#### Announcements

#### Assignments:

- HW8 (written)
  - Due tonight, 10 pm
- P4
  - Due Thu 10/31, 10 pm
  - Check your slip days!

#### Midterm:

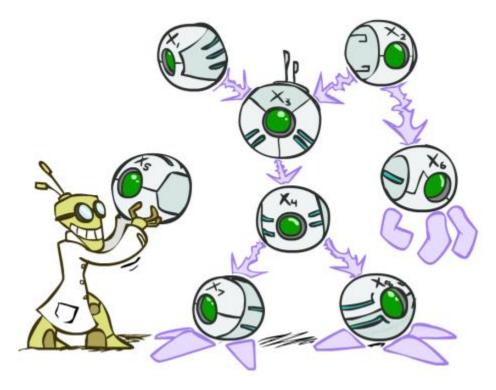
■ Tue 11/12, in-class

#### Course Feedback:

Results

## AI: Representation and Problem Solving

## **Bayes Nets**



Instructors: Pat Virtue & Fei Fang

Slide credits: CMU AI and http://ai.berkeley.edu

## **Probability Notation**

Notation and conventions in this course

$$P(+b,C) = \sum_{a \in \{a_1,a_2,a_3\}} P(a,+b,C)$$

## **Probability Notation**

#### Notation and conventions in this course

$$P(+b,C) = \sum_{a \in \{a_1,a_2,a_3\}} P(a,+b,C)$$

- Random variables:
  - Capitalized
  - Represents all potential outcomes
  - e.g. *C*
- Outcomes (values):
  - lower case
  - e.g. +b,  $a_1$ ,  $a_2$ ,  $a_3$
- Variables for values:
  - lower case
  - E.g. *a*

- For each random variable
  - Discrete outcomes
  - Disjoint outcomes
  - Not always binary

Which of the following probability tables sum to one? Select all that apply.

```
i. P(A | b)
ii. P(A,B,c)
iii. P(A,B | c)
iv. P(a,b | c)
v. P(a | B,C)
vi. P(c | A)
```

Which of the following probability tables sum to one? Select all that apply.

