

COMP90051

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# Workshop Week 11

# About the Workshops

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- 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

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- 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

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- ❑ 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

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- 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

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

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□ 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 $P(A = 1, B)$

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□ 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 $P(B|A = 1)$

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□ 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  $B$  ( $A = 1$ )

□  $P(B|A = 1) \propto P(A = 1, B)$

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

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

# $P(A)$

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□ 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 $P(A|B)$ ?

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□ 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 $P(A = 1|B) \dots$ ?

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□ 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 $P(A, B)$

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- 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 $P(A, B, C, D, E)$

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

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

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

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

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

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

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

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# How to calculate the joint distribution?

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## □ Directed PGMs

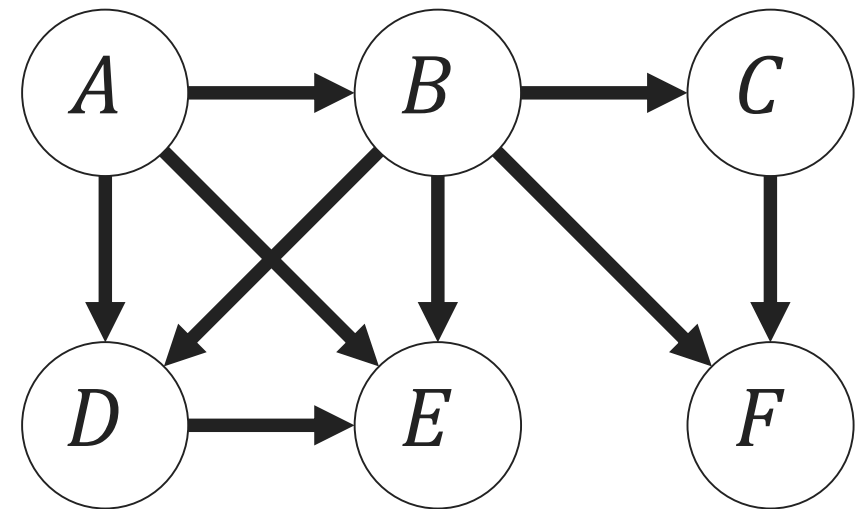
$$P(\text{all } r.v.) = \prod_{\text{every } r.v.} P(r.v. | \text{parents of } r.v.)$$

## □ Undirected PGMs

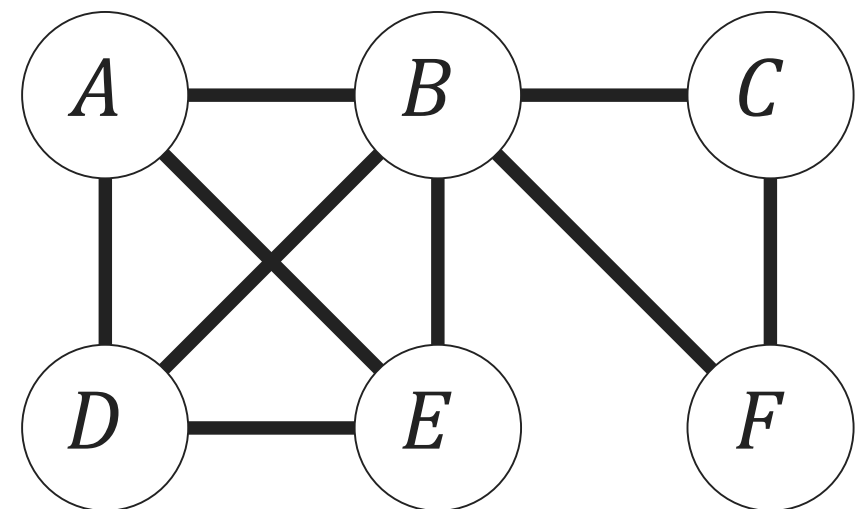
$$P(\text{all } r.v.) \propto \prod_{\text{every clique}} f_{\text{clique}}(r.v. \text{ in } \textit{clique})$$

