## Basic Probability

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Some images and slides are used from:

1. CS188 UC Berkeley
2. RN, AIMA

## Definition

- Probability theory is nothing but common sense reduced to calculation. $\sim$ Pierre Laplace
- What is probability? What does it mean when we say "the probability that a coin will land head is 0.5 "


## Frequentist Vs Bayesian

DID THE SUN JUST EXPLODE?
(TSS NGHT, SO WERE NOT SURE.)


FREQUENTIST STATISTICIAN:
BAYESIAN STATISTICIAN:


## Random variables

## What is a random variable?

Suppose that the variable a denotes the outcome of a role of a single six-sided die:

$a$ is a random variable
this is the domain of a

Another example:
Suppose $b$ denotes whether it is raining or clear outside:

$$
b \in\{\text { rain, clear }\}=B
$$

## Probability distribution

A probability distribution associates each with a probability of occurrence.
A probability table is one way to encode the distribution:

$$
a \in\{1,2,3,4,5,6\}=A \quad b \in\{\text { rain }, \text { clear }\}=B
$$

| a | $\mathrm{P}(\mathrm{a})$ |
| :---: | :---: |
| 1 | $1 / 6$ |
| 2 | $1 / 6$ |
| 3 | $1 / 6$ |
| 4 | $1 / 6$ |
| 5 | $1 / 6$ |
| 6 | $1 / 6$ |


| b | $\mathrm{P}(\mathrm{b})$ |
| :---: | :---: |
| rain | $1 / 4$ |
| clear | $3 / 4$ |

All probability distributions must satisfy the following:

1. $\forall a \in A, a \geq 0$
2. $\sum_{a \in A}, a=1$

## Writing probabilities

| a | $\mathrm{P}(\mathrm{a})$ |
| :---: | :---: |
| 1 | $1 / 6$ |
| 2 | $1 / 6$ |
| 3 | $1 / 6$ |
| 4 | $1 / 6$ |
| 5 | $1 / 6$ |
| 6 | $1 / 6$ |


| b | $\mathrm{P}(\mathrm{b})$ |
| :---: | :---: |
| rain | $1 / 4$ |
| clear | $3 / 4$ |

For example: $\quad p(a=2)=1 / 6$

$$
p(b=c l e a r)=3 / 4
$$

But, sometimes we will abbreviate this as: $\quad p(2)=1 / 6$

$$
p(\text { clear })=3 / 4
$$

## Joint probability distributions

Given random variables: $X_{1}, X_{2}, \ldots, X_{n}$
The joint distribution is a probability assignment to all combinations:

$$
P\left(X_{1}=x_{1}, X_{2}=x_{2}, \ldots, X_{n}=x_{n}\right)
$$

$$
\text { or: } \quad P\left(x_{1}, x_{2}, \ldots, x_{n}\right)
$$

As with single-variate distributions, joint distributions must satisfy:

1. $\quad P\left(x_{1}, x_{2}, \ldots, x_{n}\right) \geq 0$
2. $\sum_{x_{1}, \ldots, x_{n}} P\left(x_{1}, x_{2}, \ldots, x_{n}\right)=1$

## Joint probability distributions

Joint distributions are typically written in table form:

| T | W | $\mathrm{P}(\mathrm{T}, \mathrm{W})$ |
| :---: | :---: | :---: |
| hot | sun | 0.4 |
| hot | rain | 0.1 |
| cold | sun | 0.2 |
| cold | rain | 0.3 |

## Marginalization

Given $P(T, W)$, calculate $P(T)$ or $P(W)$...

|  |  |  | $P(T)=\sum_{w \in W} P(T, w)$ | T | $\mathrm{P}(\mathrm{T})$ |
| :---: | :---: | :---: | :---: | :---: | :---: |
|  |  |  | hot | 0.5 |
| T | W | $\mathrm{P}(\mathrm{T}, \mathrm{W})$ |  | cold | 0.5 |
| hot | sun | 0.4 |  | $P(W)=\sum_{t \in T} P(t, W)$ |  |  |
| hot | rain | 0.1 |  |  |  |
| cold | sun | 0.2 | W |  | P(W) |
| cold | rain | 0.3 | sun |  | 0.5 |
|  |  |  | rain |  | 0.4 |

## Marginalization



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## Conditional Probabilities

$$
P(\operatorname{sun} \mid h o t) \equiv \quad \text { Probability that it is sunny given that it is hot. }
$$

| T | W | $\mathrm{P}(\mathrm{T}, \mathrm{W})$ |
| :---: | :---: | :---: |
| hot | sun | 0.4 |
| hot | rain | 0.1 |
| cold | sun | 0.2 |
| cold | rain | 0.3 |

## Conditional Probabilities

Calculate the conditional probability using the product rule:

