

Properties of Joints:

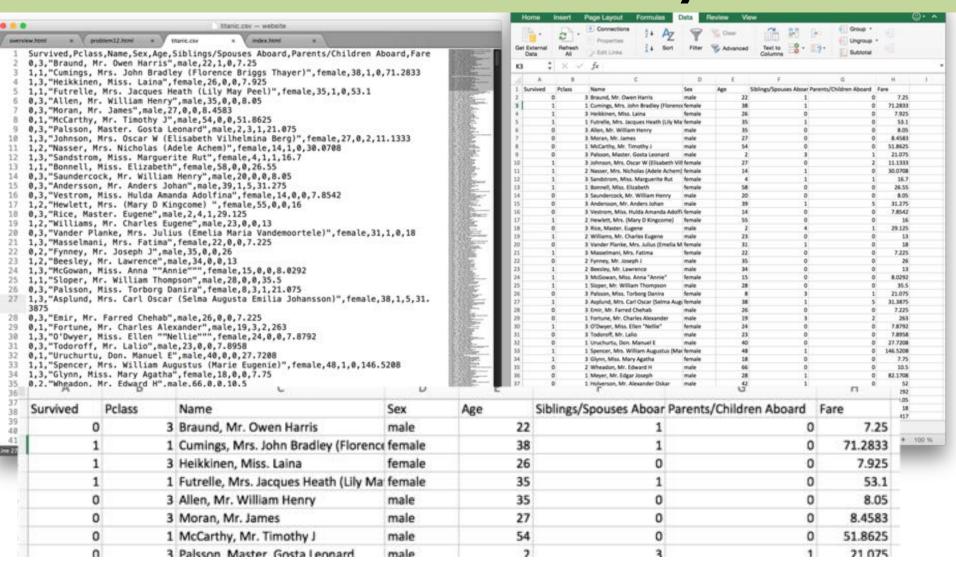
Expectation, Independence, Convolution

Chris Piech

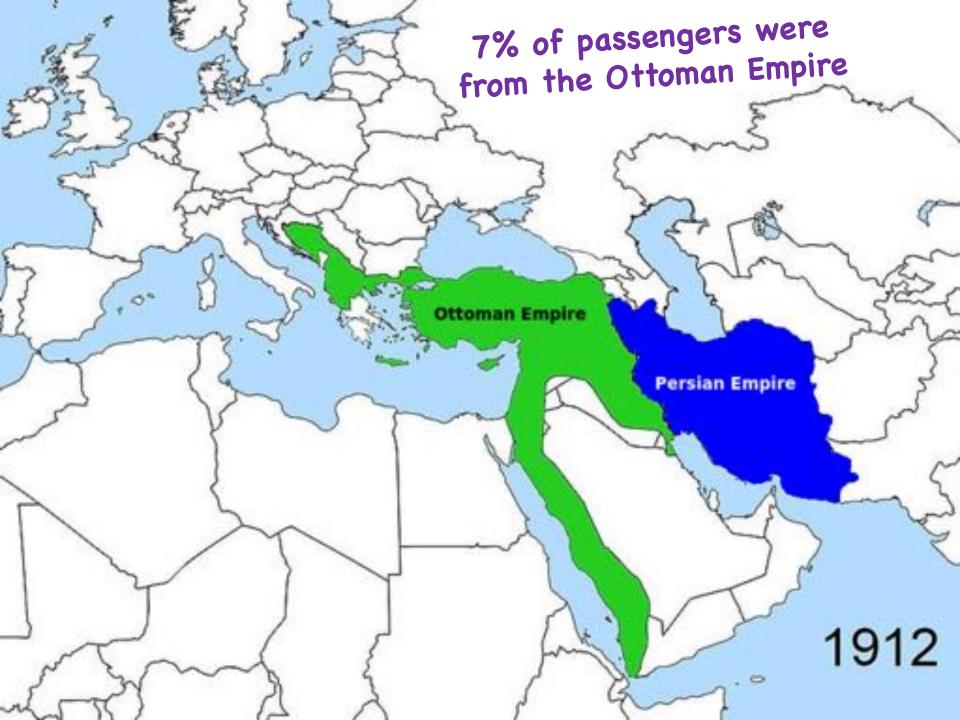
CS109, Stanford University



# **Titanic Probability**

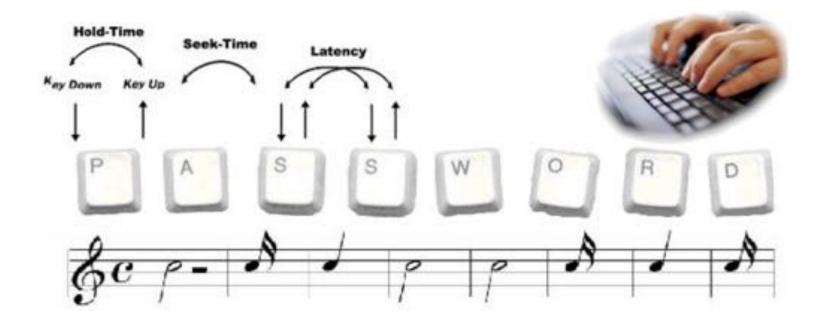








# **Biometric Keystrokes**

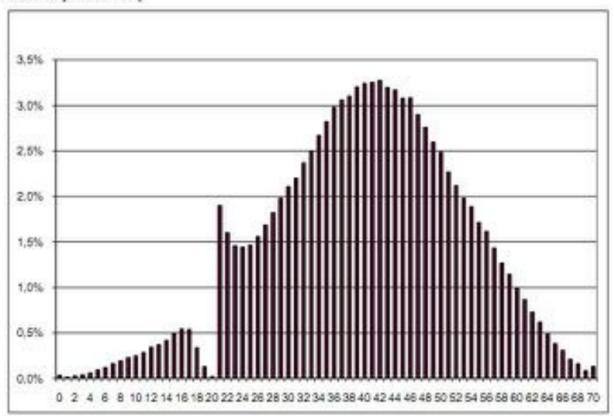


### Altruism?

# Scores for a standardized test that students in Poland are required to pass before moving on in school

See if you can guess the minimum score to pass the test.

#### 2.1. Poziom podstawowy



Wykres 1. Rozkład wyników na poziomie podstawowym

### Joint Random Variables



Use a joint table, density function or CDF to solve probability question



Think about **conditional** probabilities with joint variables (which might be continuous)



Use and find **expectation** of multiple RVS



