

COMP90051

Workshop Week 11

About the Workshops

- 7 sessions in total
 - Tue 12:00-13:00 AH211
 - Tue 12:00-13:00 AH108 *
 - Tue 13:00-14:00 AH210
 - Tue 16:15-17:15 AH109
 - Tue 17:15-18:15 AH236 *
 - Tue 18:15-19:15 AH236 *
 - Fri 14:15-15:15 AH211

About the Workshops

- Homepage

- <https://trevorcohn.github.io/comp90051-2017/workshops>

- Solutions will be released on next Friday (a week later).

Syllabus

1	Introduction; Probability theory	Probabilistic models; Parameter fitting	
2	Linear regression; Intro to regularization	Logistic regression; Basis expansion	
3	Optimization; Regularization	Perceptron	
4	Backpropagation	CNNs; Auto-encoders	
5	Hard-margin SVMs	Soft-margin SVMs	
6	Kernel methods	Ensemble Learning	
7	Clustering	EM algorithm	
8	Principal component analysis; Multidimensional Scaling	Manifold Learning; Spectral clustering	
9	Bayesian inference (uncertainty, updating)	Bayesian inference (conjugate priors)	
10	PGMs, fundamentals	PGMs, independence	←
11	Guest lecture (TBC)	PGMs, inference	
12	PGMs, statistical inference	Subject review	

Outline

- Review the lecture, background knowledge, etc.
 - Formal notations for probability
 - Joint probability for probabilistic graphical models (PGMs)
 - Directed PGMs
 - Undirected PGMs
 - Independence in PGMs
 - Directed PGMs
 - Undirected PGMs
- Worksheet

Formal notations for probability

- Upper-case letters for random variables (r.v.'s)

- A, B, C

- Lower-case letters for specific values

- a, b, c

- P for the operator to calculate the probability

- $P(A = a), P(A = a|B = b), P(A < 10), P(A = 1|B \geq 5)$

Formal notations for probability

- Suppose A is a binary random variable
- $P(A = 0)$ is a number
- $P(A = 1)$ is a number
- $P(A)$ is the distribution for the random variable A
- $P(A)$ represents $P(A = 0)$ and $P(A = 1)$

$P(A, B)$ and $P(A = a, B = b)$

□ Suppose A and B are both binary random variables

□ $P(A, B) = P(A)P(B)$ is equivalent to ...

□ $P(A = a, B = b) = P(A = a)P(B = b) \quad \forall a, b \in \{0, 1\}$

□ $P(A = 0, B = 0) = P(A = 0)P(B = 0)$

□ $P(A = 1, B = 0) = P(A = 1)P(B = 0)$

□ $P(A = 0, B = 1) = P(A = 0)P(B = 1)$

□ $P(A = 1, B = 1) = P(A = 1)P(B = 1)$

Sometimes we write partially $P(A = 1, B)$

□ Given the probability table:

$P(A, B)$	$B = 0$	$B = 1$
$A = 0$	0.1	0.2
$A = 1$	0.3	0.4

□ $P(A = 1, B)$ can be

□ interpreted as an unnormalized distribution for B

$P(A = 1, B)$ is unnormalized $P(B|A = 1)$

□ Given the probability table:

$P(A, B)$	$B = 0$	$B = 1$
$A = 0$	0.1	0.2
$A = 1$	0.3	0.4

□ $P(A = 1, B)$ can be

□ interpreted as an unnormalized distribution for B ($A = 1$)

□ $P(B|A = 1) \propto P(A = 1, B)$

□ $P(A = 1, B = 0) = 0.3 \quad \rightarrow \quad P(B = 0|A = 1) = 3/7$

□ $P(A = 1, B = 1) = 0.4 \quad \rightarrow \quad P(B = 1|A = 1) = 4/7$

$P(A)$

□ Given the probability table:

$P(A, B)$	$B = 0$	$B = 1$
$A = 0$	0.1	0.2
$A = 1$	0.3	0.4

□ $P(A)$ is

□ the marginal distribution for A

$$\square P(A) = \sum_B P(A, B)$$

$$\begin{aligned} \square P(A = 0) &= \sum_B P(A = 0, B) = \sum_{b \in \{0,1\}} P(A = 0, B = b) \\ &= P(A = 0, B = 0) + P(A = 0, B = 1) = 0.3 \end{aligned}$$

$$\square P(A = 1) = 1 - P(A = 0) = 0.7$$

What about $P(A|B)$?