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iv. P(a,b | c)
v. P(a | B,C)
vi. P(c | A)
```

#### How many valid equations can we compose using:

$$P(x)$$
,  $P(y)$ ,  $P(x,y)$ ,  $P(x|y)$ ,  $P(y|x)$  and  $=$ ,  $\times$ ,  $\div$ 

First one: P(x|y) = P(x,y)/P(y)

- A) 2
- B) 4
- C) 7
- D) Other

At most one use per probability term

e.g. Not 
$$P(x) = P(x)$$

Must be different

e.g. Cannot also use

$$P(x,y)/P(y) = P(x|y)$$

Also (less meaningful): 
$$P(Y) = P(Y|X)P(X) / P(X|Y)$$
  

$$P(X) = P(X|Y)P(Y) / P(Y|X)$$

How many valid equations can we compose using:

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,  $P(Y)$ ,  $P(X,Y)$ ,  $P(X|Y)$ ,  $P(Y|X)$  and  $=$ ,  $\times$ ,  $\div$ 

First one: P(X|Y) = P(X,Y)/P(Y)

A) 2
$$P(Y|X) = P(X,Y)/P(X)$$
B) 4
$$P(X,Y) = P(X|Y)P(Y)$$
C) 7
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D) Other
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At most one use per probability term e.g. Not P(X) = P(X)

Must be different e.g. Cannot also use P(X,Y)/P(Y) = P(X|Y)

## **Probability Tools Summary**

#### Our toolbox

1. Definition of conditional probability