Use and find independence of multiple RVS



What happens when you add random variables?

# All the Bayes Belong to Us

M,N are discrete. X, Y are continuous

$$\begin{array}{ll} \text{OG BaYes} & p_{\scriptscriptstyle M|N}(m|n) = \frac{P_{\scriptscriptstyle N|M}(n|m)p_{\scriptscriptstyle M}(m)}{p_{\scriptscriptstyle N}(n)} \\ \\ \text{Mix BaYes} & f_{\scriptscriptstyle X|N}(x|n) = \frac{P_{\scriptscriptstyle N|X}(n|x)f_{\scriptscriptstyle X}(x)}{P_{\scriptscriptstyle N}(n)} \\ \\ \text{Mix BaYes} & p_{\scriptscriptstyle N|X}(n|x) = \frac{f_{\scriptscriptstyle X|N}(x|n)p_{\scriptscriptstyle N}(n)}{f_{\scriptscriptstyle X}(x)} \\ \\ \text{Mix Continuous} & f_{\scriptscriptstyle X|Y}(x|y) = \frac{f_{\scriptscriptstyle Y|X}(y|x)f_{\scriptscriptstyle X}(x)}{f_{\scriptscriptstyle Y}(y)} \end{array}$$

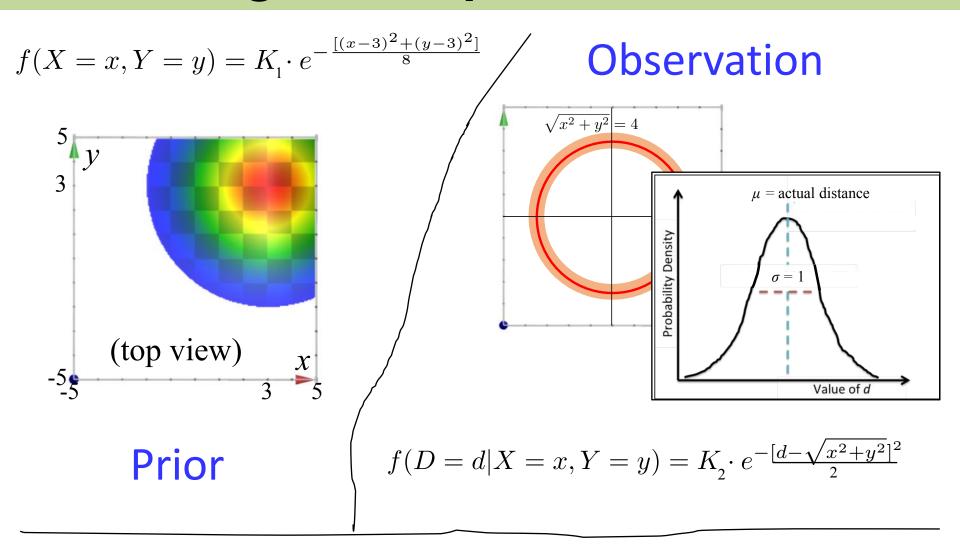
# Tracking in 2D Space?



