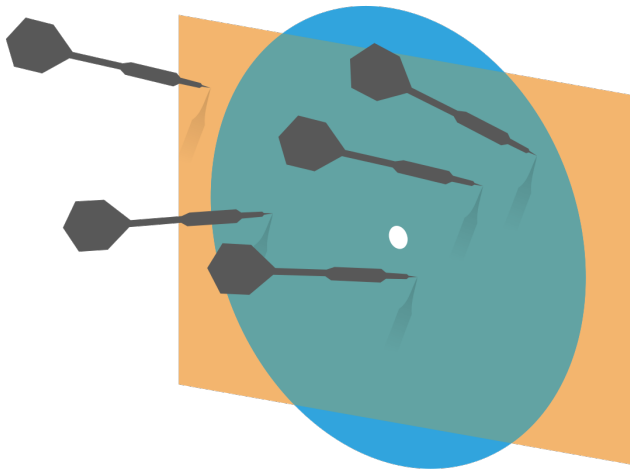


Probability and Statistics for Computer Science

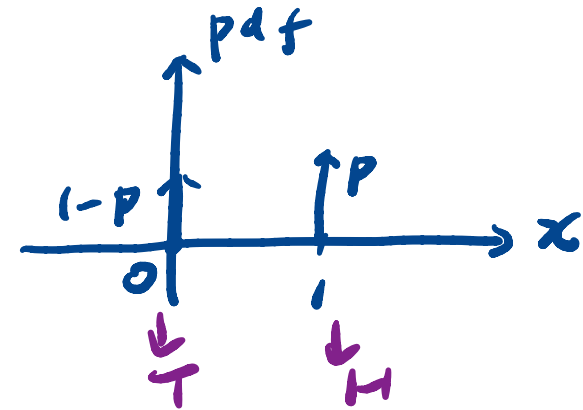


Credit: wikipedia

“I have now used each of the terms mean, variance, covariance and standard deviation in two slightly different ways.” ---Prof. Forsythe

Last time

- ✱ Random Variable
- ✱ Probability distribution
- ✱ Cumulative distribution



$$P(X = x) ?$$

$$P(X = 1)$$

Objectives

- ✱ Random Variable
 - ✱ *Conditional & Joint Probability*
 - ✱ *Expected value*
 - ✱ *Variance & covariance*

Conditional probability distribution of random variables

- ✱ The conditional probability distribution of X given Y is

$$P(x|y) = \frac{P(x, y)}{P(y)} \quad P(y) \neq 0$$

(Note: In the original image, $P(x, y)$ is written in purple and has an arrow pointing to the y in the denominator.)

$$P(x, y) = P(X=x \cap Y=y)$$

$$P(y) = P(Y=y)$$

$$P(x|y) = P(X=x | Y=y)$$

Conditional probability distribution of random variables

- ✱ The conditional probability distribution of X given Y is

$$P(x|y) = \frac{P(x, y)}{P(y)} \quad P(y) \neq 0$$

for all x

- ✱ The joint probability distribution of two random variables X and Y is

$$P(\{X = x\} \cap \{Y = y\})$$

$$\sum_x P(x|y) = 1$$

Get the marginal from joint distri.

- ✱ We can recover the individual probability distributions from the joint probability distribution

$$P(x) = \sum_y P(x, y) \quad \text{RHS} = \sum_y P(y|x) P(x) = P(x) \cdot \sum_y P(y|x)$$

$$P(y) = \sum_x P(x, y)$$

Joint probabilities sum to 1

- ✱ The sum of the joint probability distribution

$$\sum_y \sum_x P(x, y) = 1$$

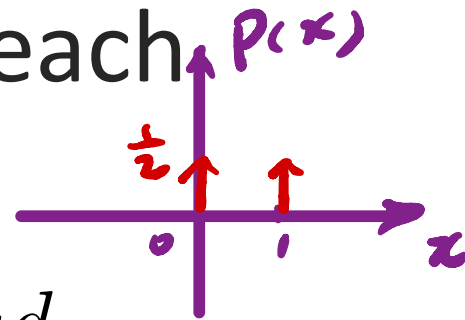
$$\begin{aligned} \text{LHS} &= \sum_y P(y) \\ &= 1 \end{aligned}$$

