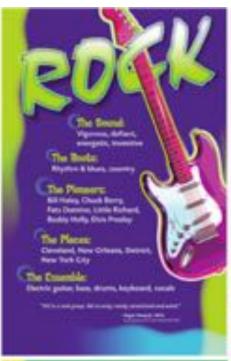


Your random variables are correlated

Covariance and Correlation

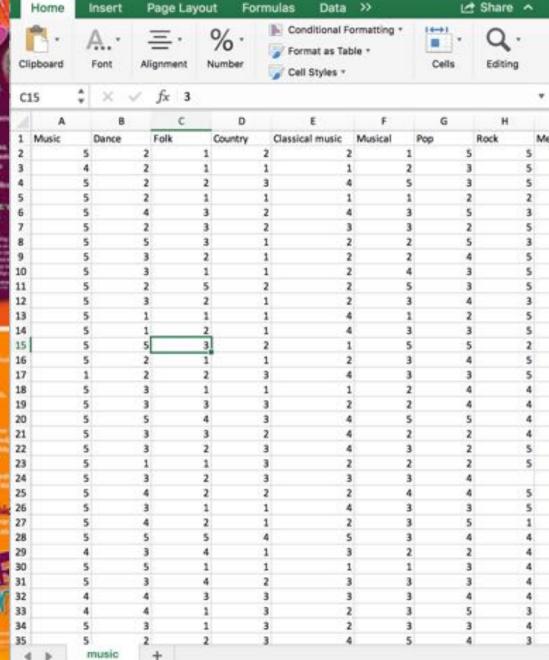
Chris Piech CS109, Stanford University







Ready



---+

100%

8

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п



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?



How do multiple variables **covary**?

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 ~ Poi(λ_i) for 1 ≤ i ≤ N, then

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

Convolution of Probability Distributions



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

Uniform, Normal.

Summation: not just for the 1%

What about the general case?

Were talking about the sum of uniforms

```
sum.py
   import random
  def main():
     x = random.random()
5
6
     y = random.random()
    z = x + y
     print(z)
  if __name__ == '__main
     main()
```

The Insight to Convolution Proofs

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

$$\begin{array}{c|cccc} X & Y & \mathbf{k} \\ \hline 0 & \mathbf{n} & 0 & P(X=0,Y=n) \end{array}$$

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

P(X=n,Y=0)

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

 \mathbf{n}

The Insight to Convolution Proofs

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

$$f(X+Y=\alpha) = \int_{k=-\infty}^{\infty} f(X=k, Y=\alpha-k) dk$$

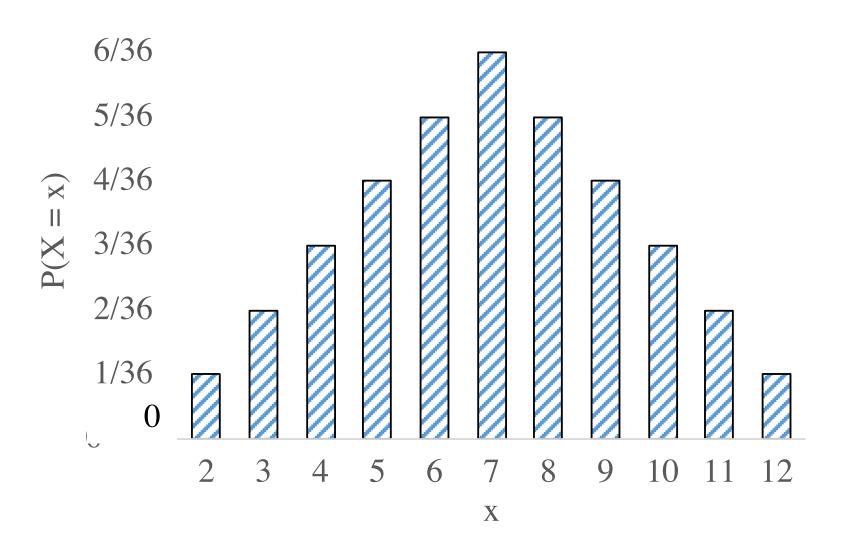
The Insight to Convolution Proofs

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

$$f_{X+Y}(\alpha) = \int_{k=-\infty}^{\infty} f(X=k, Y=\alpha-k) dk$$

Sum of Two Dice

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



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

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

$$f_{X+Y}(\alpha) = \int_{k=-\infty}^{\infty} f(X=k, Y=\alpha-k) \ dk$$

$$f_{X+Y}(\alpha) = \int_{k=-\infty}^{\infty} f(X=k)f(Y=\alpha-k) dk$$

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

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

$$f_{X+Y}(\alpha) = \int_{k=-\infty}^{\infty} f(X=k)f(Y=\alpha-k) dk$$

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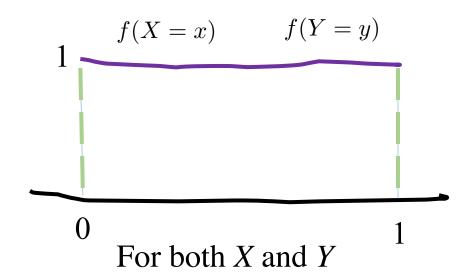
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?