$$f(X = x, Y = y | D = d)$$



# Tracking in 2D Space: New Belief



What is your *new* belief for the location of the object being tracked? Your joint probability density function can be expressed with a constant

# Tracking in 2D Space: New Belief

$$f(X = x, Y = y | D = 4) = \frac{f(D = 4 | X = x, Y = y) \cdot f(X = x, Y = y)}{f(D = 4)}$$

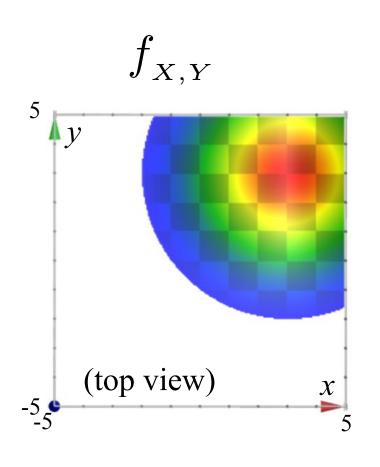
$$= \frac{K_1 \cdot e^{-\frac{[4 - \sqrt{x^2 + y^2})^2]}{2} \cdot K_2 \cdot e^{-\frac{[(x - 3)^2 + (y - 3)^2]}{8}}}{f(D = 4)}$$

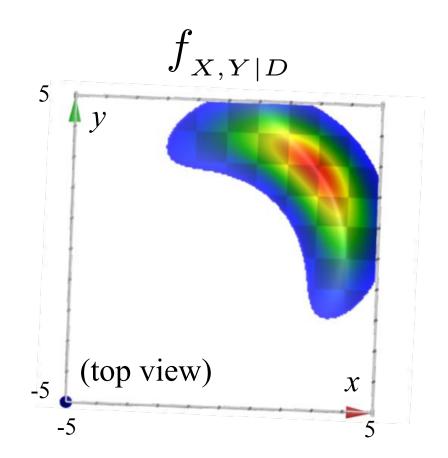
$$= \frac{K_3 \cdot e^{-\left[\frac{[4 - \sqrt{x^2 + y^2})^2]}{2} + \frac{[(x - 3)^2 + (y - 3)^2]}{8}\right]}{f(D = 4)}$$

$$= K_4 \cdot e^{-\left[\frac{(4 - \sqrt{x^2 + y^2})^2}{2} + \frac{[(x - 3)^2 + (y - 3)^2]}{8}\right]}$$

For your notes...

# Tracking in 2D Space: Posterior



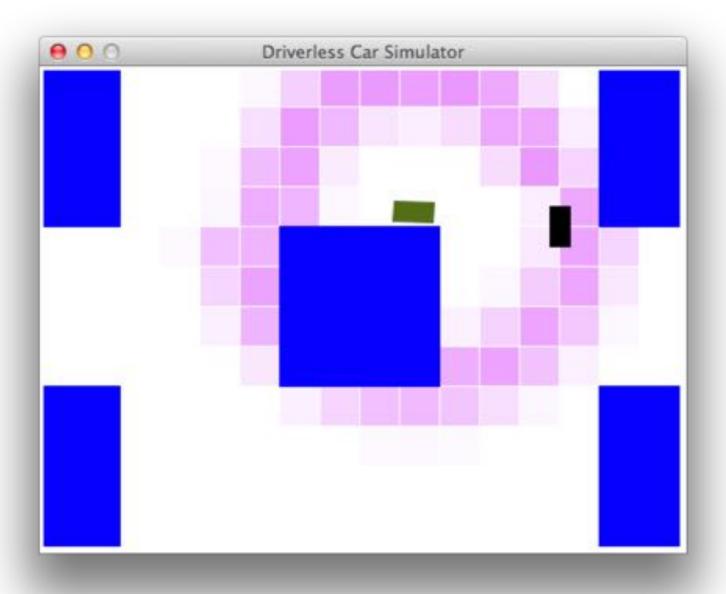


**Prior** 

**Posterior** 

$$= K_4 \cdot e^{-\left[\frac{(4-\sqrt{x^2+y^2})^2}{2} + \frac{[(x-3)^2+(y-3)^2]}{8}\right]}$$

