

# Factor Zoo I

- Joint distribution:  $P(X, Y)$

- Entries  $P(x, y)$  for all  $x, y$
- Sums to 1

- Selected joint:  $P(x, Y)$

- A slice of the joint distribution
- Entries  $P(x, y)$  for fixed  $x$ , all  $y$
- Sums to  $P(x)$

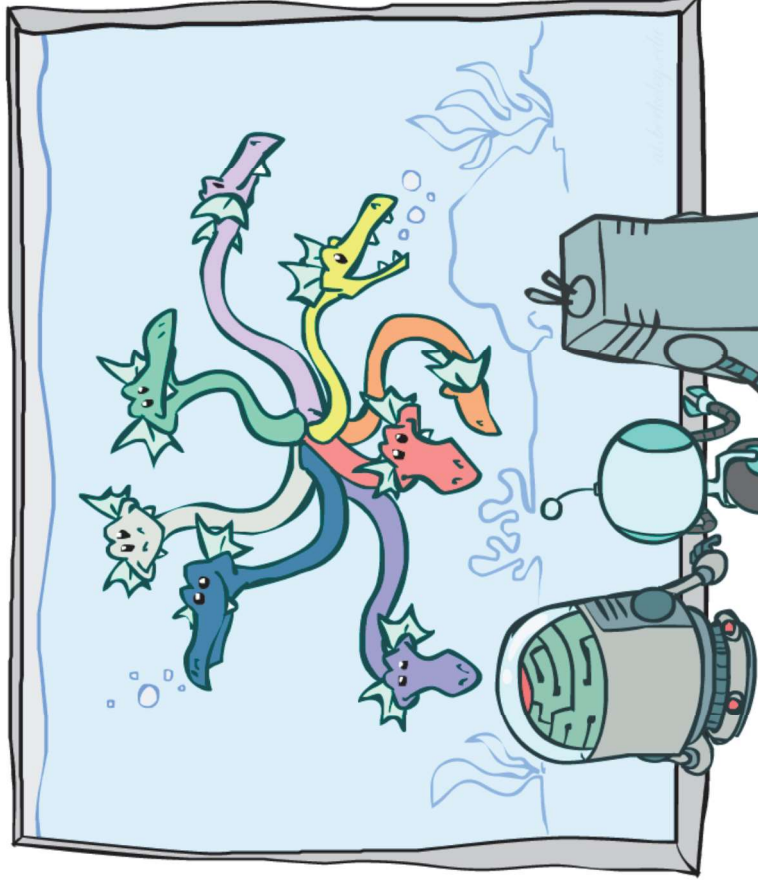
- Number of capitals = dimensionality of the table

$P(T, W)$

T	W	P
hot	sun	0.4
hot	rain	0.1
cold	sun	0.2
cold	rain	0.3

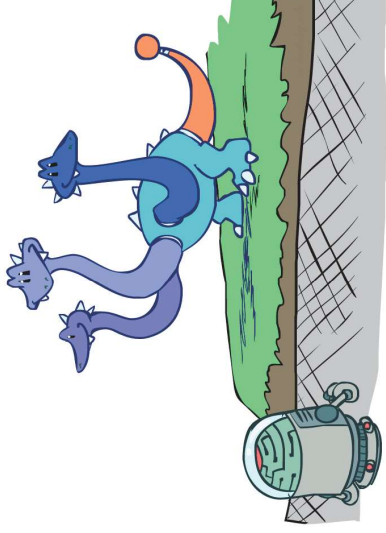
$P(\text{cold}, W)$

T	W	P
cold	sun	0.2
cold	rain	0.3



# Factor Zoo II

- Single conditional:  $P(Y | x)$ 
  - Entries  $P(y | x)$  for fixed  $x$ , all  $y$
  - Sums to 1



