

# Sampling Based Inference

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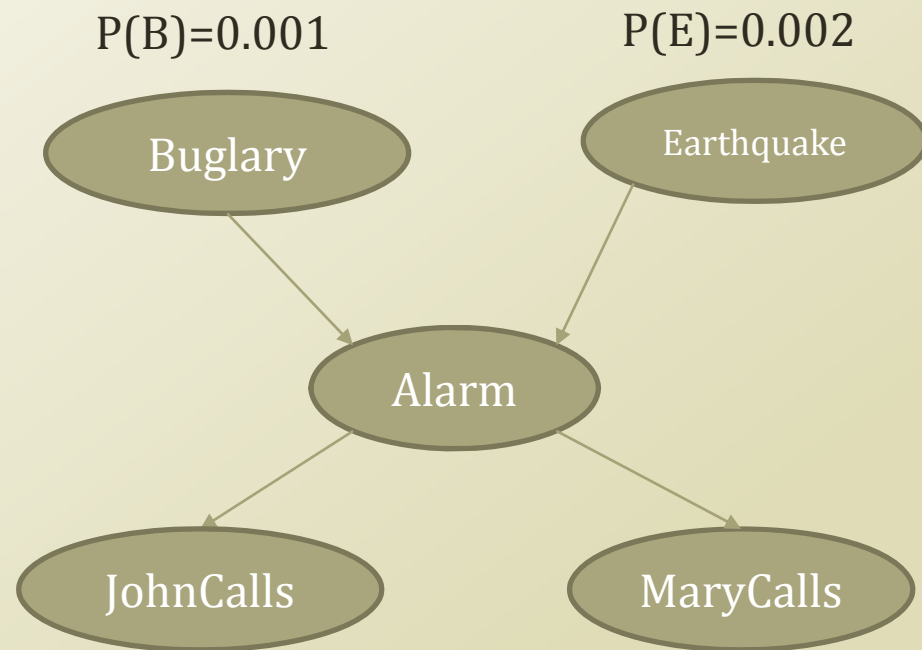
# Weekly Objectives

- Learn basic sampling methods
  - Understand the concept of Markov chain Monte Carlo
  - Able to apply MCMC to the parameter inference of Bayesian networks
  - Know the mechanism of rejection sampling
  - Know the mechanism of importance sampling
- Learn sampling based inference
  - Understand the concept of Metropolis-Hastings algorithm
  - Know the mechanism of Gibbs sampling
- Know a case study of sampling based inference
  - Understand the latent Dirichlet allocation model
  - Know the collapsed Gibbs sampling
  - Know how to derive Gibbs sampling formula for LDA

# BASIC SAMPLING METHODS

# Forward Sampling

- Generate a sample from the Bayesian network
  - Follow topological order
    - Buglary  $\rightarrow$  false
    - Earthquake  $\rightarrow$  false
    - Alarm|B=F,E=F $\rightarrow$ true
    - JC|A=T $\rightarrow$ true
    - MC|A=T $\rightarrow$ false
  - Create such sample many, many, many times
- Then, count the samples match the case
  - $P(E=T|MC=T)=?$ 
    - Count the cases of E=T and MC=T
    - Count the cases of MC=T
- Any problem?



B	E	$P(A B,E)$
T	T	0.95
T	F	0.94
F	T	0.29
F	F	0.001

A	$P(J A)$
T	0.90
F	0.05

A	$P(M A)$
T	0.70
F	0.01

# Forward Sampling in GMM

$$\begin{aligned} P(x) &= \sum_{k=1}^K P(z_k)P(x|z) \\ &= \sum_{k=1}^K \pi_k N(x|\mu_k, \Sigma_k) \end{aligned}$$

- Forward sampling of GMM
  - Sample  $z$  from  $\pi$ 
    - $z$  is the indicator of the mixture distribution
  - With selected  $z$ , sample  $x$  from  $N(\mu_z, \Sigma_z)$
- After many, many sampling, you can draw the histogram of the mixture distribution
- You have an empirical PDF, so you can ask a query like  $P(0 \leq x \leq 5 | \pi, \mu, \Sigma)$