Joint Probability Example

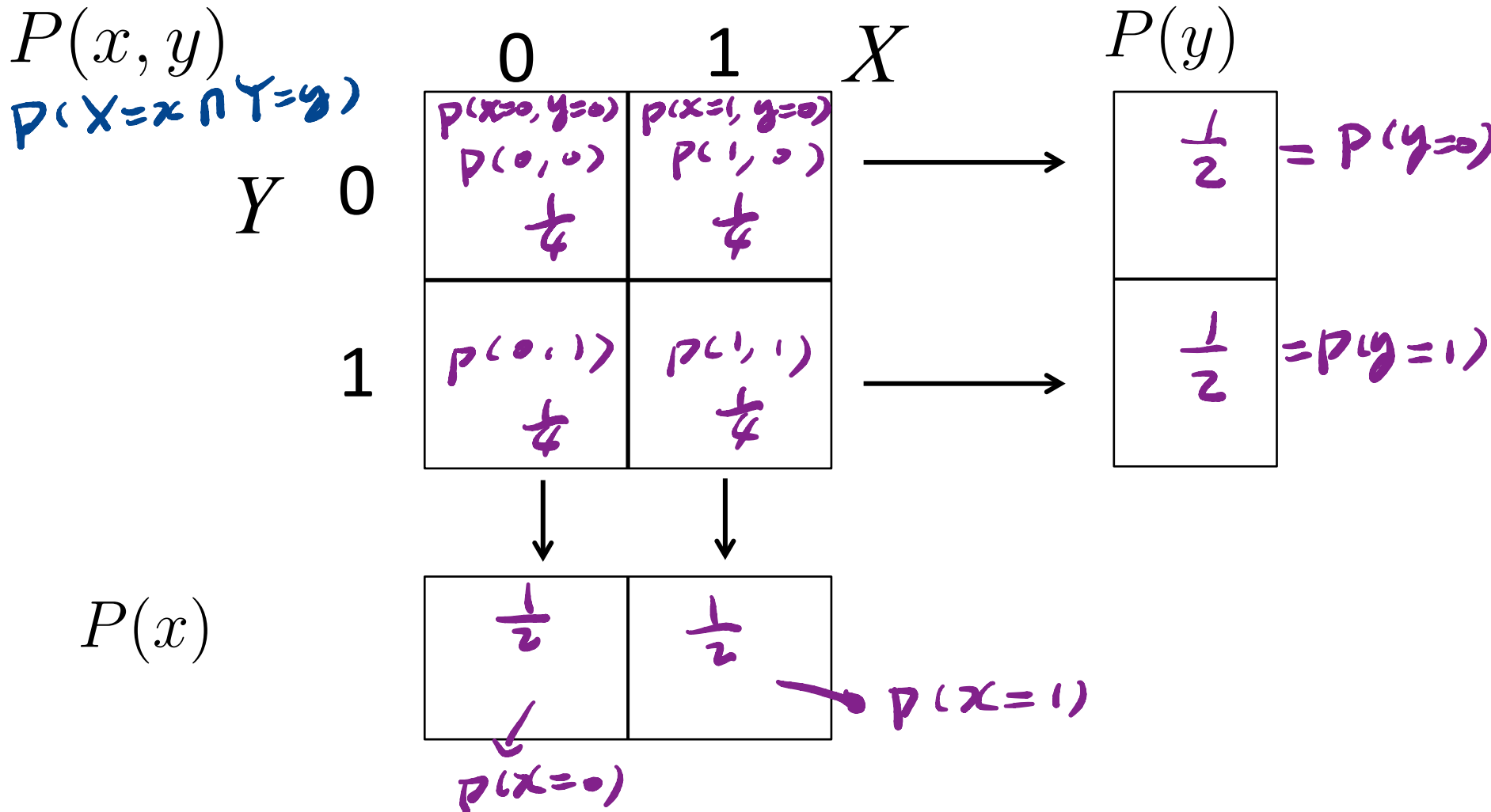
- ✱ Tossing a ^{fair} coin twice, we define random variable X and Y for each toss.

$$X(\omega) = \begin{cases} 1 & \text{outcome of } \omega \text{ is head} \\ 0 & \text{outcome of } \omega \text{ is tail} \end{cases}$$

$$Y(\omega) = \begin{cases} 1 & \text{outcome of } \omega \text{ is head} \\ 0 & \text{outcome of } \omega \text{ is tail} \end{cases}$$