$$P(A|B) = \frac{P(A,B)}{P(B)}$$

2. Product Rule

$$P(A,B) = P(A|B)P(B)$$

3. Bayes' theorem

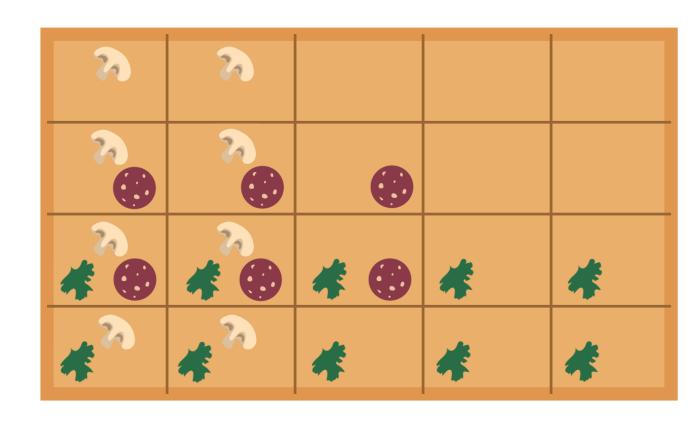
$$P(B|A) = \frac{P(A|B)P(B)}{P(A)}$$

4. Chain Rule

$$P(X_1, ..., X_N) = \prod_{n=1}^{N} P(X_n \mid X_1, ..., X_{n-1})$$

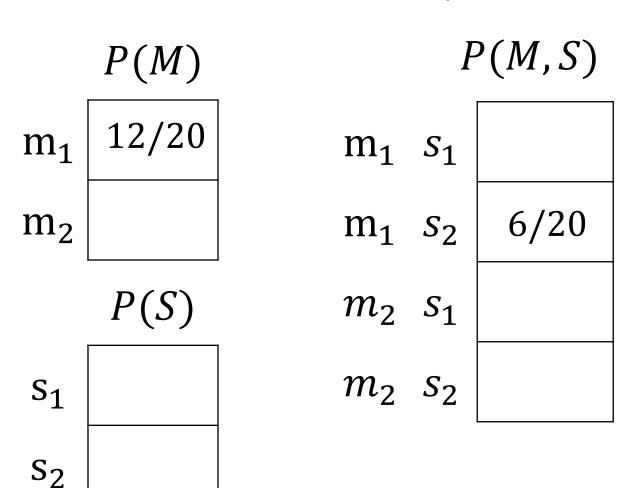
#### What is the probability of getting a slice with:

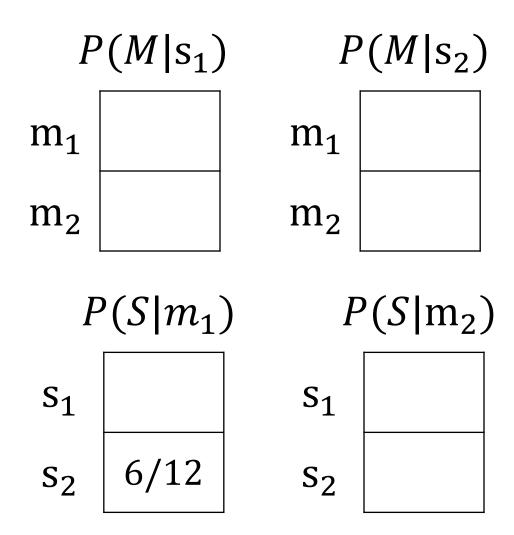
- 1) No mushrooms
- 2) Spinach and no mushrooms
- 3) Spinach, when asking for slice with no mushrooms
- Mushrooms
- Spinach
- No spinach
- No spinach and mushrooms
- No spinach when asking for no mushrooms
- No spinach when asking for mushrooms
- Spinach when asking for mushrooms



Icons: CC, https://openclipart.org/detail/296791/pizza-slice

#### You can answer all of these questions:





P(Weather)?

P(Weather | winter)?

P(Weather | winter, hot)?

Season	Temp	Weather	P(S, T, W)
summer	hot	sun	0.30
summer	hot	rain	0.05
summer	cold	sun	0.10
summer	cold	rain	0.05
winter	hot	sun	0.10
winter	hot	rain	0.05
winter	cold	sun	0.15
winter	cold	rain	0.20

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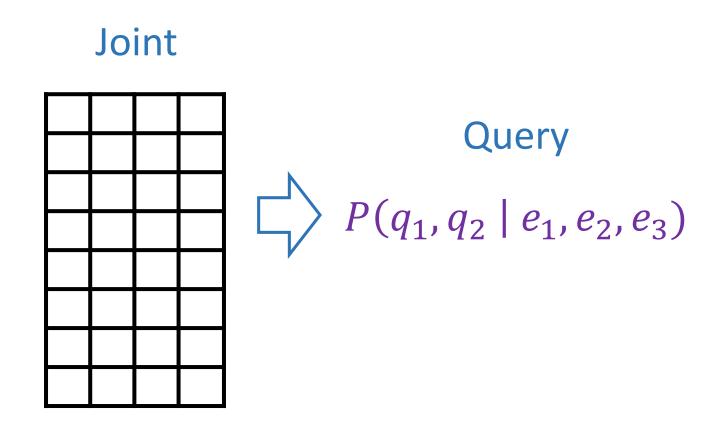