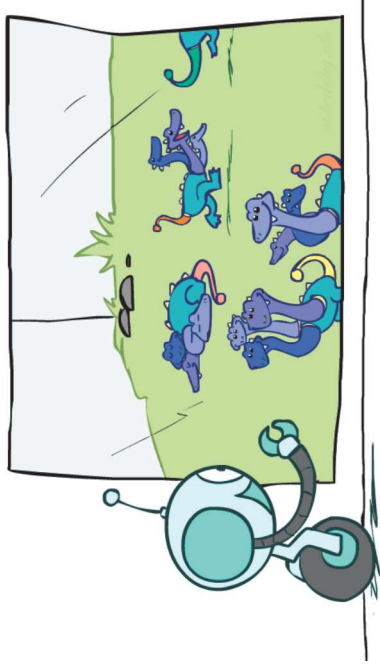
$$P(W|cold)$$

T	W	P
cold	sun	0.4
cold	rain	0.6

- Family of conditionals:

$$P(Y | X)$$

- Multiple conditionals
  - Entries  $P(y | x)$  for all  $x, y$
  - Sums to  $|X|$



$$P(W|T)$$

T	W	P
hot	sun	0.8
hot	rain	0.2
cold	sun	0.4
cold	rain	0.6

$$P(W|hot)$$

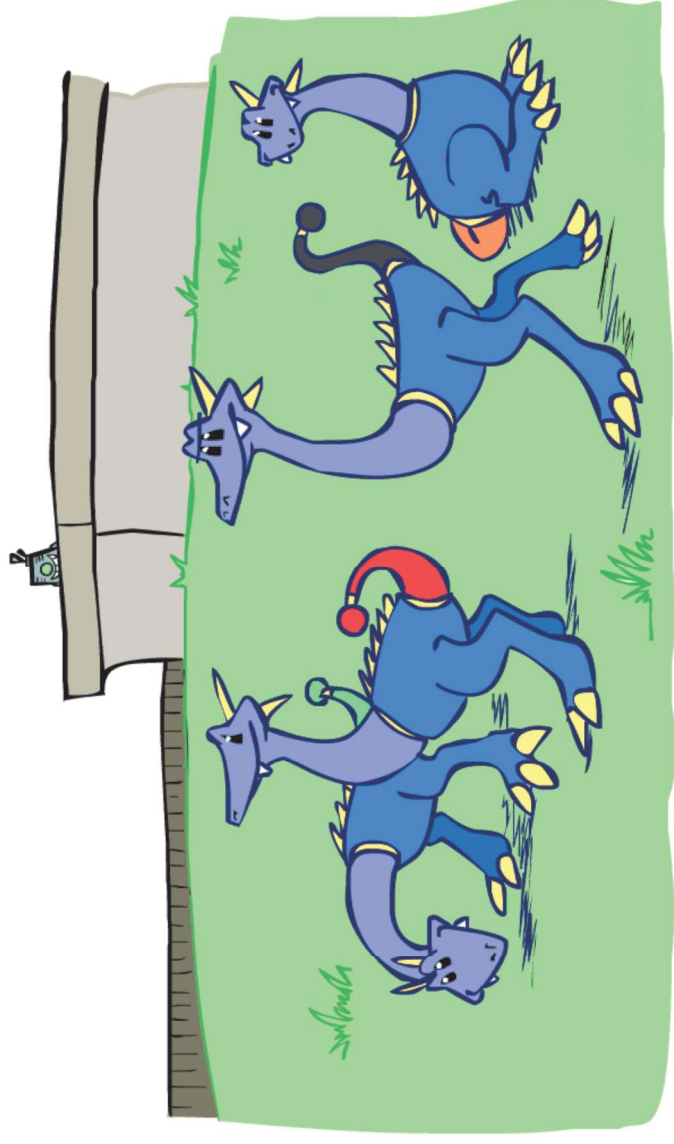
$$P(W|cold)$$

# Factor Zoo III

- Specified family:  $P(y \mid X)$ 
  - Entries  $P(y \mid x)$  for fixed  $y$ , but for all  $x$
  - Sums to ... who knows!

