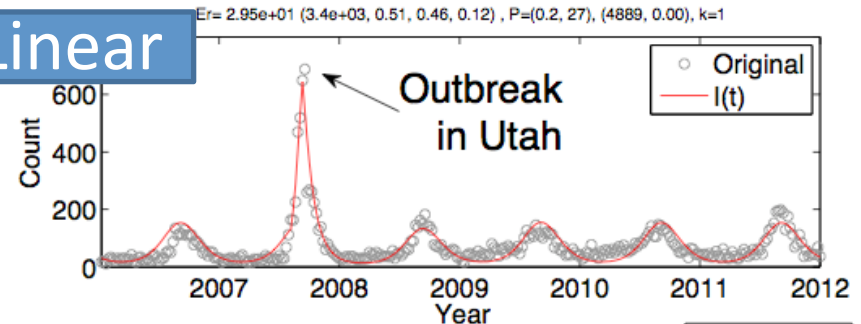
Q1. Sense-making

P4 external shock events

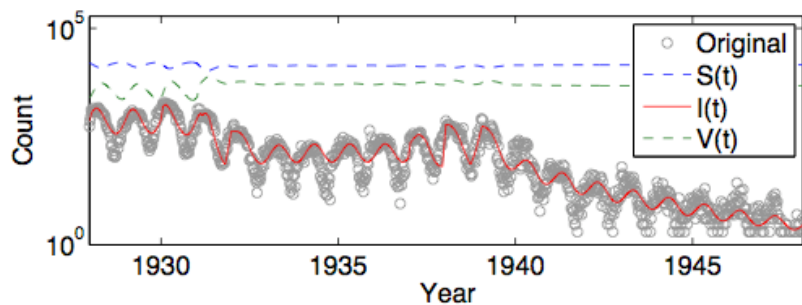
We can detect external shocks “**automatically**” !!