Joint probability distribution example



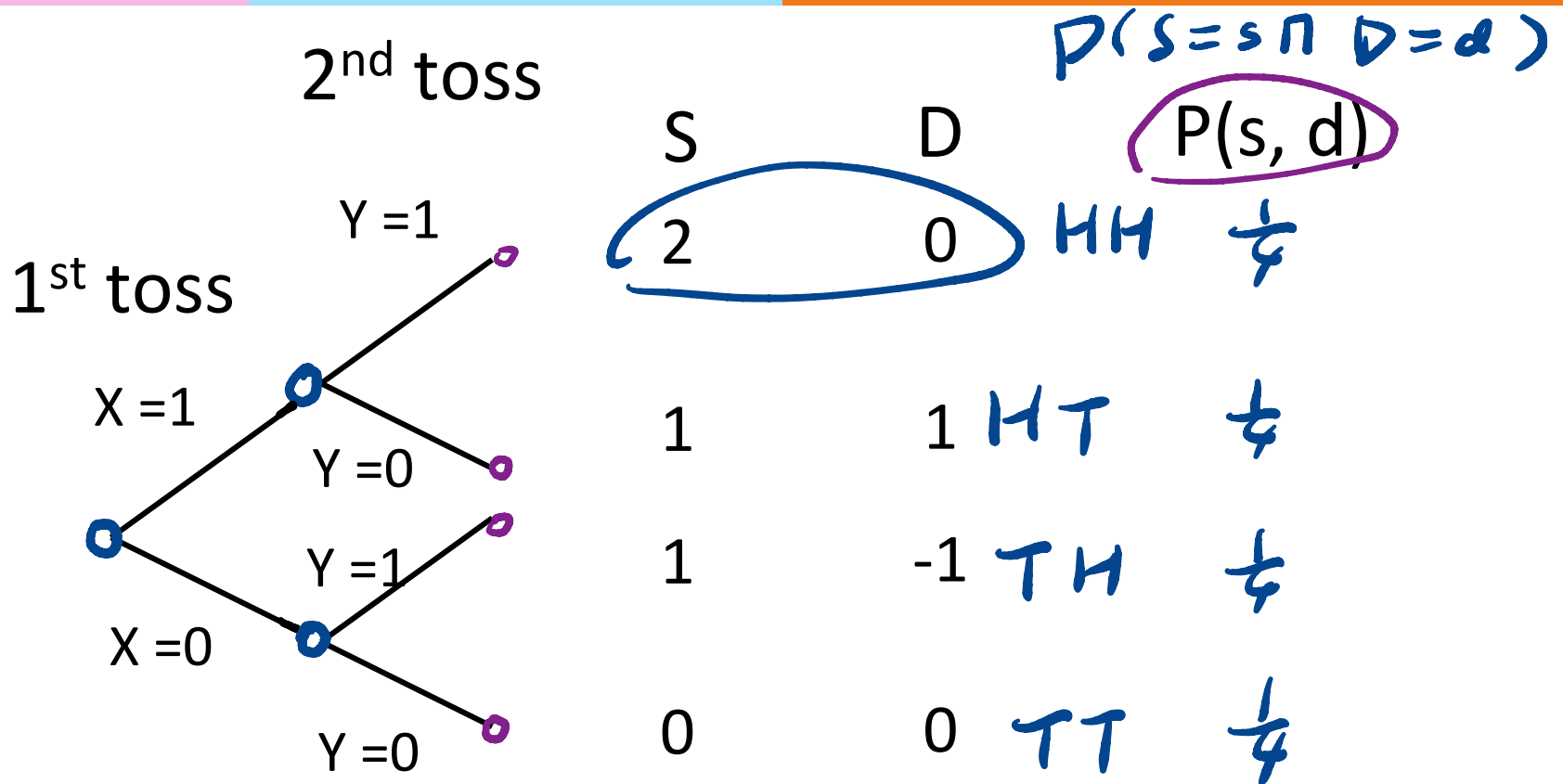
Joint Probability Example

Now we define Sum $\mathbf{S} = X + Y$, Difference $\mathbf{D} = X - Y$. \mathbf{S} takes on values $\{0, 1, 2\}$ and \mathbf{D} takes on values $\{-1, 0, 1\}$

$$X(\omega) = \begin{cases} 1 & \text{outcome of } \omega \text{ is head} \\ 0 & \text{outcome of } \omega \text{ is tail} \end{cases}$$

$$Y(\omega) = \begin{cases} 1 & \text{outcome of } \omega \text{ is head} \\ 0 & \text{outcome of } \omega \text{ is tail} \end{cases}$$

Joint Probability Example



Suppose coin is fair, and the tosses are independent

Joint probability distribution example

$P(s, d)$

		-1	0	1
S	0	0	$\frac{1}{4}$	0
	1	$\frac{1}{4}$	0	$\frac{1}{4}$
	2	0	$\frac{1}{4}$	0