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Joint distributions are the best!



#### Two tools to go from joint to query

1. Definition of conditional probability

$$P(A|B) = \frac{P(A,B)}{P(B)}$$

2. Law of total probability (marginalization, summing out)

$$P(A) = \sum_{b} P(A, b)$$

$$P(Y \mid U, V) = \sum_{x} \sum_{z} P(x, Y, z \mid U, V)$$

Two tools to go from joint to query

Joint:  $P(H_1, H_2, Q, E)$ 

Query:  $P(Q \mid e)$ 

1. Definition of conditional probability

$$P(Q|e) = \frac{P(Q,e)}{P(e)}$$

2. Law of total probability (marginalization, summing out)

$$P(Q,e) = \sum_{h_1} \sum_{h_2} P(h_1, h_2, Q, e)$$

$$P(e) = \sum_{q} \sum_{h_1} \sum_{h_2} P(h_1, h_2, q, e)$$

P(Weather)?

P(Weather | winter)?

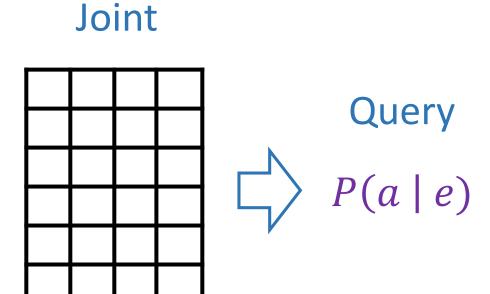
P(Weather | winter, hot)?

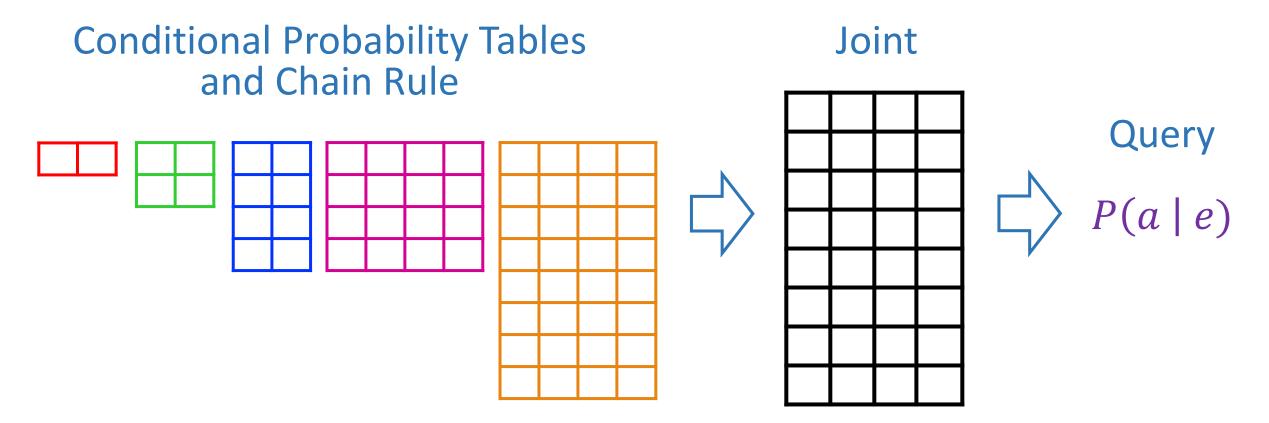
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winter	cold	sun	0.15
winter	cold	rain	0.20

Joint distributions are the best!

#### Problems with joints

- We aren't given the joint table
  - Usually some set of conditional probability tables





P(A) P(B|A) P(C|A,B) P(D|A,B,C) P(E|A,B,C,D)