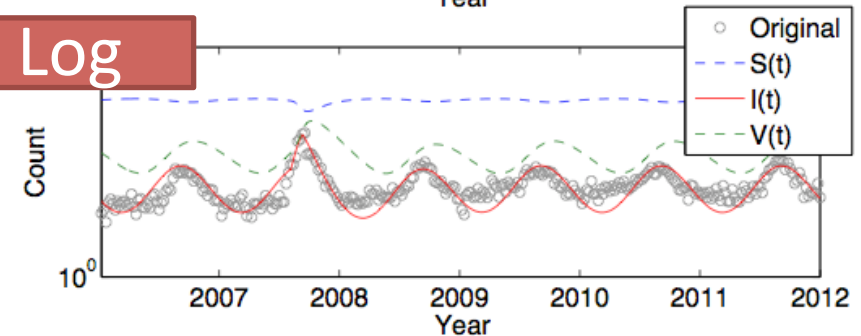
Linear



Log



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

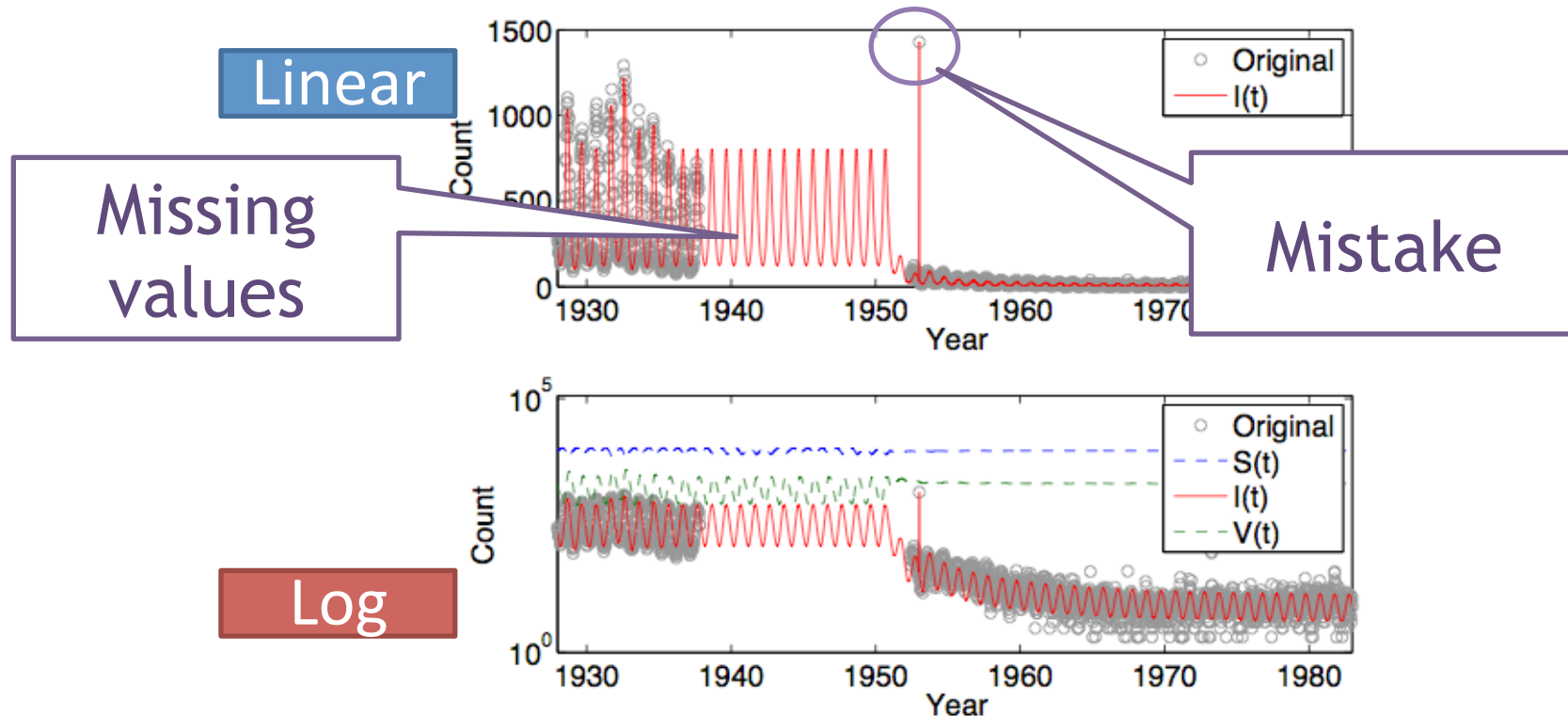


(j) Cryptosporidiosis (P1), (P3), (P4)

Q1. Sense-making

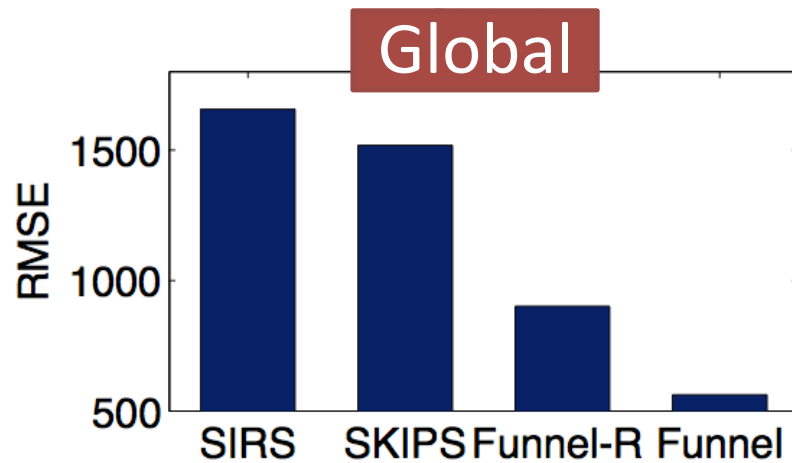
P5 mistakes, incorrect values

We can also detect typos “**automatically**” !!

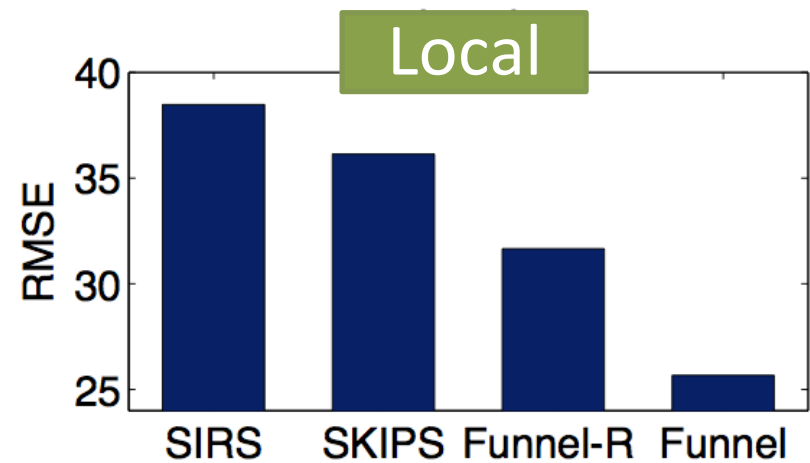


Q2. Accuracy

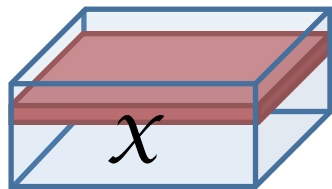
Fitting accuracy for **Global** **Local** sequences
(lower is better)



