

# Hierarchical models in Stan

QoCT surveys

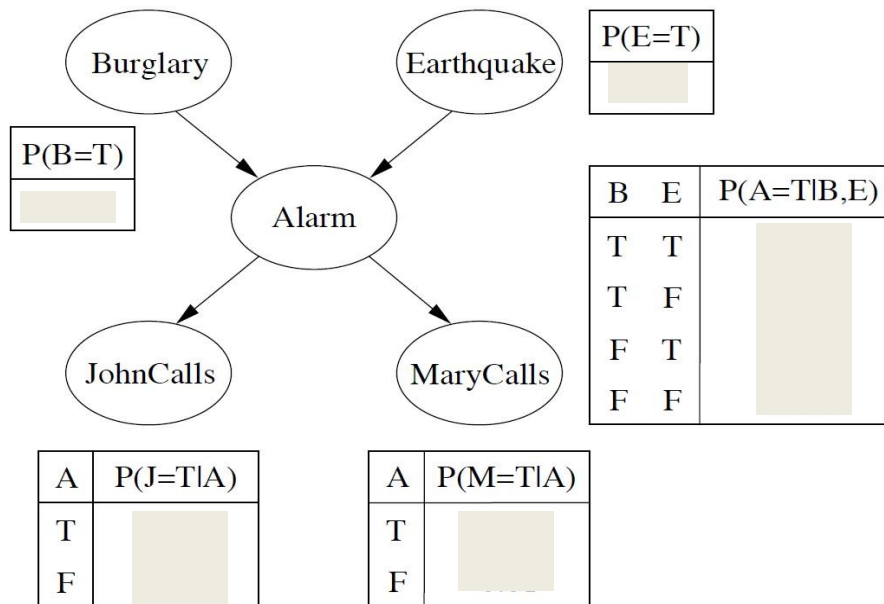
The link is at: <https://apps.eng.unimelb.edu.au/casmas/index.php?r=qoct/subjects>.

**Workshop # 11**

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# Bayesian Network: Training

- Constructing the structure of the network
  - \* domain expert to decide the causal relations
  - \* *structure learning* algorithms exist, but complicated
- Parameter learning (filling the table)