#### Two tools to construct joint distribution

#### 1. Product rule

$$P(A,B) = P(A \mid B)P(B)$$
  
 
$$P(A,B) = P(B \mid A)P(A)$$

#### 2. Chain rule

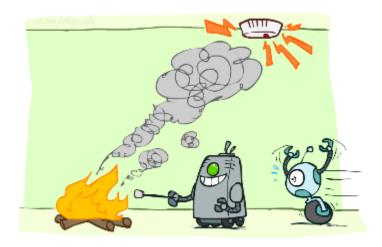
$$P(X_1,X_2,\ldots,X_n)=\prod_i P(X_i\mid X_1,\ldots,X_{i-1})$$
 
$$P(A,B,C)=P(A)P(B\mid A)P(C\mid A,B)\quad\text{for ordering A, B, C}$$

$$P(A,B,C) = P(A)P(C \mid A)P(B \mid A,C)$$
 for ordering A, C, B

$$P(A,B,C) = P(C)P(B \mid C)P(A \mid C,B)$$
 for ordering C, B, A

#### Binary random variables

- Fire
- Smoke
- Alarm



#### **Variables**

■ B: Burglary

■ A: Alarm goes off

■ M: Mary calls

■ J: John calls

■ E: Earthquake!

#### How many different ways can we write the chain rule?

*A.* 1

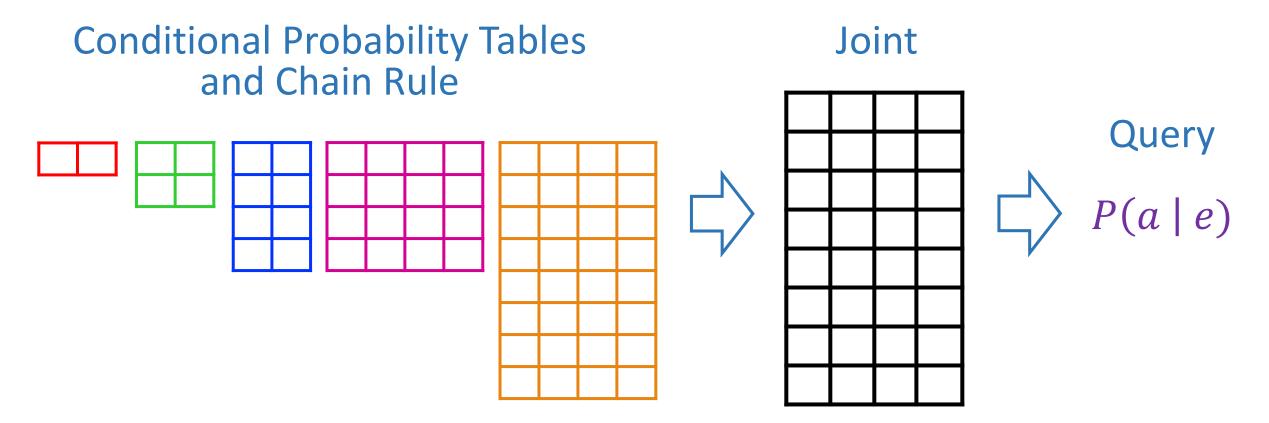
*B.* 5

C. 5 choose 5

*D.* 5!

 $E. 5^5$ 





P(A) P(B|A) P(C|A,B) P(D|A,B,C) P(E|A,B,C,D)

### Answer Any Query from Condition Probability Tables

Process to go from (specific) conditional probability tables to query

- 1. Construct the joint distribution
  - 1. Product Rule or Chain Rule
- 2. Answer query from joint
  - 1. Definition of conditional probability
  - 2. Law of total probability (marginalization, summing out)