# Tracking in 2D Space: CS221



# Expectation of Multiple RVs

# Joint Expectation

$$E[X] = \sum_{x} xp(x)$$

- Expectation over a joint isn't nicely defined because it is not clear how to compose the multiple variables:
  - Add them? Multiply them?
- Lemma: For a function g(X,Y) we can calculate the expectation of that function:

$$E[g(X,Y)] = \sum_{x,y} g(x,y)p(x,y)$$

Recall, this also holds for single random variables:

$$E[g(X)] = \sum g(x)p(x)$$

# **Expected Values of Sums**

Big deal lemma: first stated without proof



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

Generalized: 
$$E\left[\sum_{i=1}^{n} X_i\right] = \sum_{i=1}^{n} E[X_i]$$

Holds regardless of dependency between  $X_i$ 's

# **Skeptical Chris Wants a Proof!**

Let 
$$g(X,Y) = [X + Y]$$

$$E[X+Y] = E[g(X,Y)] = \sum_{x,y} g(x,y) p(x,y) \qquad \text{What a useful lemma}$$
 
$$= \sum_{x,y} [x+y] p(x,y) \qquad \text{By the definition of } g(x,y)$$
 
$$= \sum_{x,y} x p(x,y) + \sum_{x,y} y p(x,y)$$
 
$$\text{Change the sum of } (x,y) \text{ into separate sums} \qquad = \sum_{x} x \sum_{y} p(x,y) + \sum_{y} y \sum_{x} p(x,y)$$
 
$$= \sum_{x} x \sum_{y} p(x,y) + \sum_{y} y \sum_{x} p(x,y)$$
 That is the definition of marginal probability 
$$= \sum_{x} x p(x) + \sum_{y} y p(y)$$
 That is the definition of marginal probability 
$$= E[X] + E[Y]$$

expectation



# Independent Discrete Variables

 Two discrete random variables X and Y are called <u>independent</u> if:

$$p(x,y) = p_X(x)p_Y(y) \text{ for all } x, y$$
$$P(X = x, Y = y) = P(X = x) \cdot P(Y = y)$$

- Intuitively: knowing the value of X tells us nothing about the distribution of Y (and vice versa)
  - If two variables are <u>not</u> independent, they are called <u>dependent</u>
- Similar conceptually to independent events, but we are dealing with multiple <u>variables</u>
  - Keep your events and variables distinct (and clear)!

# Is Year Independent of Lunch?

	Joint Proba				
	Dining Hall	Eating Club	Cafe	Self-made	Marginal Year
Freshman	0.03	0.00	0.02	0.00	0.05
Sophomore	0.50	0.15	0.03	0.03	0.68
Junior	0.08	0.02	0.02	0.02	0.12
Senior	0.02	0.05	0.01	0.01	0.08
5+	0.02	0.01	0.05	0.05	0.07
Marginal Status	0.65	0.22	0.12	0.11	

For all values of Year, Status:

# Is Year Independent of Lunch?

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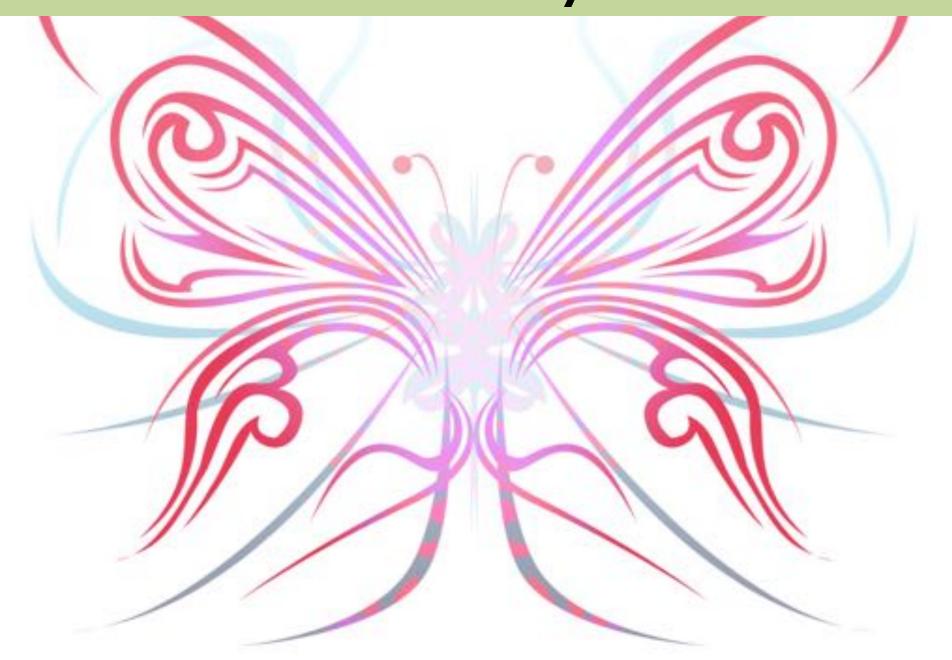
For all values of Year, Status:

P(Year = y, Lunch= s) = P(Year = y)P(Lunch = s)  

$$0.03$$
  $0.68$   $0.12$ 



# Aside: Butterfly Effect



# Coin Flips

- Flip coin with probability p of "heads"
  - Flip coin a total of n + m times
  - Let X = number of heads in first n flips
  - Let Y = number of heads in next m flips

$$P(X = x, Y = y) = \binom{n}{x} p^{x} (1-p)^{n-x} \binom{m}{y} p^{y} (1-p)^{m-y}$$

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

- X and Y are independent
- Let Z = number of total heads in n + m flips
- Are X and Z independent?
  - ⋄ What if you are told Z = 0?