□ Given the probability table:

$P(A, B)$	$B = 0$	$B = 1$
$A = 0$	0.1	0.2
$A = 1$	0.3	0.4

□ $P(A|B)$ can be

□ interpreted as two distributions for A given $B = 0$ and 1

□ $P(A|B = 0)$ is a distribution

□ $P(A = 0|B = 0) = 1/4, P(A = 1|B = 0) = 3/4$

□ $P(A|B = 1)$ is another one

□ $P(A = 0|B = 1) = 1/3, P(A = 1|B = 1) = 2/3$

And $P(A = 1|B) \dots$?

□ Given the probability table:

$P(A, B)$	$B = 0$	$B = 1$
$A = 0$	0.1	0.2
$A = 1$	0.3	0.4

□ $P(A = 1|B)$ is

□ the likelihood of observing $A = 1$ under different values of B

□ We have calculated that

□ $P(A = 0|B = 0) = 1/4$, $P(A = 1|B = 0) = 3/4$

□ $P(A = 0|B = 1) = 1/3$, $P(A = 1|B = 1) = 2/3$

If we know the joint distribution $P(A, B)$

- We can calculate everything, such as
- Marginal distributions
 - $P(A)$ and $P(B)$
 - Summation or integration over other random variables
 - $P(A) = \sum_B P(A, B) = \sum_{b \in \{0,1\}} P(A, B = b)$
- Conditional distributions
 - $P(A|B)$ and $P(B|A)$
 - Division of two unconditional distributions
 - $P(A|B) = P(A, B)/P(B)$

In general, if we know $P(A, B, C, D, E)$

$$\square P(A) = \sum_{B,C,D,E} P(A, B, C, D, E)$$

$$\square P(A, B) = \sum_{C,D,E} P(A, B, C, D, E)$$

$$\square P(A, B, C) = \sum_{D,E} P(A, B, C, D, E)$$

$$\square P(A, B, C, D) = \sum_E P(A, B, C, D, E)$$

$$\square P(B|A) = P(A, B)/P(A)$$

$$\square P(B, C|A) = P(A, B, C)/P(A)$$

$$\square P(C, D|A, B) = P(A, B, C, D)/P(A, B)$$

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 - ❑ Undirected PGMs
- ❑ Worksheet

How to calculate the joint distribution?

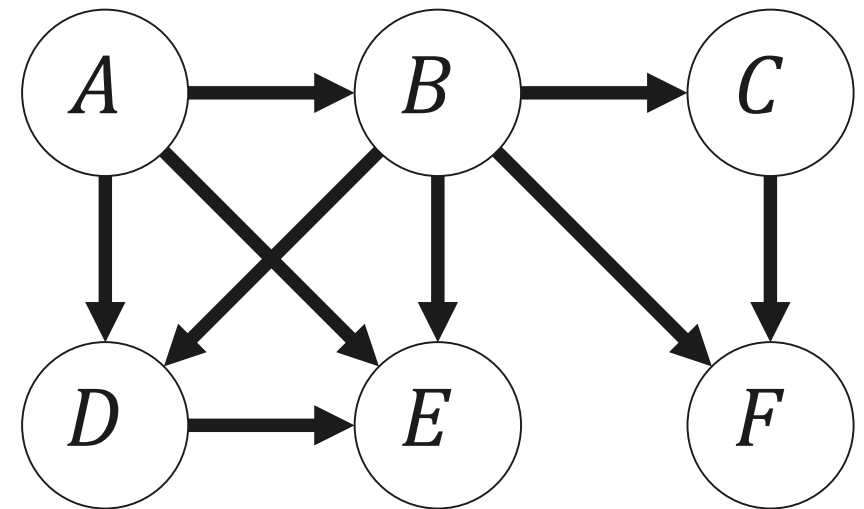
□ Directed PGMs

$$P(\text{all } r.v.) = \prod_{\text{every } r.v.} P(r.v. | \text{parents of } r.v.)$$