### Answer Any Query from Condition Probability Tables

#### Bayes' rule as an example

Given: 
$$P(E|Q)$$
,  $P(Q)$  Query:  $P(Q|e)$ 

- 1. Construct the joint distribution
  - 1. Product Rule or Chain Rule

$$P(E,Q) = P(E|Q)P(Q)$$

- 2. Answer query from joint
  - 1. Definition of conditional probability

$$P(Q \mid e) = \frac{P(e,Q)}{P(e)}$$

2. Law of total probability (marginalization, summing out)

$$P(Q \mid e) = \frac{P(e,Q)}{\sum_{q} P(e,q)}$$

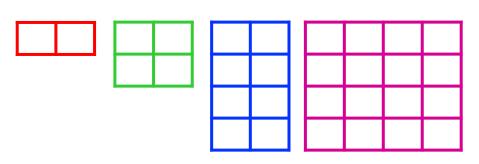
#### Bayesian Networks

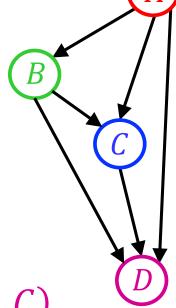
#### Bayes net

One node per random variable

DAG

One CPT per node: P(node | Parents(node) )





$$P(A,B,C,D) = P(A) P(B|A) P(C|A,B) P(D|A,B,C)$$

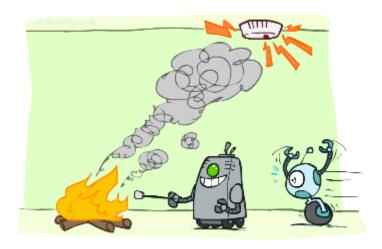
Encode joint distributions as product of conditional distributions on each variable

$$P(X_1, ..., X_N) = \prod_i P(X_i | Parents(X_i))$$

## Build Bayes Net Using Chain Rule

#### Binary random variables

- Fire
- Smoke
- Alarm



#### Question

#### Variables

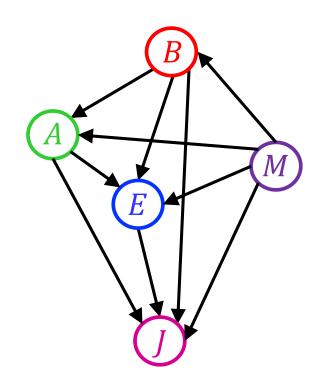
■ B: Burglary

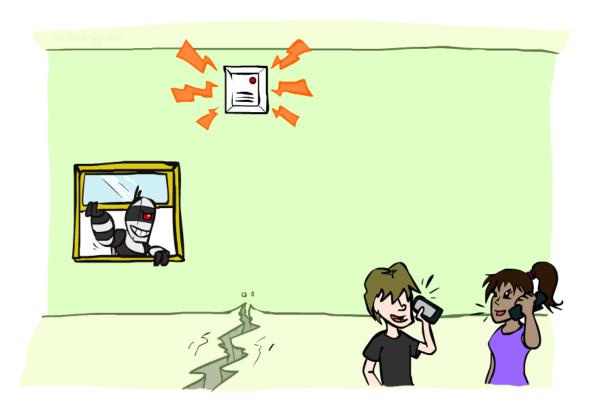
■ A: Alarm goes off