$P(d)$

$\frac{1}{4}$	$\frac{1}{2}$	$\frac{1}{4}$
---------------	---------------	---------------

D

$\frac{1}{4}$	$\rightarrow P(S=0)$
$\frac{1}{2}$	
$\frac{1}{4}$	

$P(S)$

$P(S=s)$

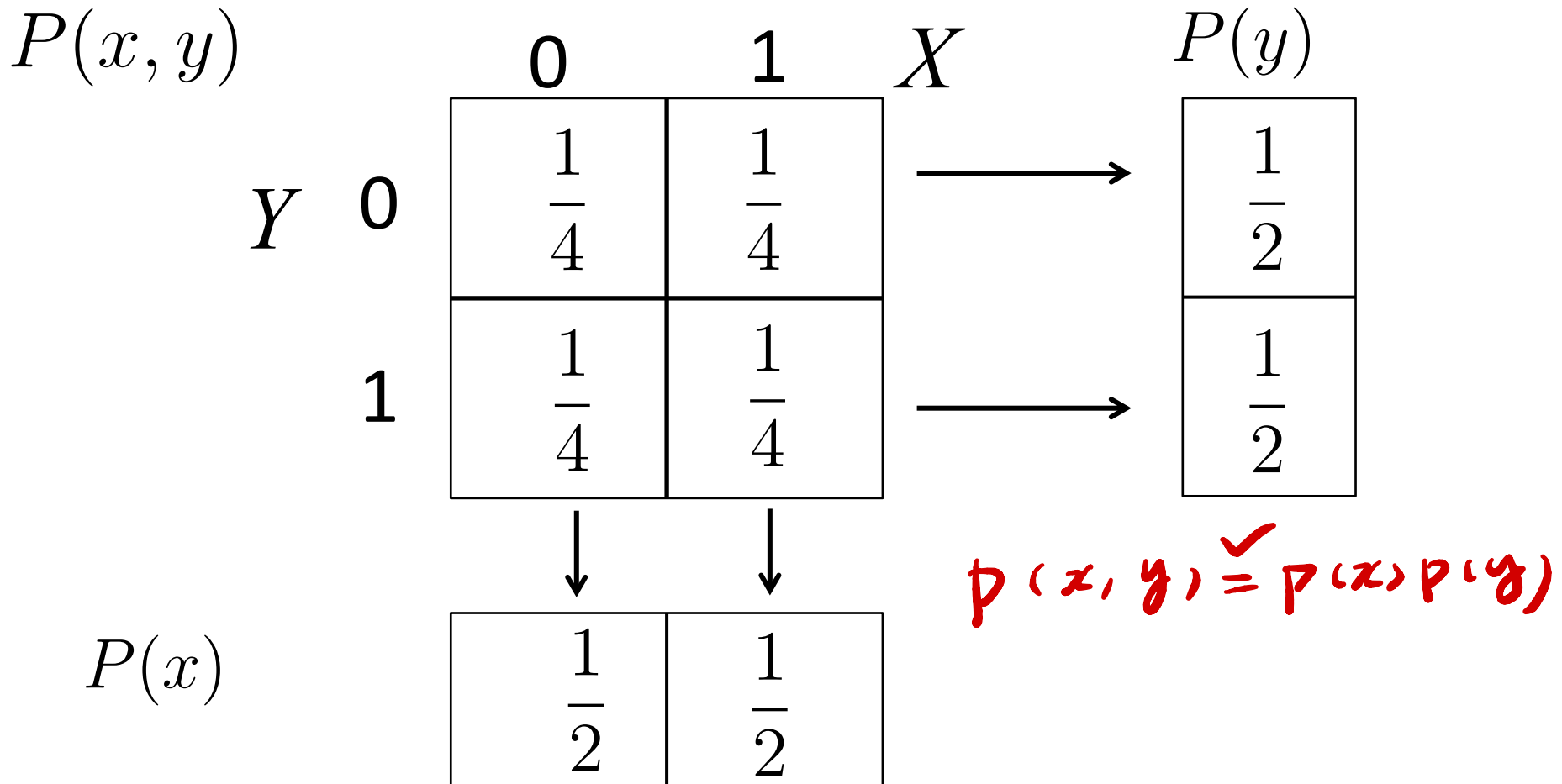
Independence of random variables

- ✱ Random variable X and Y are independent if

$$P(x, y) = P(x)P(y) \text{ for all } x \text{ and } y$$

- ✱ In the previous coin toss example
 - ✱ Are X and Y independent?
 - ✱ Are S and D independent?