(a) Global fitting

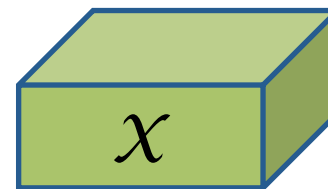


(b) Local fitting



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

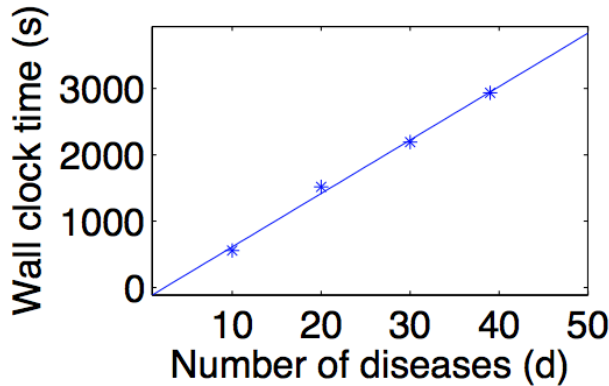
$$\bar{x}_i(t) = \sum_{j=1}^l x_{ij}(t).$$



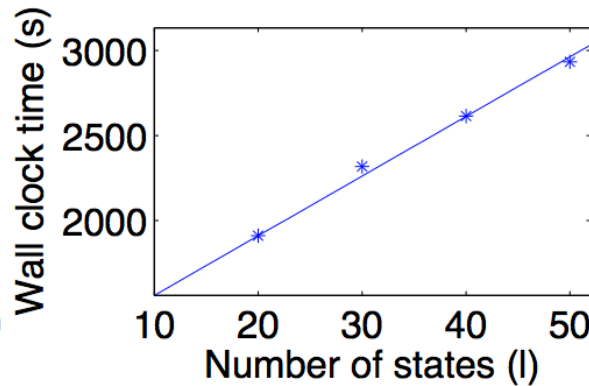
$$\{x_{ij}(t)\}_{i,j,t}^{d,l,n}$$

Q3. Scalability

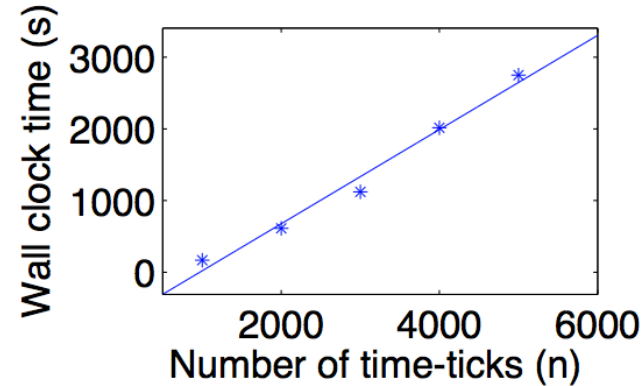
Wall clock time vs. diseases , states , Time



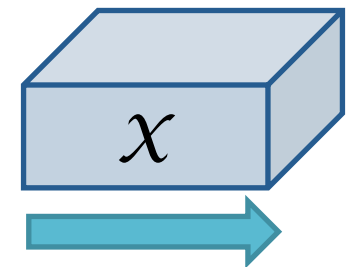
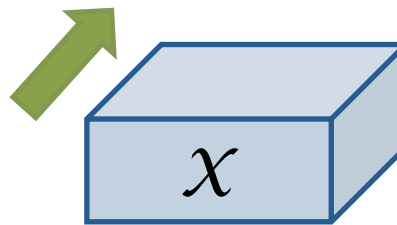
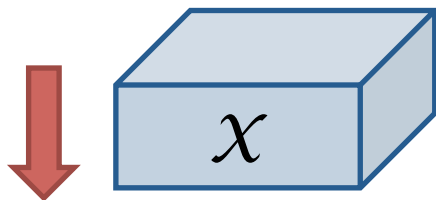
(a) Diseases (d)



(b) States (l)



(c) Duration (n)



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