# Independent Continuous Variables

 Two continuous random variables X and Y are called <u>independent</u> if:

$$P(X \le a, Y \le b) = P(X \le a) P(Y \le b)$$
 for any  $a, b$ 

Equivalently:

$$F_{X,Y}(a,b) = F_X(a)F_Y(b)$$
 for all  $a,b$   
 $f_{X,Y}(a,b) = f_X(a)f_Y(b)$  for all  $a,b$ 

More generally, joint density factors separately:

$$f_{XY}(x,y) = h(x)g(y)$$
 where  $-\infty < x, y < \infty$ 

# Is the Blur Distribution Independent?



In image processing, a Gaussian blur is the result of blurring an image by a Gaussian function. It is a widely used effect in graphics software, typically to reduce image noise.

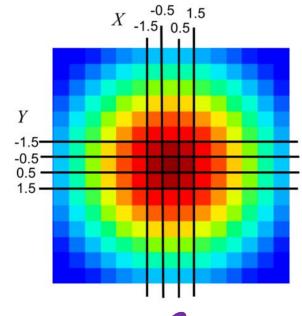
Gaussian blurring with StDev = 3, is based on a joint probability distribution:

#### **Joint PDF**

$$f_{X,Y}(x,y) = \frac{1}{2\pi \cdot 3^2} e^{-\frac{x^2 + y^2}{2 \cdot 3^2}}$$

#### **Joint CDF**

$$F_{X,Y}(x,y) = \Phi\left(\frac{x}{3}\right) \cdot \Phi\left(\frac{y}{3}\right)$$



Used to generate this weight matrix

# Pop Quiz (just kidding)

Consider joint density function of X and Y:

$$f_{X,Y}(x,y) = 6e^{-3x}e^{-2y}$$
 for  $0 < x, y < \infty$ 

Are X and Y independent? Yes!

Let 
$$h(x) = 3e^{-3x}$$
 and  $g(y) = 2e^{-2y}$ , so  $f_{X,Y}(x,y) = h(x)g(y)$ 

Consider joint density function of X and Y:

$$f_{X,Y}(x,y) = 4xy$$
 for  $0 < x, y < 1$ 

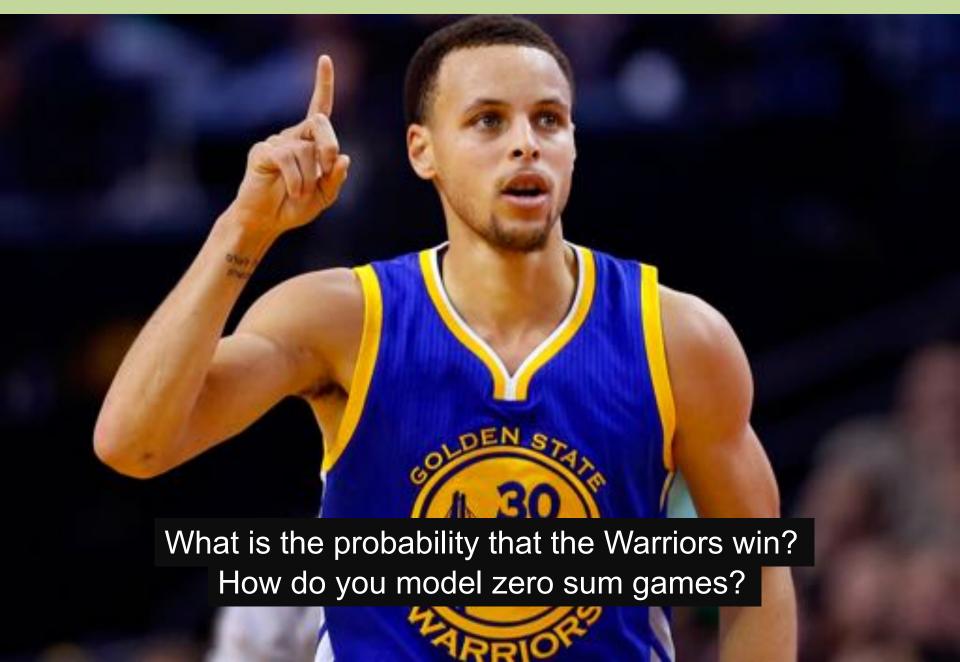
Are X and Y independent? Yes!

Let 
$$h(x) = 2x$$
 and  $g(y) = 2y$ , so  $f_{X,Y}(x,y) = h(x)g(y)$ 

- Now add constraint that: 0 < (x + y) < 1
- Are X and Y independent? No!
  - Cannot capture constraint on x + y in factorization!