Joint probability distribution example



Joint probability distribution example

$P(s, d)$

S

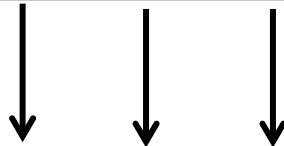
0

1

2

-1 0 1

0	$\frac{1}{4}$	0
$\frac{1}{4}$	0	$\frac{1}{4}$
0	$\frac{1}{4}$	0



$\frac{1}{4}$	$\frac{1}{2}$	$\frac{1}{4}$
---------------	---------------	---------------

$P(d)$

D

$P(s)$



$\frac{1}{4}$
$\frac{1}{2}$
$\frac{1}{4}$

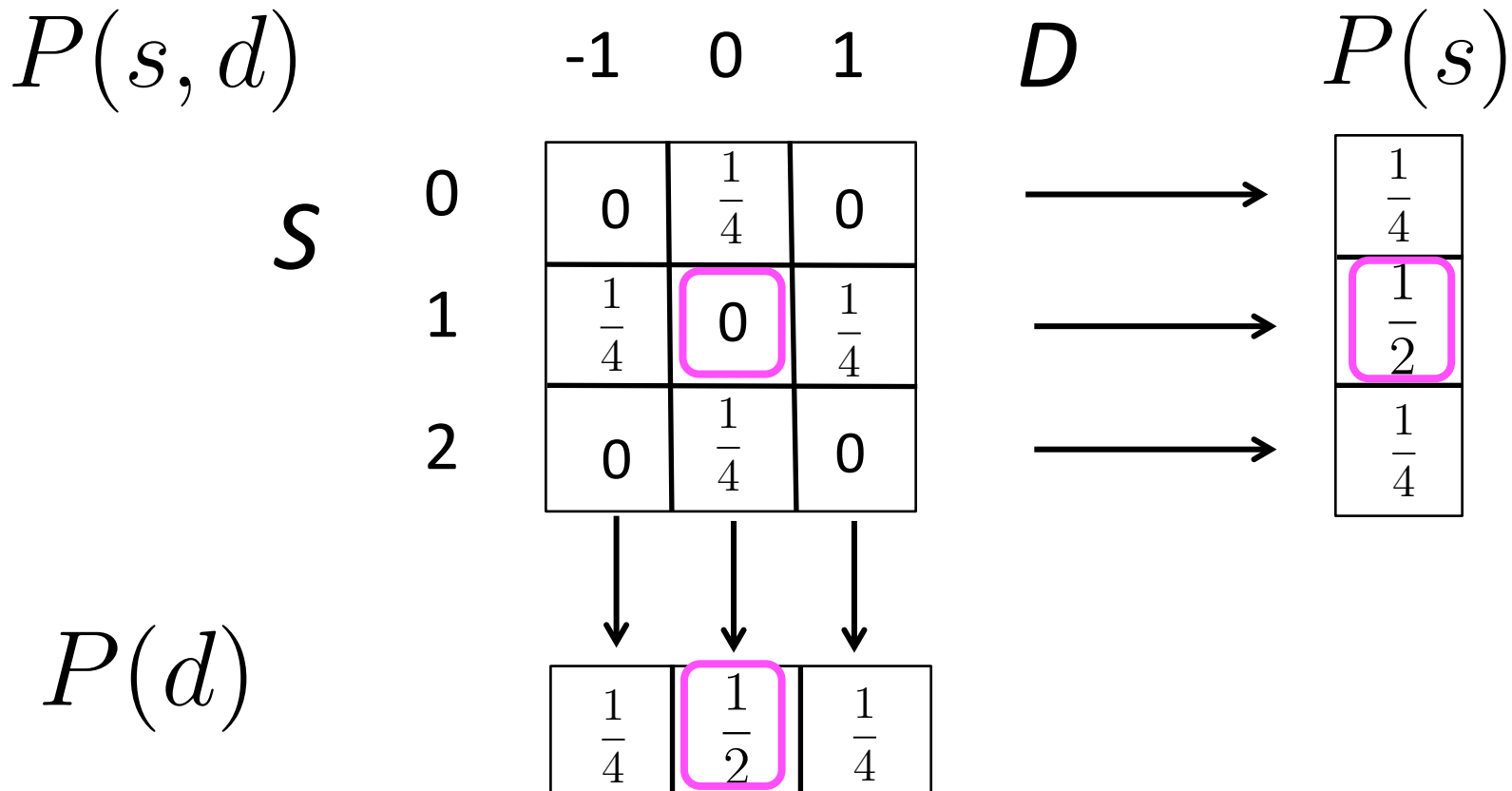
$P(s=1 \cap d=0) = 0$

$P(s=1) = \frac{1}{2}$

$P(d=0) = \frac{1}{2}$

$P(s, d) \neq P(s)P(d)$

Joint probability distribution example



$$P(S = 1, D = 0) \neq P(S = 1)P(D = 0)$$

Conditional probability distribution example

$$P(s|d) = \frac{P(s, d)}{P(d)}$$

		-1	0	1	<i>D</i>
<i>S</i>	0	0	$\frac{1}{2}$	0	
	1	1	0	1	
	2	0	$\frac{1}{2}$	0	

$$P(S=1 | D=-1) = \frac{P(S=1, D=-1)}{P(D=-1)} = \frac{\frac{1}{4}}{\frac{1}{4}} = 1$$

Bayes rule for random variable

- ✱ Bayes rule for events generalizes to random variables

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

$$\rightarrow P(x|y) = \frac{P(y|x)P(x)}{P(y)}$$

$$= \frac{P(y|x)P(x)}{\sum_x P(y|x)P(x)}$$

Total Probability

Conditional probability distribution example

$$P(s|d) = \frac{P(s, d)}{P(d)}$$

		-1	0	1	<i>D</i>
<i>S</i>	0	0	$\frac{1}{2}$	0	
	1	1	0	1	
	2	0	$\frac{1}{2}$	0	

$$P(D = -1|S = 1) = \frac{P(S = 1|D = -1)P(D = -1)}{P(S = 1)} = \frac{1 \times \frac{1}{4}}{\frac{1}{2}}$$

Three important facts of Random variables