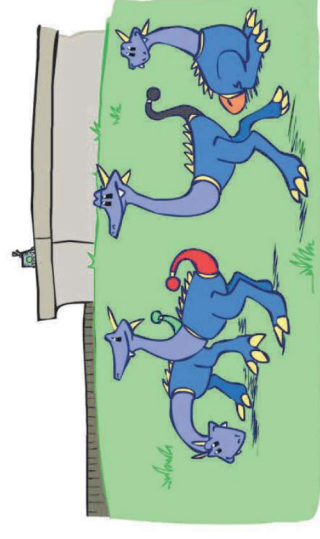
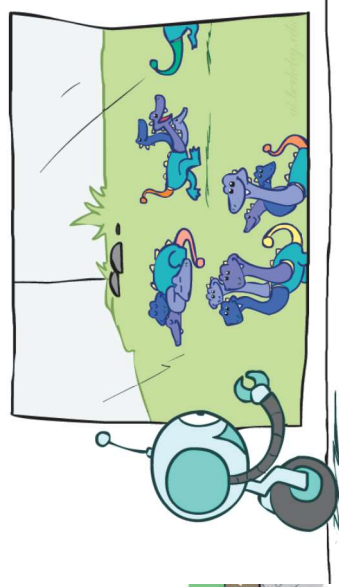
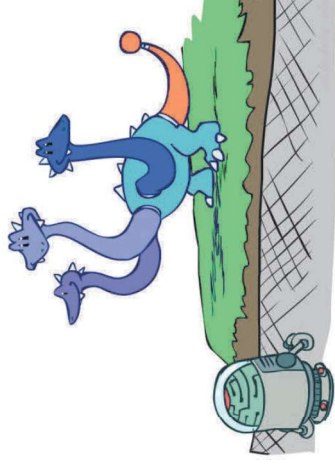
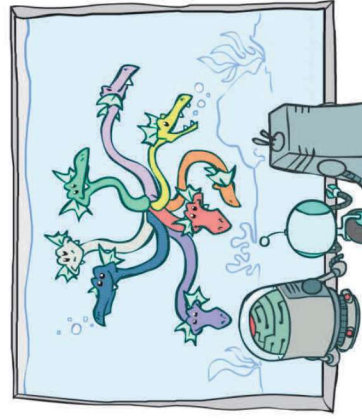
$P(\text{rain} \mid T)$

T	W	P	
hot	rain	0.2	$P(\text{rain} \mid \text{hot})$
cold	rain	0.6	$P(\text{rain} \mid \text{cold})$



# Factor Zoo Summary

- In general, when we write  $P(Y_1 \dots Y_N \mid X_1 \dots X_M)$ 
  - It is a “factor,” a multi-dimensional array
  - Its values are  $P(y_1 \dots y_N \mid x_1 \dots x_M)$
  - Any assigned (=lower-case) X or Y is a dimension missing (selected) from the array



# Example: Traffic Domain

- Random Variables

- R: Raining
- T: Traffic
- L: Late for class!



$$P(L) = ?$$

$$= \sum_{r,t} P(r, t, L)$$

$$= \sum_{r,t} P(r)P(t|r)P(L|t)$$

$$P(R)$$

+r	0.1
-r	0.9

$$P(T|R)$$

+r	+t	0.8
+r	-t	0.2
-r	+t	0.1
-r	-t	0.9

$$P(L|T)$$

+t	+l	0.3
+t	-l	0.7
-t	+l	0.1
-t	-l	0.9



# Inference by Enumeration: Procedural Outline

- Track objects called **factors**

- Initial factors are local CPTs (one per node)