# What happens when you add random variables?

### Zero Sum Games



# Motivating Idea: Zero Sum Games

### How it works:

- Each team has an "ELO" score S, calculated based on their past performance.
- Each game, the team has ability  $A \sim N(S, 200^2)$
- The team with the higher sampled ability wins.



Arpad Elo

$$A_B \sim \mathcal{N}(1555, 200^2)$$
  $A_W \sim \mathcal{N}(1797, 200^2)$ 

$$P(\text{Warriors win}) = P(A_W > A_B)$$

# Motivating Idea: Zero Sum Games

$$A_W \sim \mathcal{N}(1797, 200^2)$$

$$A_B \sim \mathcal{N}(1555, 200^2)$$

$$P(\text{Warriors win}) = P(A_W > A_B)$$

$$P(\text{Warriors win}) = P(A_W - A_B > 0)$$

In class we solved this by sampling. But that is a bit of a "cheat" and is computationally expensive.

Sums (or subtractions) of random variables show up all the time. But we have no explicit tools for dealing with them!

Challenge: try and come up with the way to solve this by the end of class

# Sum of Independent Binomials

- Let X and Y be independent binomials with the same value for p:
  - $X \sim Bin(n_1, p)$  and  $Y \sim Bin(n_2, p)$
  - $X + Y \sim Bin(n_1 + n_2, p)$
- Intuition:
  - X has n<sub>1</sub> trials and Y has n<sub>2</sub> trials
    - Each trial has same "success" probability p
  - Define Z to be n<sub>1</sub> + n<sub>2</sub> trials, each with success prob. p
  - $Z \sim Bin(n_1 + n_2, p)$ , and also Z = X + Y

# If only it were always that simple

# The Insight to Convolution

Imagine a game where each player *independently* scores between 0 and 100 points:

Let X be the amount of points you score. Let Y be the amount of points your opponent scores. Let's say you know P(X = x) and P(Y = y).

What is the probability of a tie?

$$P(\text{tie}) = \sum_{i=0}^{100} P(X = i, Y = i)$$
$$= \sum_{i=0}^{100} P(X = i)P(Y = i)$$

# The Insight to Convolution Proofs

What is the probability that X + Y = n?

$$P(X + Y = n)?$$

$$P(X + Y = n) = \sum_{i=0}^{n} P(X = i, Y = n - i)$$

X	Y	i	
0	n	0	P(X=0,Y=n)

1 
$$P(X = 1, Y = n - 1)$$

$$P(X = 2, Y = n - 2)$$

$$P(X=n,Y=0)$$

# The Insight to Convolution Proofs

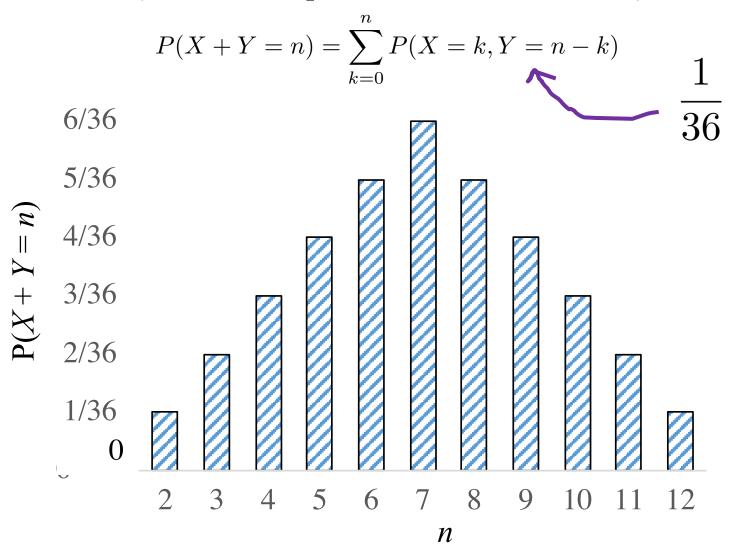
$$P(X+Y=n)?$$

Since this is the OR or mutually exclusive events 
$$P(X+Y=n) = \sum_{k=0}^n P(X=k,Y=n-k)$$

If the random variables are independent 
$$= \sum_{k=0}^{n} P(X=k)P(Y=n-k)$$

### Sum of Two Dice

Let *X*+*Y* be the value of the sum of two dice (aka two independent random variables)