- ✱ Random variables have **probability functions**
- ✱ Random variables can be **conditioned** on events or other random variables
- ✱ Random variables have **averages**

Expected value

- ✱ The **expected value** (or **expectation**) of a random variable X is

$$E[X] = \sum_x x P(x)$$

Handwritten annotations:
- A red arrow points from the term $P(x)$ to the expression $\begin{matrix} \leq 1 \\ \geq 0 \end{matrix}$.
- A red arrow points from the term $P(x)$ to the expression $P(X=x)$.
- Below the summation, the expression $\sum_x P(x) = 1$ is written.

The expected value is a **weighted sum** of **all** the values X can take

Expected value

- ✱ The **expected value** of a random variable X is

$$E[X] = \sum_x x P(x)$$

$\sum_x P(x) = 1$

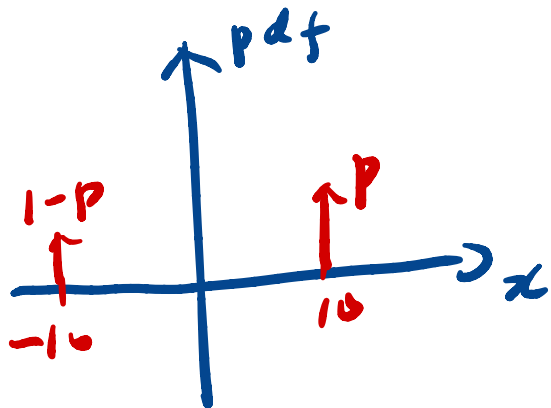
≤ 1

$E[x] = 0.5$ if $p = 1/2$
Theoretical mean

The expected value is a **weighted sum** of all the values X can take

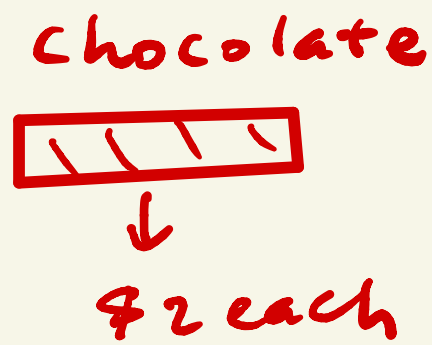
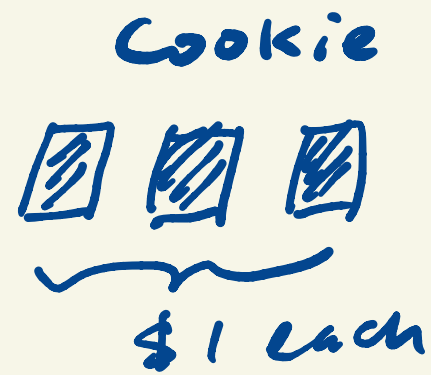
Expected value: profit

- ✱ A company has a project that has p probability of earning 10 million and $1-p$ probability of losing 10 million.
- ✱ Let X be the return of the project.



$$\begin{aligned} E[X] &= \sum x P(x) \\ &= (-10) \cdot (1-p) \\ &\quad + 10 \cdot p \\ &= 20p - 10 \geq 0 \\ p &\geq \frac{1}{2} \end{aligned}$$

After class



A) random draw 1
out of 4

Expected value ?