$$P(R)$$

$+r$	0.1
$-r$	0.9

$$P(T|R)$$

+r	+t	0.8
+r	-t	0.2
-r	+t	0.1
-r	-t	0.9

$$P(L|T)$$

+t	+l	0.3
+t	-l	0.7
-t	+l	0.1
-t	-l	0.9

- Any known values are selected

- E.g. if we know  $L = +\ell$  the initial factors are

$$P(R)$$

$+r$	0.1
$-r$	0.9

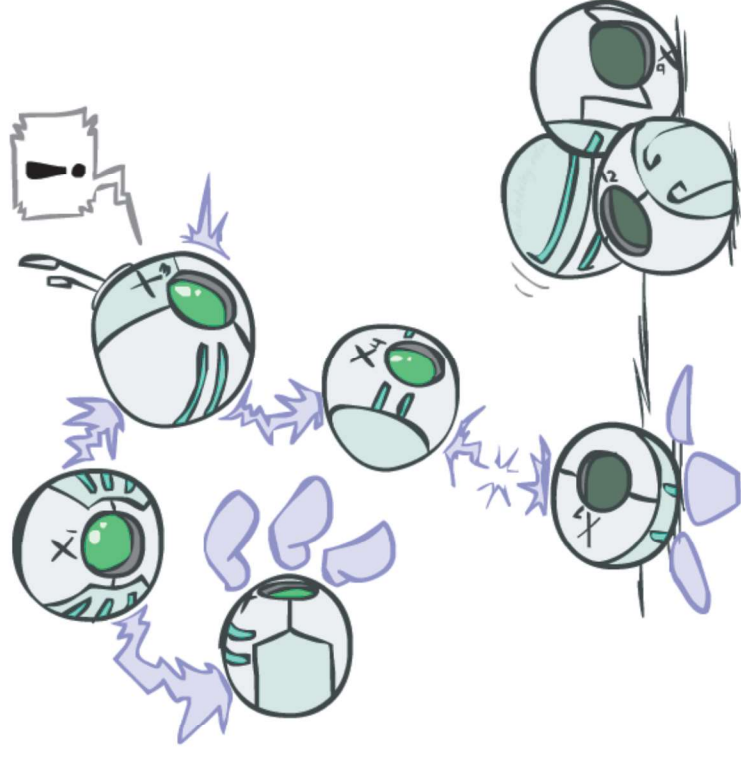
$$P(T|R)$$

+r	+t	0.8
+r	-t	0.2
-r	+t	0.1
-r	-t	0.9

$$P(+\ell|T)$$

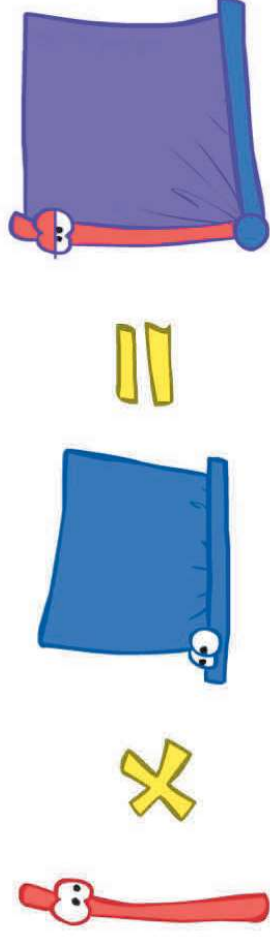
+t	+l	0.3
-t	+l	0.1

- Procedure: Join all factors, eliminate all hidden variables, normalize