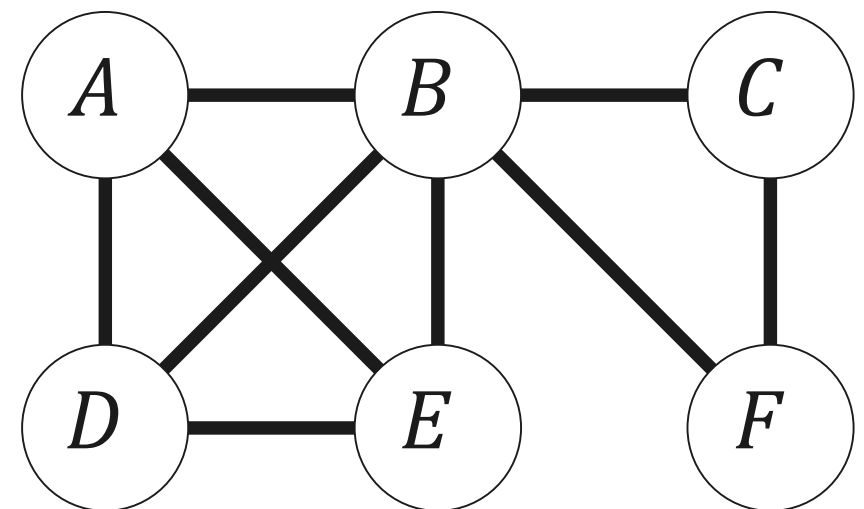
□ Undirected PGMs

$$P(\text{all } r.v.) \propto \prod_{\text{every clique}} f_{\text{clique}}(r.v. \text{ in } \textit{clique})$$

$$\begin{aligned}
 &P(A, B, C, D, E, F) \\
 &= P(A)P(B|A)P(C|B) \\
 &\quad \cdot P(D|A, B)P(E|A, B, D)P(F|B, C)
 \end{aligned}$$



$$\begin{aligned}
 &P(A, B, C, D, E, F) \\
 &\propto f_1(A, B, D, E)f_2(B, C, F) \\
 &= \frac{1}{Z} f_1(A, B, D, E)f_2(B, C, F) \\
 &Z = \sum_{A, B, C, D, E, F} f_1(A, B, D, E)f_2(B, C, F)
 \end{aligned}$$



Outline

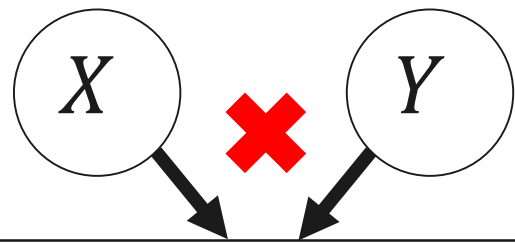
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Independence in directed PGMs

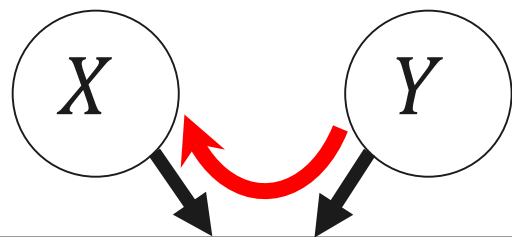
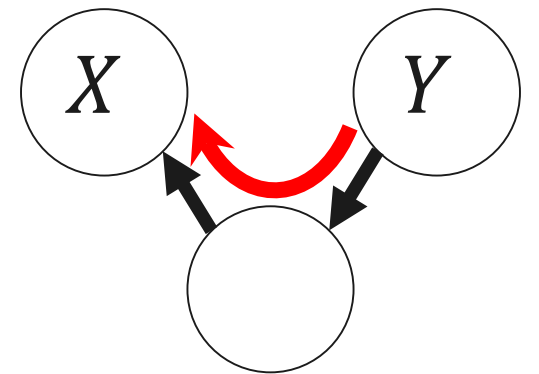
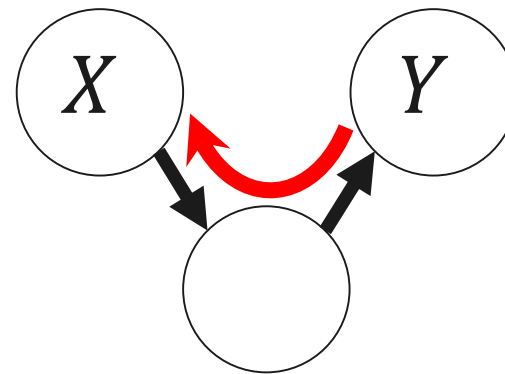
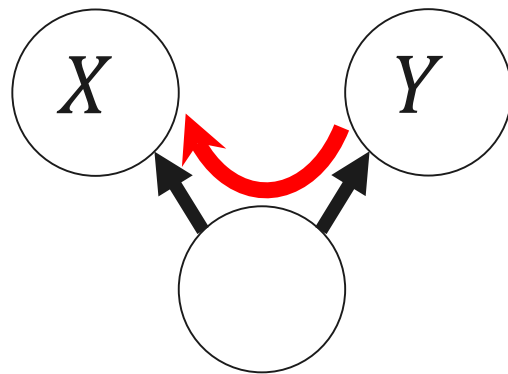
□ Paths from Y to X