B) random draw 1 twice with replacement
when the two are the same,
you get the prize.

Expected value ?

Linearity of Expectation

✱ For random variables X and Y and constants k, c

✱ Scaling property

$$E[kX] = kE[X]$$

✱ Additivity

$$E[X + Y] = E[X] + E[Y]$$

✱ And $E[kX + c] = kE[X] + c$

Linearity of Expectation

✱ Proof of the additive property

$$E[X + Y] = E[X] + E[Y] \quad S = X + Y$$

$$E[X + Y] = E(S) = \sum_s s P(s)$$

$$P(s) = P(S = s)$$

Note. $P(S = s) = 0$

if $x + y \neq s$

$$= \sum_{\{S=x+y\}} s \sum_{\{S=x+y\}} P(x, y)$$

$$= \sum_x \sum_y (x + y) P(x, y)$$

Proof conti.

$$E[X+Y] = \sum_x \sum_y (x+y) P(x, y)$$

$$E[aX + bY + cZ] = ?$$

$$aE[X] + bE[Y] + cE[Z]$$

$$= \sum_x \sum_y x P(x, y) + \sum_x \sum_y y P(x, y)$$

$$= \sum_x x \sum_y P(x, y) + \sum_y \sum_x y P(x, y)$$

$$= \sum_x x \cdot P(x) + \sum_y y \sum_x P(x, y)$$

$$= E[X] + \sum_y y P(y)$$

$$= E[X] + E[Y]$$

Q. What's the value?

* What is $E[E[X] + 1]$? ^{Const}

- A. $E[X] + 1$ B. 1 C. 0

$$\begin{aligned} E[E[X] + 1] \\ &= E[E[X]] + E[1] \\ &= E[X] + 1 \end{aligned}$$

Expected value of a function of X


- ✱ If f is a function of a random variable X , then $Y = f(X)$ is a random variable too
- ✱ The expected value of $Y = f(X)$ is

$$E[f(X)] = ?$$

$$E[Y] = \sum_y y P(y)$$

Expected value of a function of X

- ✱ If f is a function of a random variable X , then $Y = f(X)$ is a random variable too
- ✱ The expected value of $Y = f(X)$ is

$$E[Y] = E[f(X)] = \sum_x f(x)P(x)$$


The exchange of variable theorem

$$E[f(x)] = E[Y] = \sum y P(y)$$

if $\{x\} \rightarrow \{y\}$ is bijection $P(y) = P(x)$

$$= \sum y P(x)$$

$$= \sum_x f(x) P(x)$$

if several x in $\{x_s\} \rightarrow$ one y_s value

$$E[Y] = \sum_{y \in I} y P(x) + y_s P(y_s)$$

$$= \sum_{y \in I} y P(x) + y_s \sum_{x \in \{x_s\}} P(x)$$

$$= \sum y P(x) = \sum_x f(x) P(x)$$

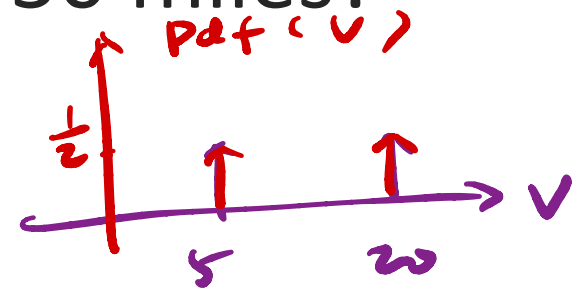
Expected time of cat

- ✱ A cat moves with random constant speed V , either 5 mile/hr or 20 mile/hr with equal probability, what's the expected time for it to travel 50 miles?

$$E[T] = ? \quad \neq \frac{D}{E[V]}$$

$$T = \frac{50}{V} = f(V)$$

$$E\left[\frac{50}{V}\right] = \sum f(v) P(v) = \frac{50}{v_1} \cdot P(v_1) + \frac{50}{v_2} \cdot P(v_2) = 6 \dots$$