■ M: Mary calls

■ J: John calls

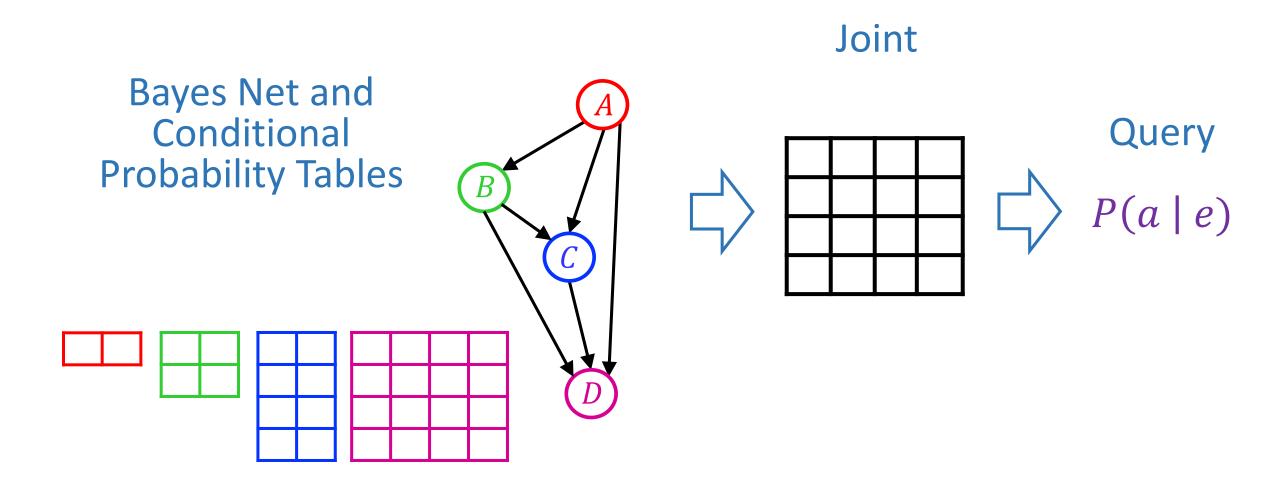
■ E: Earthquake!



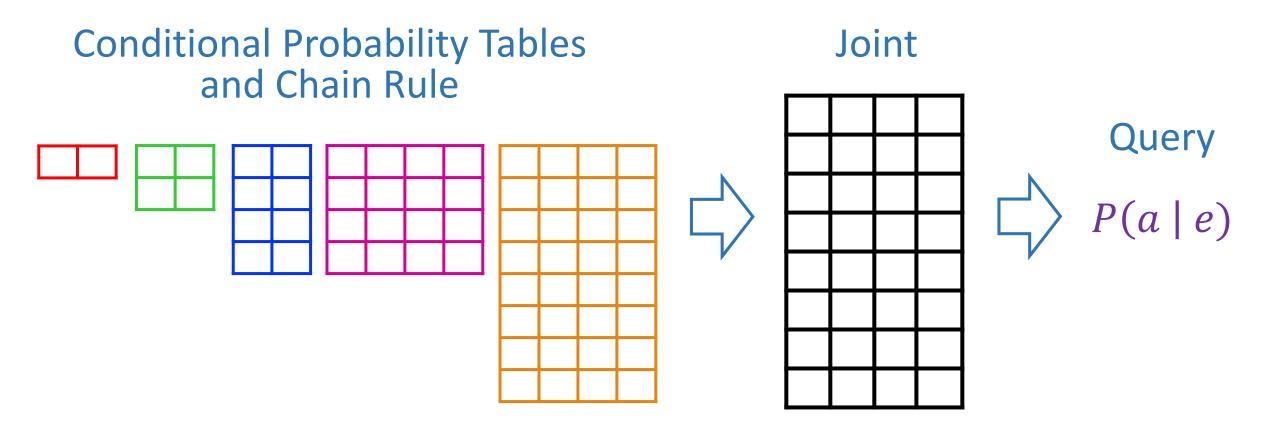


Given the Bayes net, write the joint distribution?

## Answer Any Query from Bayes Net



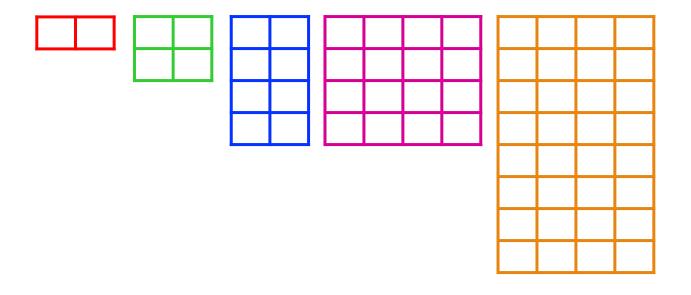
## Answer Any Query from Condition Probability Tables



P(A) P(B|A) P(C|A,B) P(D|A,B,C) P(E|A,B,C,D)

### Answer Any Query from Condition Probability Tables

# Conditional Probability Tables and Chain Rule



#### **Problems**

- Huge
  - n variables with d values
  - $d^n$  entries
- We aren't given the right tables

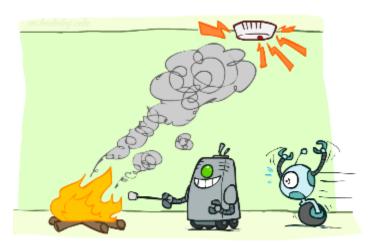
P(A) P(B|A) P(C|A,B) P(D|A,B,C) P(E|A,B,C,D)

Danielle Belgrave, Microsoft Research

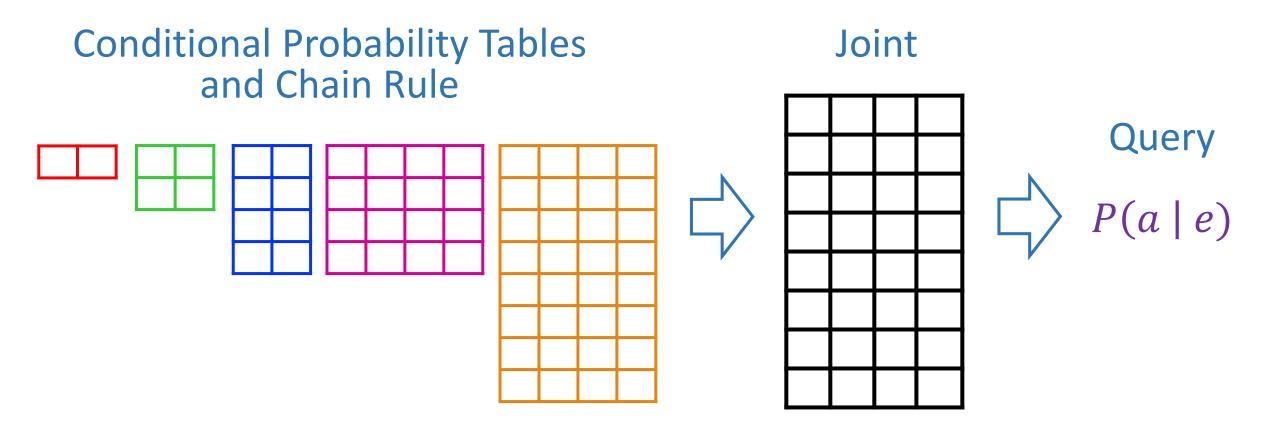
#### Do We Need the Full Chain Rule?

#### Binary random variables

- Fire
- Smoke
- Alarm

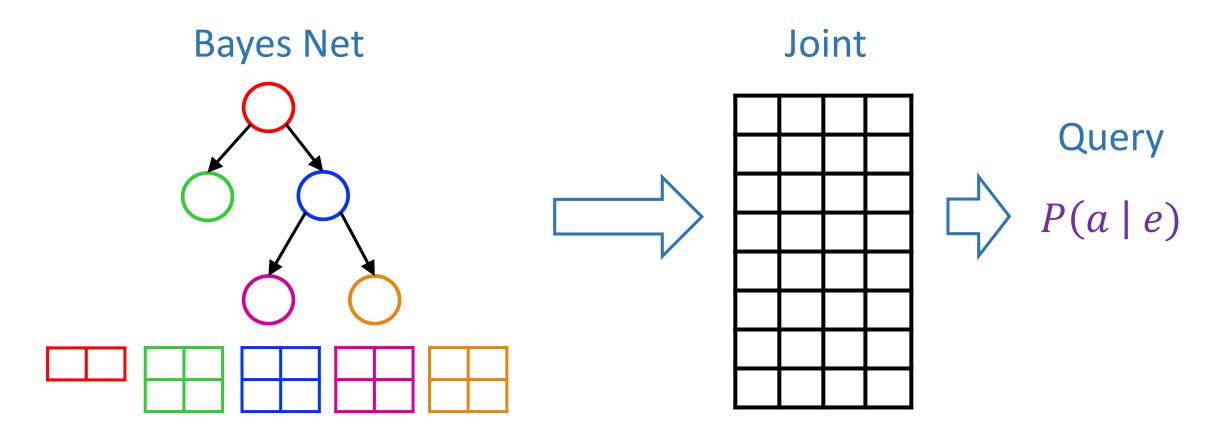


## Answer Any Query from Condition Probability Tables



P(A) P(B|A) P(C|A,B) P(D|A,B,C) P(E|A,B,C,D)

## Answer Any Query from Condition Probability Tables



P(A) P(B|A) P(C|A) P(D|C) P(E|C)

$$P(X_1, ..., X_N) = \prod_{i} P(X_i | Parents(X_i))$$