□ If a path exists, then X and Y are dependent*

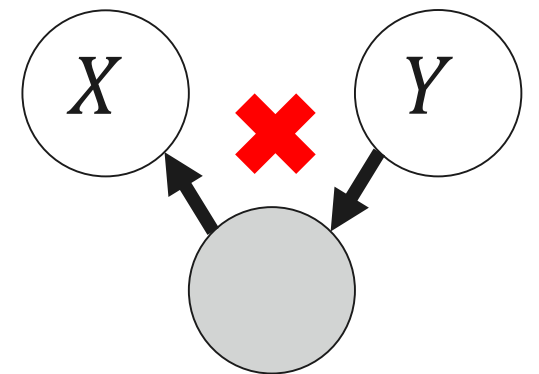
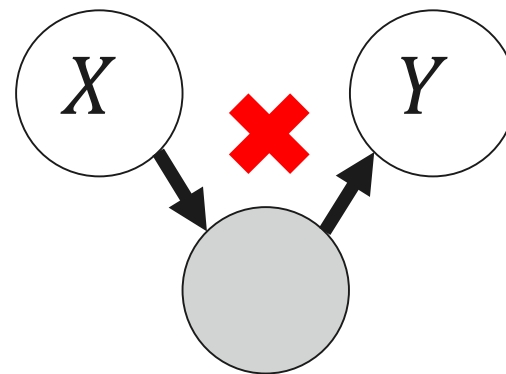
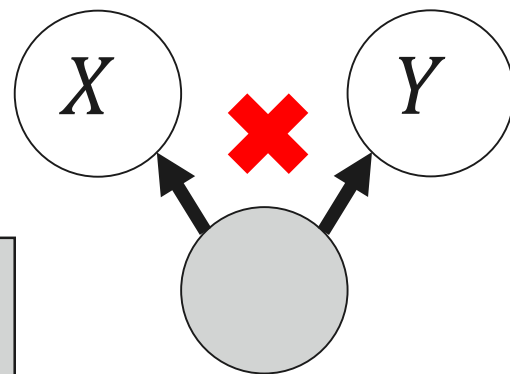
□ * the PGM doesn't require they should be independent



every descendant
is not observed

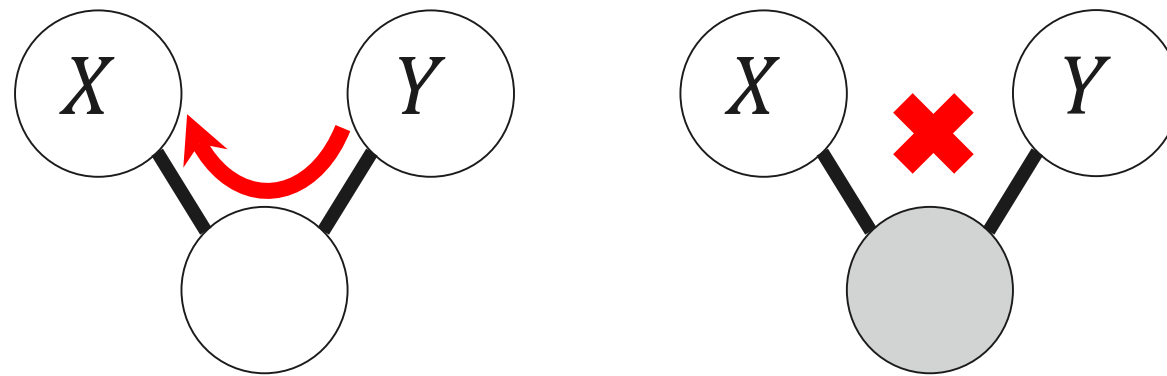


any descendant
is observed



Independence in undirected PGMs

- Paths from Y to X
- If the path exists, then X and Y are dependent



A practice for independence in PGMs

□ <http://web.mit.edu/jmn/www/6.034/d-separation.pdf>

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