$v_1 = 5$
 $v_2 = 20$

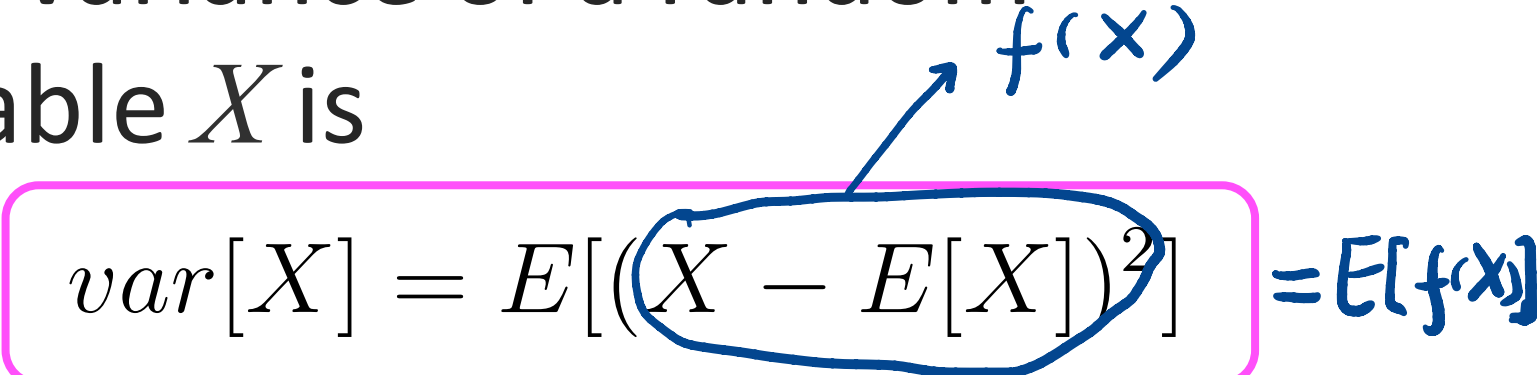
Q: Is this statement true?

If there exists a constant such that $P(X \geq a) = 1$, then $E[X] \geq a$. It is:

- A. True
- B. False

Variance and standard deviation

- ✱ The variance of a random variable X is

$$\text{var}[X] = E[(X - E[X])^2] = E[f(x)]$$


- ✱ The standard deviation of a random variable X is

$$\text{std}[X] = \sqrt{\text{var}[X]}$$

Properties of variance

- ✱ For random variable X and constant k

$$\text{var}[X] \geq 0$$

$$\text{var}[kX] = k^2 \text{var}[X]$$

A neater expression for variance

- ✱ Variance of Random Variable X is defined as:

$$\mathit{var}[X] = E[(X - E[X])^2]$$

- ✱ It's the same as:

$$\mathit{var}[X] = E[X^2] - E[X]^2$$

A neater expression for variance

$$\mathit{var}[X] = E[(X - E[X])^2]$$

A neater expression for variance

$$\mathit{var}[X] = E[(X - E[X])^2]$$

$$\mathit{var}[X] = E[(X - \mu)^2] \quad \text{where } \mu = E[X]$$

A neater expression for variance

$$\text{var}[X] = E[(X - E[X])^2]$$

$$\text{var}[X] = E[(X - \mu)^2] \quad \text{where } \mu = E[X]$$

$$= E[X^2 - 2X\mu + \mu^2]$$

$$= E[X^2] - E[2X\mu] + E[\mu^2]$$

$$= E[X^2] - 2E[X\mu] + E[\mu^2]$$

$$= E[X^2] - 2\overset{\mu}{\underset{\mu}{E[X]}} + \mu^2$$

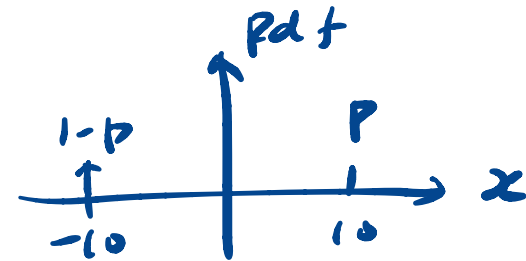
$$= E[X^2] - E[X]^2$$

Variance: the profit example

- For the profit example, what is the variance of the return? We know $E[X] = 20p - 10$

$$E[f(X)] = \sum_x f(x)P(x)$$

$$\text{var}[X] = E[X^2] - (E[X])^2$$



$$= \sum_x x^2 p(x) - (20p - 10)^2$$

$$= 10^2 \cdot p + (-10)^2 (1-p) - (20p - 10)^2$$

Additional References

- ✱ Charles M. Grinstead and J. Laurie Snell
"Introduction to Probability"
- ✱ Morris H. Degroot and Mark J. Schervish
"Probability and Statistics"

See you next time

*See
You!*