# Sum of Independent Poissons

Recall the Binomial Theorem

$$(a+b)^n = \sum_{k=0}^n \binom{n}{k} a^k b^{n-k}$$

## Sum of Independent Poissons

- Let X and Y be independent random variables
  - $X \sim Poi(\lambda_1)$  and  $Y \sim Poi(\lambda_2)$
  - $X + Y \sim Poi(\lambda_1 + \lambda_2)$
- Proof: (just for reference)
  - Rewrite (X + Y = n) as (X = k, Y = n k) where  $0 \le k \le n$

$$P(X+Y=n) = \sum_{k=0}^{n} P(X=k, Y=n-k) = \sum_{k=0}^{n} P(X=k)P(Y=n-k)$$

$$=\sum_{k=0}^{n}e^{-\lambda_{1}}\frac{\lambda_{1}^{k}}{k!}e^{-\lambda_{2}}\frac{\lambda_{2}^{n-k}}{(n-k)!}=e^{-(\lambda_{1}+\lambda_{2})}\sum_{k=0}^{n}\frac{\lambda_{1}^{k}\lambda_{2}^{n-k}}{k!(n-k)!}=\frac{e^{-(\lambda_{1}+\lambda_{2})}}{n!}\sum_{k=0}^{n}\frac{n!}{k!(n-k)!}\lambda_{1}^{k}\lambda_{2}^{n-k}$$

- Noting Binomial theorem:  $(\lambda_1 + \lambda_2)^n = \sum_{k=0}^n \frac{n!}{k!(n-k)!} \lambda_1^k \lambda_2^{n-k}$   $P(X+Y=n) = \frac{e^{-(\lambda_1 + \lambda_2)}}{n!} (\lambda_1 + \lambda_2)^n$  so,  $X+Y=n \sim \text{Poi}(\lambda_1 + \lambda_2)$

## Reference: Sum of Independent RVs

- Let X and Y be independent Binomial RVs
  - $X \sim Bin(n_1, p)$  and  $Y \sim Bin(n_2, p)$
  - $X + Y \sim Bin(n_1 + n_2, p)$
  - More generally, let  $X_i \sim Bin(n_i, p)$  for  $1 \le i \le N$ , then

$$\left(\sum_{i=1}^{N} X_i\right) \sim \operatorname{Bin}\left(\sum_{i=1}^{N} n_i, p\right)$$

- Let X and Y be independent Poisson RVs
  - $X \sim Poi(\lambda_1)$  and  $Y \sim Poi(\lambda_2)$
  - $X + Y \sim Poi(\lambda_1 + \lambda_2)$
  - More generally, let  $X_i \sim Poi(\lambda_i)$  for  $1 \le i \le N$ , then

$$\left(\sum_{i=1}^{N} X_{i}\right) \sim \operatorname{Poi}\left(\sum_{i=1}^{N} \lambda_{i}\right)$$

# Sum of Independent Normals

- Let X and Y be independent random variables
  - $X \sim N(\mu_1, \sigma_1^2)$  and  $Y \sim N(\mu_2, \sigma_2^2)$
  - $X + Y \sim N(\mu_1 + \mu_2, \sigma_1^2 + \sigma_2^2)$

Generally, have n independent random variables
 X<sub>i</sub> ~ N(μ<sub>i</sub>, σ<sub>i</sub><sup>2</sup>) for i = 1, 2, ..., n:

$$\left(\sum_{i=1}^{n} X_i\right) \sim N\left(\sum_{i=1}^{n} \mu_i, \sum_{i=1}^{n} \sigma_i^2\right)$$

### Virus Infections

- Say you are working with the WHO to plan a response to a the initial conditions of a virus:
  - Two exposed groups
  - P1: 50 people, each independently infected with p = 0.1
  - P2: 100 people, each independently infected with p = 0.4
  - Question: Probability of more than 40 infections?

Sanity check: Should we use the Binomial Sum-of-RVs shortcut?

- A. YES!
- B. NO!
- C. Other/none/more

### Virus Infections

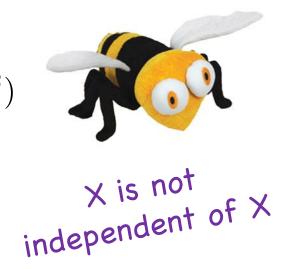
- Say you are working with the WHO to plan a response to a the initial conditions of a virus:
  - Two exposed groups
  - P1: 50 people, each independently infected with p = 0.1
  - P2: 100 people, each independently infected with p = 0.4
  - A = # infected in P1 A ~ Bin(50, 0.1)  $\approx$  X ~ N(5, 4.5)
  - B = # infected in P2 B ~ Bin(100, 0.4)  $\approx$  Y ~ N(40, 24)
  - What is P(≥ 40 people infected)?
  - $P(A + B \ge 40) \approx P(X + Y \ge 39.5)$
  - $X + Y = W \sim N(5 + 40 = 45, 4.5 + 24 = 28.5)$

$$P(W \ge 39.5) = P\left(\frac{W - 45}{\sqrt{28.5}} > \frac{39.5 - 45}{\sqrt{28.5}}\right) = 1 - \Phi(-1.03) \approx 0.8485$$

### **Linear Transform**

$$X \sim N(\mu, \sigma^2)$$
 
$$Y = X + X = 2 \cdot X$$
 
$$Y \sim N(2\mu, 4\sigma^2)$$

$$Y = X + X = 2 \cdot X$$
 
$$X + X \sim N(\mu + \mu, \sigma^2 + \sigma^2)$$
 
$$Y \sim N(2\mu, 2\sigma^2)$$
 x is



## Motivating Idea: Zero Sum Games

#### How it works:

- Each team has an "ELO" score S, calculated based on their past performance.
- Each game, the team has ability  $A \sim N(S, 200^2)$
- The team with the higher sampled ability wins.