Product rule

$$
P(a \mid b)=\frac{P(a, b)}{P(b)}
$$

| T | W | $\mathrm{P}(\mathrm{T}, \mathrm{W})$ |
| :---: | :---: | :---: |
| hot | sun | 0.4 |
| hot | rain | 0.1 |
| cold | sun | 0.2 |
| cold | rain | 0.3 |

$$
\begin{aligned}
P(W & =s \mid T=c)=\frac{P(W=s, T=c)}{P(T=c)}=\frac{0.2}{0.5}=0.4 \\
& =P(W=s, T=c)+P(W=r, T=c) \\
& =0.2+0.3=0.5
\end{aligned}
$$

## Conditional Probabilities

- $P(+x \mid+y)$ ?

$$
P(X, Y)
$$

| $X$ | $Y$ | $P$ |
| :---: | :---: | :---: |
| $+x$ | $+y$ | 0.2 |
| $+x$ | $-y$ | 0.3 |
| $-x$ | $+y$ | 0.4 |
| $-x$ | $-y$ | 0.1 |

- $P(-x \mid+y)$ ?
- $P(-y \mid+x)$ ?


## Conditional distribution

Given $\mathrm{P}(\mathrm{T}, \mathrm{W})$, calculate $\mathrm{P}(\mathrm{T} \mid \mathrm{w})$ or $\mathrm{P}(\mathrm{W} \mid \mathrm{t})$...

| T | W | $\mathrm{P}(\mathrm{T}, \mathrm{W})$ |
| :---: | :---: | :---: |
| hot | sun | 0.4 |
| hot | rain | 0.1 |
| cold | sun | 0.2 |
| cold | rain | 0.3 |

$$
P(W \mid t)=\frac{P(W, t)}{P(t)}
$$

## Conditional distribution

Given $\mathrm{P}(\mathrm{T}, \mathrm{W})$, calculate $\mathrm{P}(\mathrm{T} \mid \mathrm{w})$ or $\mathrm{P}(\mathrm{W} \mid \mathrm{t})$...

| T | W | $\mathrm{P}(\mathrm{T}, \mathrm{W})$ |
| :---: | :---: | :---: |
| hot | sun | 0.4 |
| hot | rain | 0.1 |
| cold | sun | 0.2 |
| cold | rain | 0.3 |

$$
P(W \mid t)=\frac{P(W, t)}{P(t)}
$$

$$
\begin{aligned}
P(\text { sun } \mid \text { hot })=\frac{P(\text { sun }, \text { hot })}{P(\text { hot })} & =\frac{P(\text { sun }, \text { hot })}{P(\text { sun }, \text { hot })+P(\text { rain }, \text { hot })} \\
& =\frac{0.4}{0.4+0.1}
\end{aligned}
$$

## Conditional distribution

Given $\mathrm{P}(\mathrm{T}, \mathrm{W})$, calculate $\mathrm{P}(\mathrm{T} \mid \mathrm{w})$ or $\mathrm{P}(\mathrm{W} \mid \mathrm{t}) \ldots$

| T | W | $\mathrm{P}(\mathrm{T}, \mathrm{W})$ |
| :---: | :---: | :---: |
| hot | sun | 0.4 |
| hot | rain | 0.1 |
| cold | sun | 0.2 |
| cold | rain | 0.3 |

## Conditional distribution

Given $\mathrm{P}(\mathrm{T}, \mathrm{W})$, calculate $\mathrm{P}(\mathrm{T} \mid \mathrm{w})$ or $\mathrm{P}(\mathrm{W} \mid \mathrm{t})$...

| T | W | $\mathrm{P}(\mathrm{T}, \mathrm{W})$ |
| :---: | :---: | :---: |
| hot | sun | 0.4 |
| hot | rain | 0.1 |
| cold | sun | 0.2 |
| cold | rain | 0.3 |


| - | W | $\mathrm{P}(\mathrm{W} \mid t=h o t)$ |
| :---: | :---: | :---: |
|  | sun | 0.8 |
|  | rain | 0.2 |
| $P(W \mid t)=\frac{P(W, t)}{P(t)}$ |  |  |
|  | W | $\mathrm{P}(\mathrm{W} \mid t=$ cold $)$ |
|  | sun | 0.4 |
|  | rain | 0.6 |

$$
\begin{aligned}
P(\text { sun } \mid \text { cold })=\frac{P(\text { sun }, \text { cold })}{P(\text { cold })} & =\frac{P(\text { sun }, \text { cold })}{P(\text { sun }, \text { cold })+P(\text { rain }, \text { cold })} \\
& =\frac{0.2}{0.2+0.3}
\end{aligned}
$$

## Normalization

Given $\mathrm{P}(\mathrm{T}, \mathrm{W})$, calculate $\mathrm{P}(\mathrm{T} \mid \mathrm{w})$ or $\mathrm{P}(\mathrm{W} \mid \mathrm{t})$...