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



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



# Outline

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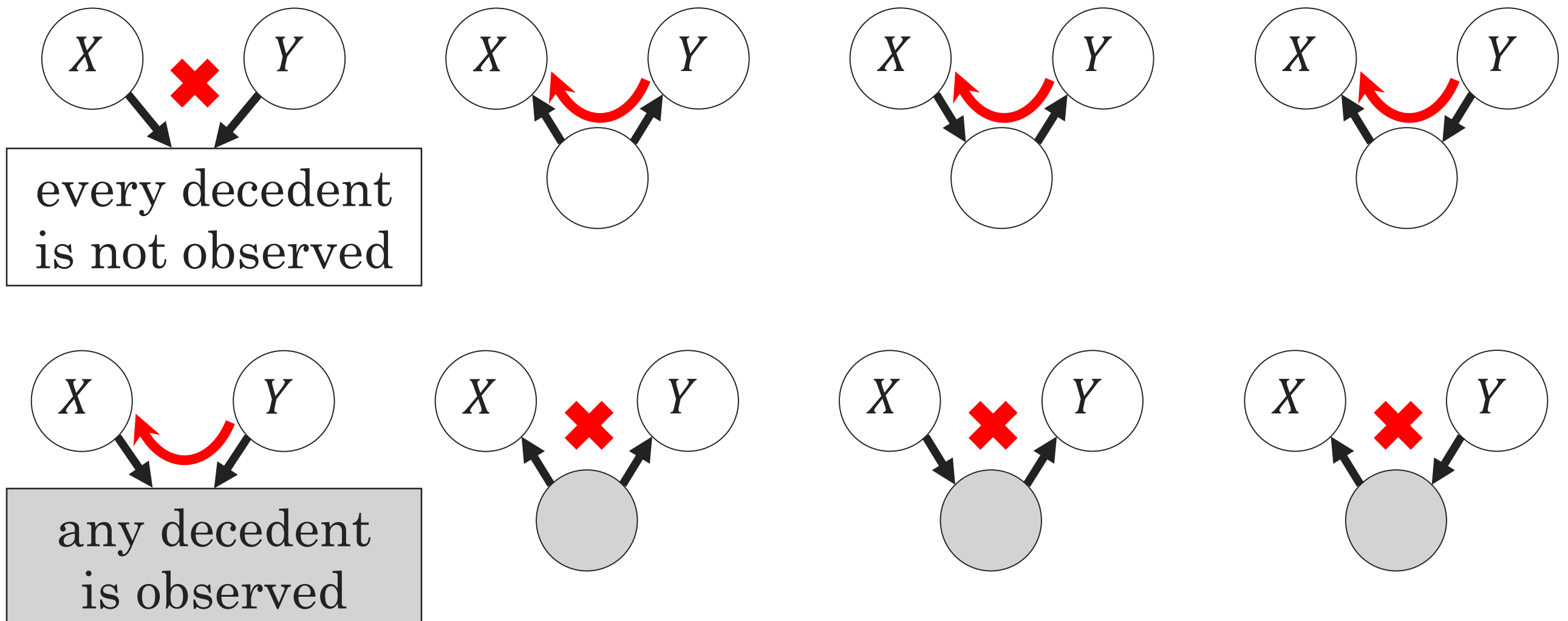
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# Independence in directed PGMs

□ Paths from  $Y$  to  $X$

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

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

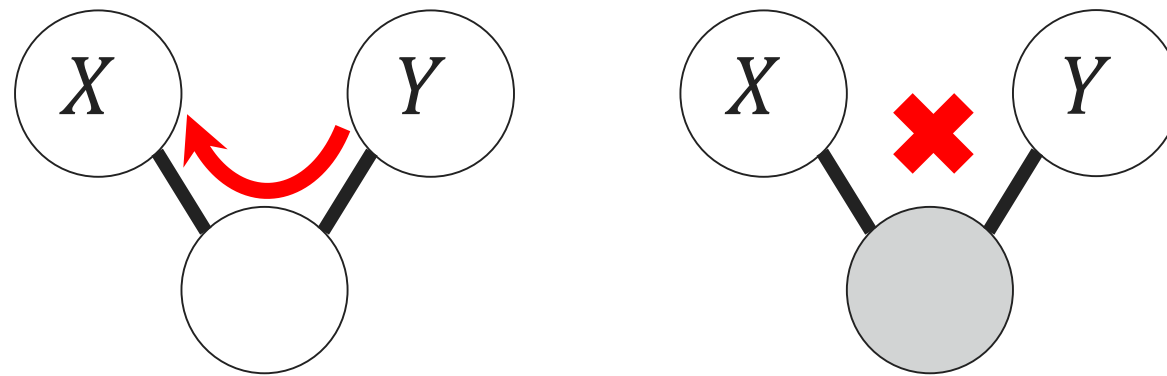


# Independence in undirected PGMs

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□ Paths from  $Y$  to  $X$

□ If the path exists, then  $X$  and  $Y$  are dependent



# A practice for independence in PGMs

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

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