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$$A_B \sim \mathcal{N}(1555, 200^2)$$
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## Motivating Idea: Zero Sum Games

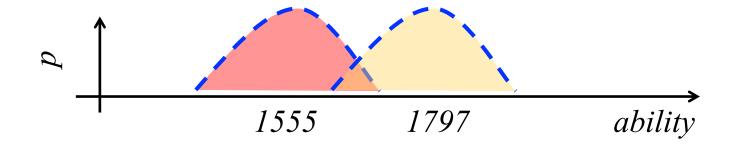
$$A_B \sim \mathcal{N}(1555, 200^2)$$
  $A_W \sim \mathcal{N}(1797, 200^2)$   $P(\text{Warriors win}) = P(A_W > A_B)$   $= P(A_W - A_B > 0)$ 

$$D = A_W - A_B$$

$$D \sim N(\mu = 1795 - 1555, \sigma_2 = 2 \cdot 200^2)$$

$$\sim N(\mu = 240, \sigma_2 = 283)$$

$$P(D > 0) = F_D(0) = 1 - \Phi\left(\frac{0 - 240}{283}\right) \approx 0.65$$



### Convolution of Probability Distributions



We talked about sum of Binomial, Normal and Poisson...who's missing from this party?

Uniform.

# Summation: not just for the 1%

### Dance, Dance Convolution

Let X and Y be independent random variables

Probability Density Function (PDF) of X + Y:

$$f_{X+Y}(a) = \int_{y=-\infty}^{\infty} f_X(a-y) f_Y(y) dy$$

• In discrete case, replace  $\int_{y=-\infty}^{\infty}$  with  $\sum_{y}$ , and f(y) with p(y)

### Dance, Dance Convolution

Let X and Y be independent random variables

Cumulative Distribution Function (CDF) of X + Y:

$$F_{X+Y}(a) = P(X+Y \le a)$$

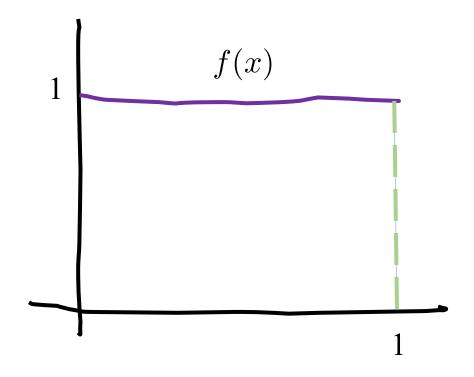
$$+ = \iint_{x+y \le a} f_X(x) f_Y(y) dx dy = \int_{y=-\infty}^{\infty} \int_{x=-\infty}^{a-y} f_X(x) dx f_Y(y) dy$$

$$= \int_{y=-\infty}^{\infty} F_X(a-y) f_Y(y) dy$$
PDF of Y

• In discrete case, replace  $\int_{y=-\infty}^{\infty}$  with  $\sum_{y}$ , and f(y) with p(y)

## Sum of Independent Uniforms

- Let X and Y be independent random variables
  - X ~ Uni(0, 1) and Y ~ Uni(0, 1)  $\rightarrow f(x) = 1$  for  $0 \le x \le 1$



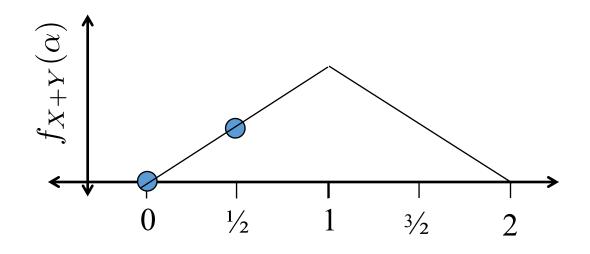
For both X and Y

### $1 < \alpha < 2$

 $X \sim \mathrm{Uni}(0,1)$   $Y \sim \mathrm{Uni}(0,1)$ X and Y are independent

$$f_{X+Y}(\alpha)$$
?

$$f_{X+Y}(a) = \begin{cases} a & 0 \le a \le 1\\ 2-a & 1 < a \le 2\\ 0 & \text{otherwise} \end{cases}$$



# That's all folks!