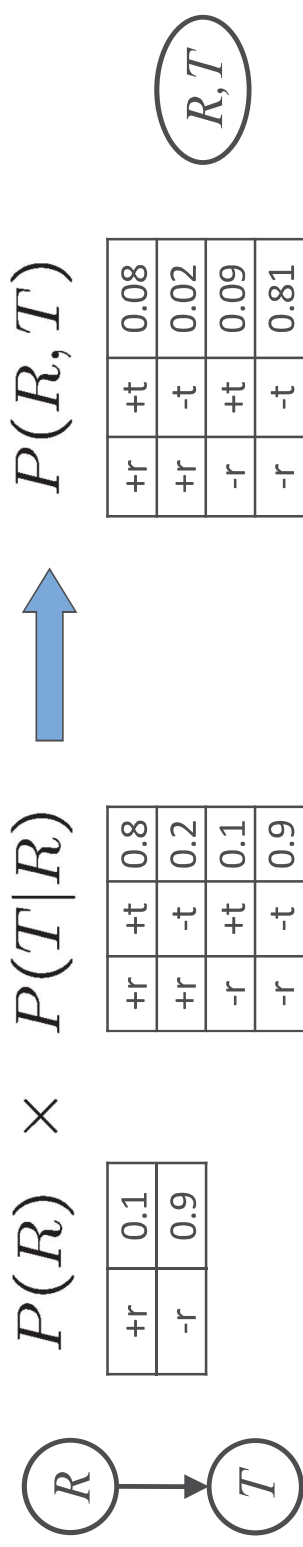


# Operation 1: Join Factors

- First basic operation: **joining factors**
- Combining factors:
  - Just like a database join**
  - Get all factors over the joining variable
  - Build a new factor over the union of the variables involved



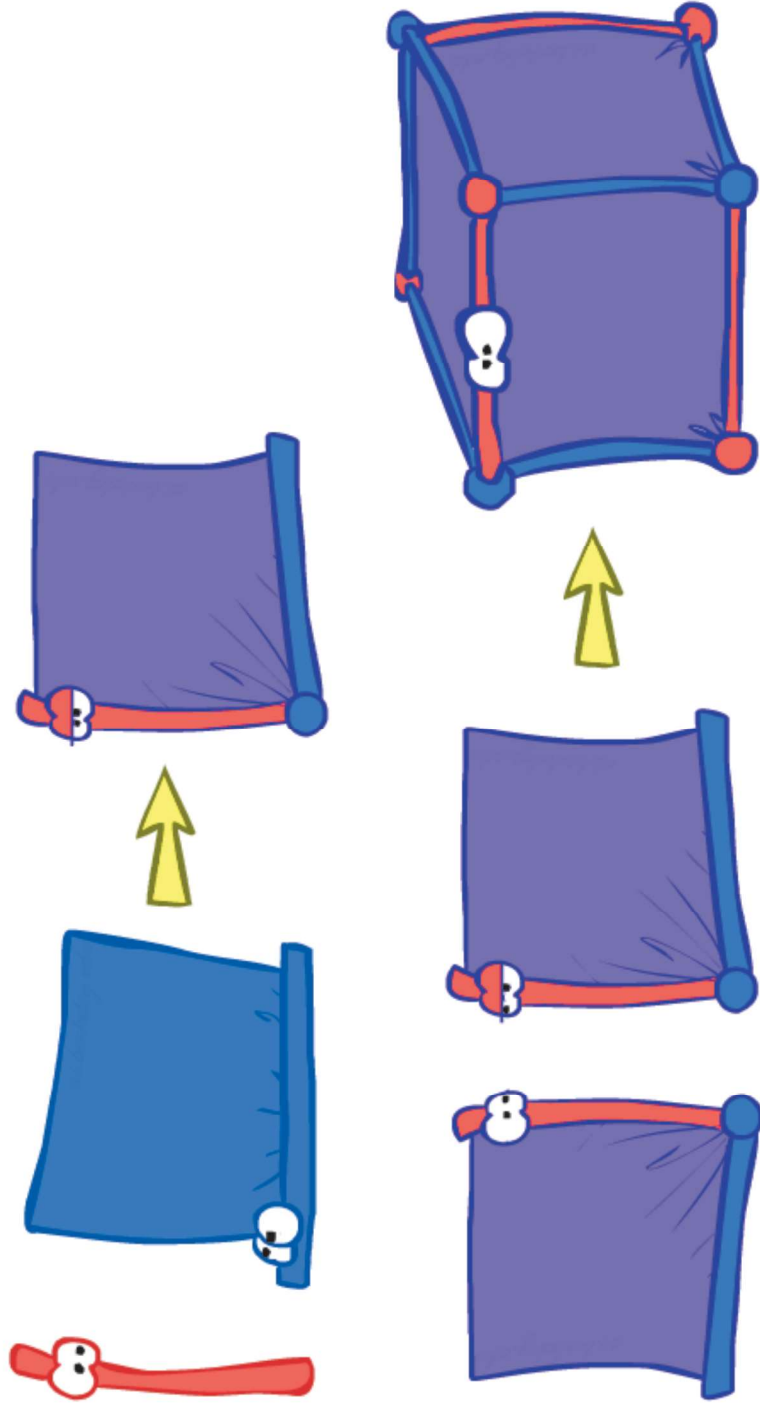
- Example: Join on R



- Computation for each entry: pointwise products  $\forall r, t : P(r, t) = P(r) \cdot P(t|r)$