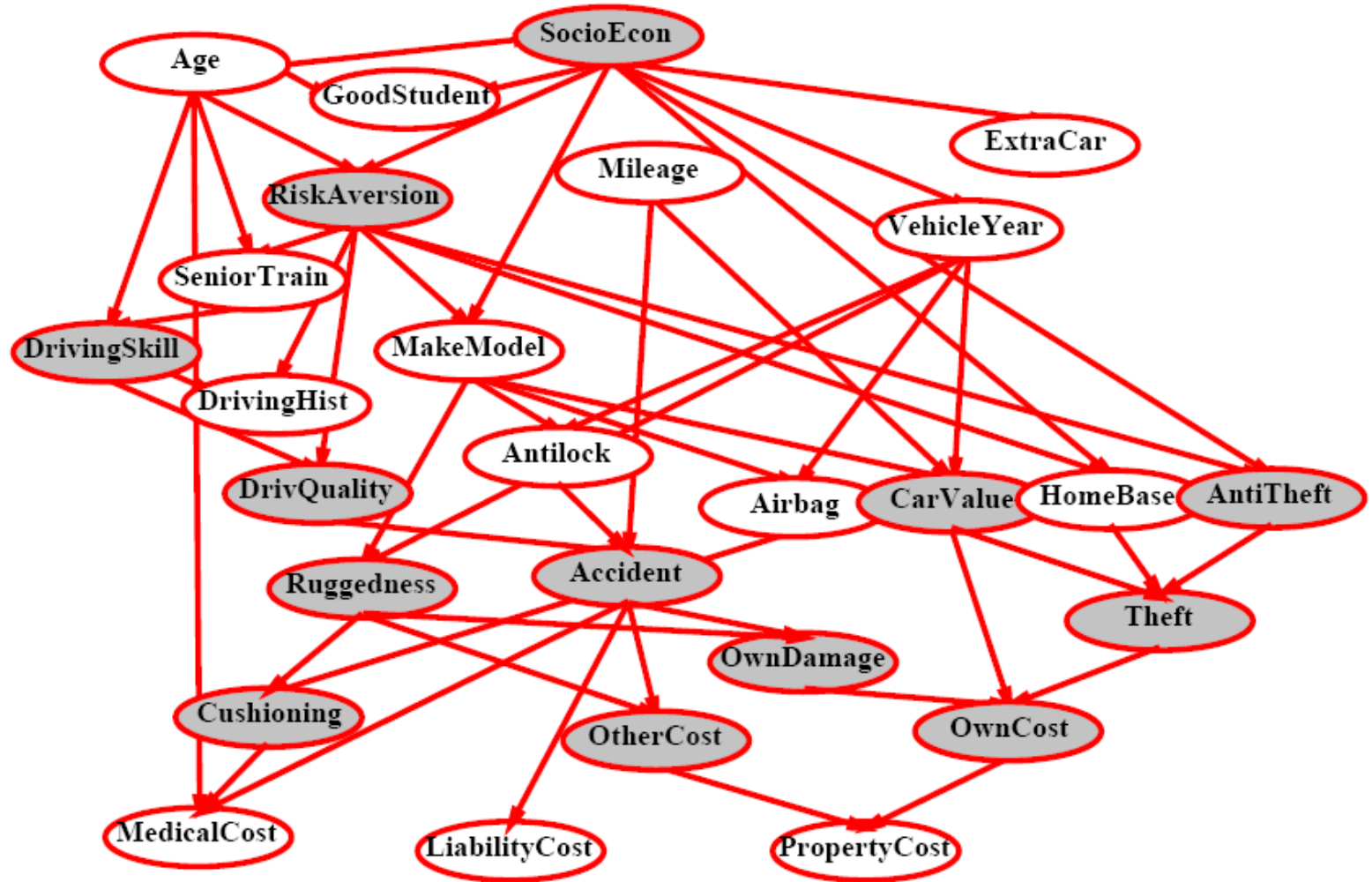


Training

| A   | B | E | J | M |
|-----|---|---|---|---|
| T   | F | T | F | T |
| F   | T | F | F | F |
| T   | F | T | F | T |
| ... |   |   |   |   |

- \* Using EM method if there are missing values

# A more realistic Bayes Network: Car insurance



in reality, parameter estimation is not an easy task !!

# Motivation for Stan

- Fit rich Bayesian statistical models
- The Process
  1. Create a statistical model
  2. Perform inference on the model
  3. Evaluate

## What is Stan?

- Statistical model specification language
- 4 cross-platform users interfaces
  - CmdStan - command line
  - RStan - R integration
  - PyStan - Python integration
  - MStan - Matlab integration (user contributed)

# The Stan Language

- Data Types
  - basic: `real`, `int`, `vector`, `row_vector`, `matrix`
  - constrained: `simplex`, `unit_vector`, `ordered`, `positive_ordered`, `corr_matrix`, `cov_matrix`
  - arrays
- Bounded variables
  - applies to `int`, `real`, and `matrix` types
  - lower example: `real<lower=0> sigma;`
  - upper example: `real<upper=100> x;`
- Program Blocks
  - `data` (optional)
  - `transformed data` (optional)
  - `parameters` (optional)
  - `transformed parameters` (optional)
  - `model`
  - `generated quantities` (optional)

# Stan Example: vectorization

```
data {  
  int<lower=0> N;  
  vector[N] y;  
  vector[N] x;  
}  
parameters {  
  real alpha;  
  real beta;  
  real<lower=0> sigma;  
}  
model {  
  alpha ~ normal(0,10);  
  beta ~ normal(0,10);  
  sigma ~ cauchy(0,5);  
  y ~ normal(alpha + beta * x, sigma);  
}
```

what is this  
model ????