## Normalization

| T | W | $\mathrm{P}(\mathrm{T}, \mathrm{W})$ |
| :---: | :---: | :---: |
| hot | sun | 0.4 |
| hot | rain | 0.1 |
| cold | sun | 0.2 |
| cold | rain | 0.3 |



Select corresponding elts from the joint distribution

Scale the numbers so that they sum to 1 .

$$
P(\text { sun } \mid \text { cold })=\frac{P(\text { sun }, \text { cold })}{P(\text { cold })}=\frac{P(\text { sun }, \text { cold })}{P(\text { sun }, \text { cold })+P(\text { rain }, \text { cold })}
$$

## Normalization

| T | W | $\mathrm{P}(\mathrm{T}, \mathrm{W})$ |
| :---: | :---: | :---: |
| hot | sun | 0.4 |
| hot | rain | 0.1 |
| cold | sun | 0.2 |
| cold | rain | 0.3 |



Select corresponding elts from the joint distribution

Scale the numbers so that they sum to 1 .

$$
P(\operatorname{sun} \mid \operatorname{cold})=\frac{P(\operatorname{sun}, \operatorname{cold})}{P(\operatorname{cold})}=\frac{P(\operatorname{sun}, \operatorname{cold})}{\underline{P(\text { sun })} \text { cold })+P(\text { rain }, \text { cold })}
$$

The only purpose of this denominator is to make the distribution sum to one.

- we achieve the same thing by scaling.


## Normalization

$$
P(X \mid Y=-y) ?
$$

$P(X, Y)$

| $x$ | $y$ | $p$ |
| :---: | :---: | :---: |
| $+x$ | $+y$ | 0.2 |
| $+x$ | $-y$ | 0.3 |
| $-x$ | $+y$ | 0.4 |
| $-x$ | $-y$ | 0.1 |

## Bayes Rule

$$
P(a \mid b)=\frac{P(b \mid a) P(a)}{P(b)}
$$



## Bayes Rule

$$
P(a \mid b)=\frac{P(b \mid a) P(a)}{P(b)}
$$

It's easy to derive from the product rule:

$$
P(a, b)=P(b \mid a) P(a)=\underbrace{P(a \mid b)} P(b)
$$

Solve for this

## Using Bayes Rule

$$
P(a \mid b)=\frac{P(b \mid a) P(a)}{P(b)}
$$

$$
P(\text { cause } \mid e f f e c t)=\frac{P(e f f e c t \mid \text { cause }) P(\text { cause })}{P(e f f e c t)}
$$

## Using Bayes Rule

$$
\begin{gathered}
P(a \mid b)=\frac{P(b \mid a) P(a)}{P(b)} \\
P(\text { cause } \mid \text { effect })=\frac{:-\cdots(e f f e c t \mid \text { cause }) P(\text { cause })}{P(\text { effect })}
\end{gathered}
$$

It's often easier to estimate this

## Bayes Rule Example

$$
P(\text { cause } \mid \text { effect })=\frac{P(\text { effect } \mid \text { cause }) P(\text { cause })}{P(e f f e c t)}
$$

Suppose you have a stiff neck...
Suppose there is a $70 \%$ chance of meningitis if you have a stiff neck:

$$
\text { stiff neck } \quad \text { meningitis }
$$

What are the chances that you have meningitis?

## Bayes Rule Example

$$
P(\text { cause } \mid e f f e c t)=\frac{P(\text { effect } \mid \text { cause }) P(\text { cause })}{P(\text { effect })}
$$

Suppose you have a stiff neck...
Suppose there is a $70 \%$ chance of meningitis if you have a stiff neck:

> stiff neck meningitis

$$
P\left(s \mid m^{\prime}\right)=0.7
$$

What are the chances that you have meningitis?

We need a little more information...

## Bayes Rule Example

$$
\begin{aligned}
& P(\text { cause } \mid e f f e c t)=\frac{P(e f f e c t \mid c a u s e) P(c a u s e)}{P(e f f e c t)} \\
& P(s \mid m)=0.7 \\
& P(s)=0.01 \\
& P(m)=\frac{1}{50000} \quad \text { Prior probability of stiff neck } \\
& P(m \mid s)=\frac{P(s \mid m) P(m)}{P(s)}=\frac{0.7 \times \frac{1}{50000}}{0.01}=0.0014
\end{aligned}
$$

## Bayes Rule Example

$$
\begin{aligned}
& P(\text { cause } \mid e f f e c t)=\frac{P(e f f e c t \mid c a u s e) P(c a u s e)}{P(e f f e c t)} \\
& P(s \mid m)=0.7 \\
& P(s)=0.01 \\
& P(m)=\frac{1}{50000} \quad \text { Prior probability of stiff neck } \\
& P(m \mid s)=\frac{P(s \mid m) P(m)}{P(s)}=\frac{0.7 \times \frac{1}{50000}}{0.01}=0.0014
\end{aligned}
$$

## Bayes Rule Example

- Given:
$P(D \mid W)$

| $D$ | $W$ | $P$ |
| :---: | :---: | :---: |
| wet | sun | 0.1 |
| dry | sun | 0.9 |
| wet | rain | 0.7 |
| dry | rain | 0.3 |

- What is P(W | dry) ?

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