

# Section 6

Sunday, December 9, 2018 8:42 PM



section6

## CS188 Fall 2018 Section 6: Probability + Bayes' Nets

### 1 Probability

Use the probability table to calculate the following values:

$X_1$	$X_2$	$X_3$	$P(X_1, X_2, X_3)$
0	0	0	0.05
1	0	0	0.1
0	1	0	0.4
1	1	0	0.1
0	0	1	0.1
1	0	1	0.05
0	1	1	0.2
1	1	1	0.0

1.  $P(X_1 = 1, X_2 = 0)$

$0.1 + 0.05 = 0.15$

$X_1$	$X_2$	$X_3$	$P(X_1, X_2, X_3)$
0	0	0	0.05
1	0	0	0.1
0	1	0	0.4
1	1	0	0.1
0	0	1	0.1
1	0	1	0.05
0	1	1	0.2
1	1	1	0.0

2.  $P(X_3 = 0)$

$0.05 + 0.1 + 0.4 + 0.1 = 0.65$

$X_1$	$X_2$	$X_3$	$P(X_1, X_2, X_3)$
0	0	0	0.05
1	0	0	0.1
0	1	0	0.4
1	1	0	0.1
0	0	1	0.1
1	0	1	0.05
0	1	1	0.2
1	1	1	0.0

3.  $P(X_2 = 1 | X_3 = 1)$

$(0.2 + 0) / (0.2 + 0.05 + 0.1) = 0.5714$

$X_1$	$X_2$	$X_3$	$P(X_1, X_2, X_3)$
0	0	0	0.05
1	0	0	0.1
0	1	0	0.4
1	1	0	0.1
0	0	1	0.1
1	0	1	0.05
0	1	1	0.2
1	1	1	0.0

4.  $P(X_1 = 0 | X_2 = 1, X_3 = 1)$

$0.2 / 0.2 = 1$

$X_1$	$X_2$	$X_3$	$P(X_1, X_2, X_3)$
0	0	0	0.05
1	0	0	0.1
0	1	0	0.4
1	1	0	0.1
0	0	1	0.1
1	0	1	0.05
0	1	1	0.2
1	1	1	0.0

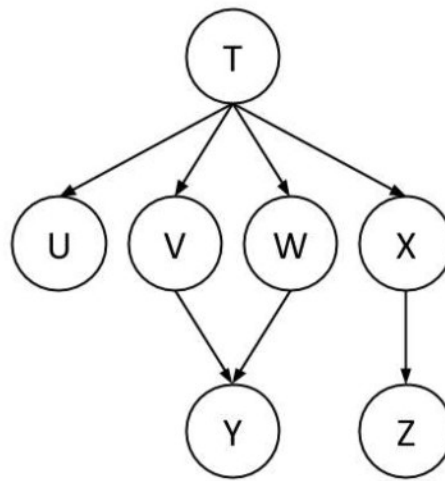
5.  $P(X_1 = 0, X_2 = 1 | X_3 = 1)$

$0.2 / (0.2 + 0.05 + 0.1) = 0.5714$

$X_1$	$X_2$	$X_3$	$P(X_1, X_2, X_3)$
0	0	0	0.05
1	0	0	0.1
0	1	0	0.4
1	1	0	0.1
0	0	1	0.1
1	0	1	0.05
0	1	1	0.2
1	1	1	0.0

## 2 D-Separation

Indicate whether each of the following conditional independence relationships is guaranteed to be true in the Bayes Net below. If the independence relationship does not hold, identify all active (d-connected) paths in the graph.



1.  $U \perp\!\!\!\perp X$  two effects from a common cause. If you see someone with an umbrella, you might expect traffic
2.  $U \perp\!\!\!\perp X|T$
3.  $V \perp\!\!\!\perp W|Y$  two causes GIVEN common effect. OJ wife is dead. OJ Killed him? But then you learn the wife had a spiteful ex-lover!!!! this is new information in the investigation!
4.  $V \perp\!\!\!\perp W|T$  -two effects given their common cause. no longer have a relationship (Markovian)
5.  $T \perp\!\!\!\perp Y|V$  markov chain broken with v. but alternative causality path thru W means T and Y are not independent even in this universe
6.  $Y \perp\!\!\!\perp Z|W$
7.  $Y \perp\!\!\!\perp Z|T$

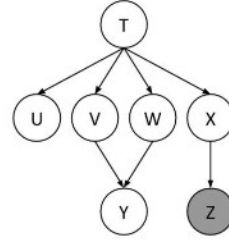
## 3 Variable Elimination

Using the same Bayes Net (shown below), we want to compute  $P(Y|+z)$ . All variables have binary domains. Assume we run variable elimination to compute the answer to this query, with the following variable elimination ordering:  $X, T, U, V, W$ . **VE on HV's. my DAWGGGG**

Complete the following description of the factors generated in this process:

After inserting evidence, we have the following factors to start out with:

$$P(T), P(U|T), P(V|T), P(W|T), P(X|T), P(Y|V, W), P(+z|X)$$



- (a) When eliminating  $X$  we generate a new factor  $f_1$  as follows, which leaves us with the factors:

$$f_1(+z|T) = \sum_x P(x|T)P(+z|x) \quad P(T), P(U|T), P(V|T), P(W|T), P(Y|V, W), f_1(+z|T)$$

- (b) When eliminating  $T$  we generate a new factor  $f_2$  as follows, which leaves us with the factors:

$$f_2(U, V, W, z) = \sum_t P(U|T)P(V|T)P(W|T)f_1(z|T) \quad P(Y|V, W) f_2(U, V, W, z)$$

- (c) When eliminating  $U$  we generate a new factor  $f_3$  as follows, which leaves us with the factors:

$$f_3(V, W, z) = \sum_u f_2(U, V, W, z) \quad P(Y|V, W) f_3(V, W, z)$$

- (d) When eliminating  $V$  we generate a new factor  $f_4$  as follows, which leaves us with the factors:

$$f_4(Y, W, z) = \sum_v P(Y|V, W)f_3(V, W, z) \quad f_4(Y, W, z)$$

- (e) When eliminating  $W$  we generate a new factor  $f_5$  as follows, which leaves us with the factors:

$$f_5(Y, z) = \sum_w f_4(W, z) \quad f_5(z, Y)$$

- (f) How would you obtain  $P(Y | +z)$  from the factors left above:

$$P(Y|z) = \alpha \sum_v P(Y|V, W) \sum_u \sum_t P(U|T)P(V|T)P(W|T)P(T) \sum_x P(x|T)P(z|x)$$

- (g) What is the size of the largest factor that gets generated during the above process?

$f_2$  has 3 unassigned variables, requires 8 entries to specify

- (m) Does there exist a better elimination ordering (one which generates smaller largest factors)?