# Example: Multiple Joins





# Example: Multiple Joins

$P(R)$

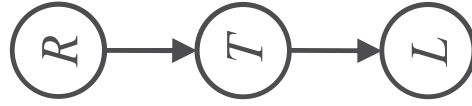
+r	0.1
-r	0.9

$P(T|R)$

+r	+t	0.8
+r	-t	0.2
-r	+t	0.1
-r	-t	0.9

$P(L|T)$

+t	+l	0.3
+t	-l	0.7
-t	+l	0.1
-t	-l	0.9



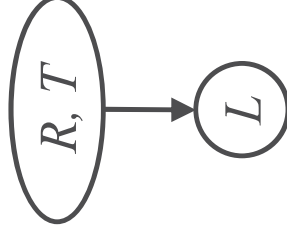
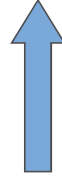
$P(R, T)$

+r	+t	0.08
+r	-t	0.02
-r	+t	0.09
-r	-t	0.81

$P(L|T)$

+t	+l	0.3
+t	-l	0.7
-t	+l	0.1
-t	-l	0.9

Join R



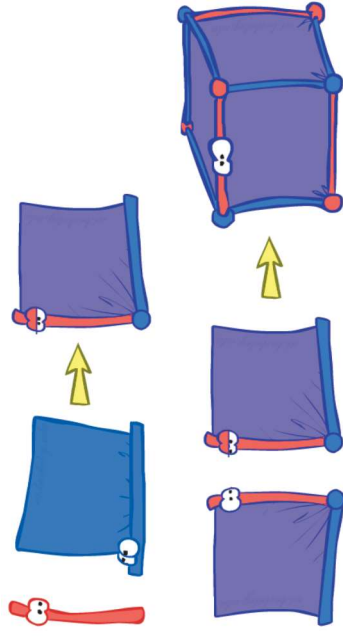
Join T



$R, T, L$

$P(R, T, L)$

+r	+t	+l	0.024
+r	+t	-l	0.056
+r	-t	+l	0.002
+r	-t	-l	0.018
-r	+t	+l	0.027
-r	+t	-l	0.063
-r	-t	+l	0.081
-r	-t	-l	0.729



# Operation 2: Eliminate

- Second basic operation: **marginalization**
- Take a factor and sum out a variable
  - Shrinks a factor to a smaller one
  - A **projection** operation

- Example:

$P(R, T)$

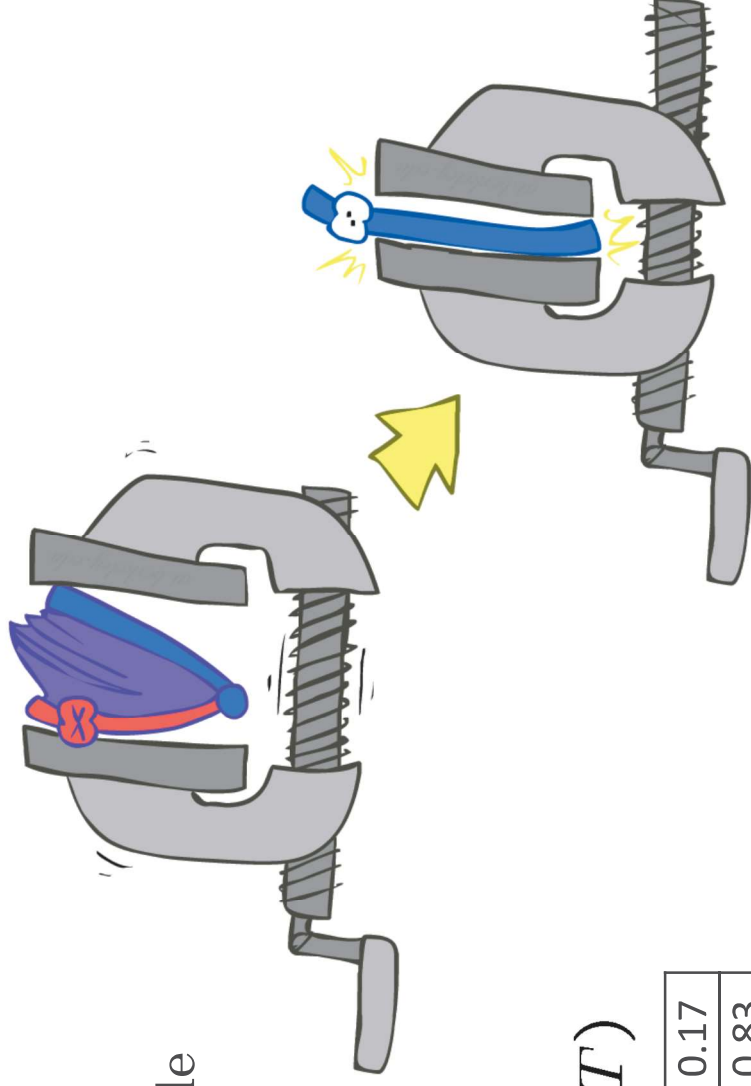
+r	+t	0.08
+r	-t	0.02
-r	+t	0.09
-r	-t	0.81

sum  $R$



$P(T)$

+t	0.17
-t	0.83



# Multiple Elimination

$P(R, T, L)$

$R, T, L$

+r	+t	+l	0.024
+r	+t	-l	0.056
+r	-t	+l	0.002
+r	-t	-l	0.018
-r	+t	+l	0.027
-r	+t	-l	0.063
-r	-t	+l	0.081
-r	-t	-l	0.729

Sum  
out R

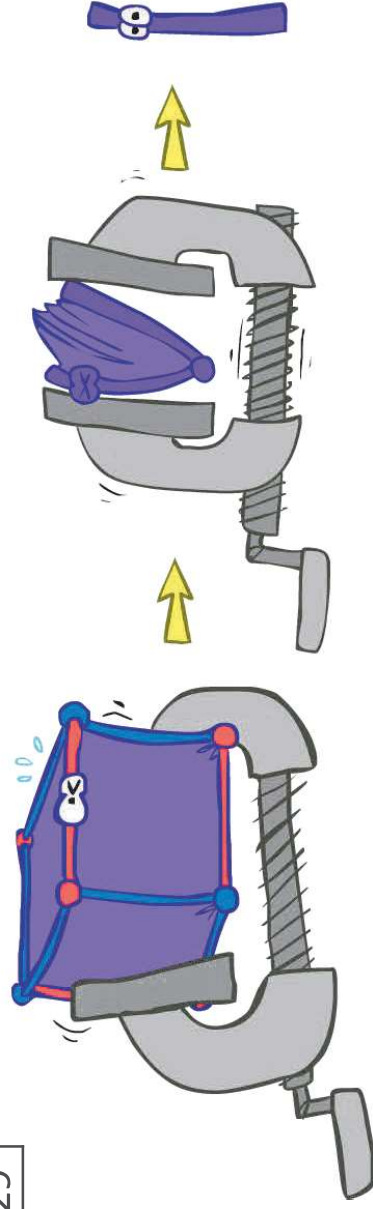
$P(T, L)$

+t	+l	0.051
+t	-l	0.119
-t	+l	0.083
-t	-l	0.747

Sum  
out T

$P(L)$

+l	0.134
-l	0.886



Thus Far: Multiple Join, Multiple Eliminate (= Inference  
by Enumeration)