RStan Demo 1

## Eight Schools: hierarchical example

- Educational Testing Service study to analyse the effects of special coaching programs on SAT-V scores in 8 high schools.
- The observed effects of special preparation are estimates based on separate analyses for the eight school experiments.
- The effects, are labelled as  $Y_j$ . Over 30 students were tested on each school.
- No prior reason to believe any program was:
  - more effective than the others
  - more similar to others

[Rubin, 1981; Gelman et al., *Bayesian Data Analysis*, 2003]

## Stan: Eight Schools Data

| School | Estimated<br>Treatment<br>Effect | Standard Error of<br>Treatment<br>Effect |
|--------|----------------------------------|--|
| A      | 28                               | 15                                       |
| B      | 8                                | 10                                       |
| C      | -3                               | 16                                       |
| D      | 7                                | 11                                       |
| E      | -1                               | 9  |
| F      | 1                                | 11                                       |
| G      | 18                               | 10                                       |
| H      | 12                               | 18                                       |



# Eight Schools: Statistical Model

- Estimate hyperparameters ????

```
data {  
  int<lower=0> J;           // # schools  
  real y[J];              // estimated treatment  
  real<lower=0> sigma[J]; // std err of effect  
}  
parameters {  
  real mu;                // school effect  
  real<lower=0> tau;      // mean for schools  
  real eta[J];           // variance between schools  
}  
transformed parameters {  
  real theta[J];  
  for (j in 1:J)  
    theta[j] <- mu + tau * eta[j];  
}  
model {  
  eta ~ normal(0, 1);  
  y ~ normal(theta, sigma);  
}
```

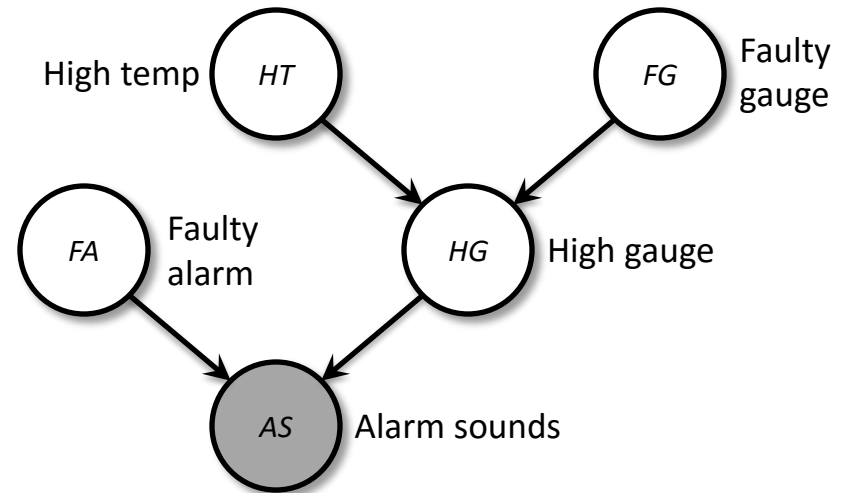
RStan Demo 2

# Example: Probability Inference

In a nuclear power station, there is an alarm that senses when a temperature gauge exceeds a given threshold.

The gauge measures the temperature of the core.

Consider the Boolean variables  $A$  (alarm sounds),  $FA$  (alarm is faulty), and  $FG$  (gauge is faulty) and the multivalued, discrete nodes  $G$  (gauge reading) and  $T$  (actual core temperature).



RStan Demo 3

What is the data block for this model ??

How many parameters ?

What about prior distribution and likelihood ?

# Reference

[1] Stan slides: <http://astrostatistics.psu.edu/su14/> (Daniel-Lee-Stan-2.pdf)

[2] Hierarchical models slides: UIUC Artificial Intelligence (CS440/ECE448)

# Stan: Help

- User Guide:  
<http://mc-stan.org/manual.html>
- Homepage:  
<http://mc-stan.org>
- Stan Users group:  
<https://groups.google.com/d/forum/stan-users>