$$f_{X+Y}(\alpha) = \int_{k=-\infty}^{\infty} f(X=k)f(Y=\alpha-k) dk$$

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$$f_{X+Y}(\alpha) = \int_{k=-\infty}^{\infty} f(X=k)f(Y=\alpha-k) dk$$



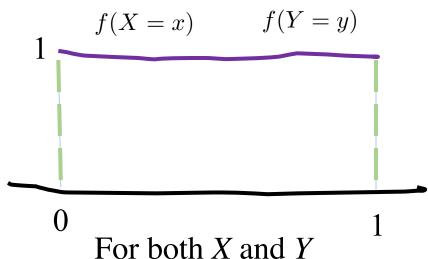
$$X \sim \mathrm{Uni}(0,1)$$
 $Y \sim \mathrm{Uni}(0,1)$
 X and Y are independent

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

$$f_{X+Y}(\alpha) = \int_{k=-\infty}^{\infty} f(X=k)f(Y=\alpha-k) dk$$

ISNT THIS JUST ONE!?!?!!!??!



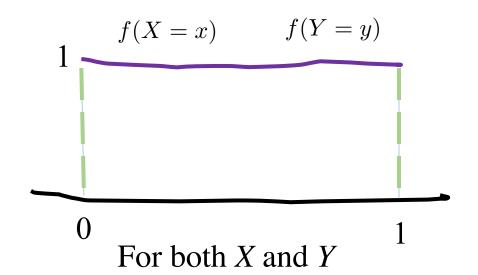


$$X \sim \mathrm{Uni}(0,1)$$
 $Y \sim \mathrm{Uni}(0,1)$
 X and Y are independent

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

$$f_{X+Y}(\alpha) = \int_{k=-\infty}^{\infty} f(X=k)f(Y=\alpha-k) dk$$

$$0 < k < 1 \qquad \qquad 0 < \alpha - k < 1$$



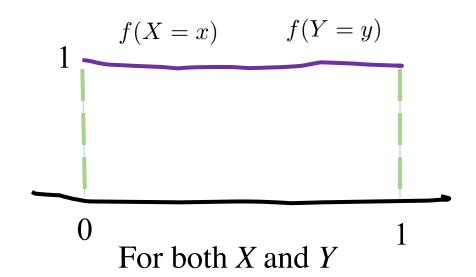
$$X \sim \mathrm{Uni}(0,1)$$
 $Y \sim \mathrm{Uni}(0,1)$
 X and Y are independent

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

$$f_{X+Y}(\alpha) = \int_{k=-\infty}^{\infty} f(X=k)f(Y=\alpha-k) dk$$

For these values of k, the densities of f_X and f_Y are 1

$$0 < k < 1 \qquad -\alpha < -k < 1 - \alpha$$



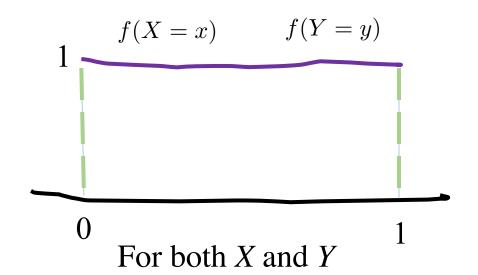
$$X \sim \mathrm{Uni}(0,1)$$
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of k, the densities of f_X and f_Y are 1

$$0 < k < 1 \qquad \qquad \alpha - 1 < k < \alpha$$



$$\alpha = \frac{1}{2}$$

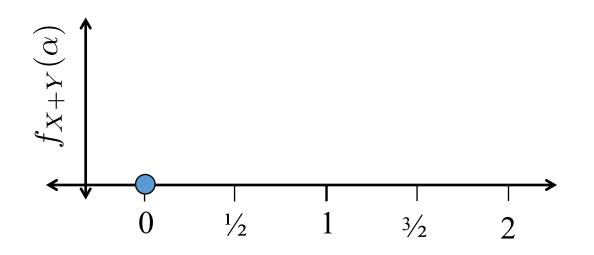
 $X \sim \mathrm{Uni}(0,1)$ $Y \sim \mathrm{Uni}(0,1)$ X and Y are independent

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

$$f_{X+Y}(\alpha) = \int_{k=-\infty}^{\infty} f(X=k)f(Y=\alpha-k) dk$$

For these values of k, the densities of f_X and f_Y are 1

$$0 < k < 1 \qquad \qquad \alpha - 1 < k < \alpha$$



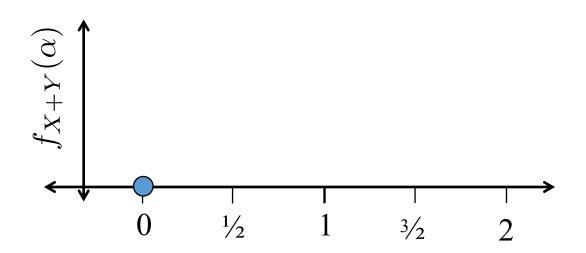
$$\alpha = \frac{1}{2}$$

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

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

$$f_{X+Y}(1/2) = \int_{k=-\infty}^{\infty} f(X=k)f(Y=1/2-k) dk$$
 $\alpha = 1/2$

$$0 < k < 1 \qquad \qquad \alpha - 1 < k < \alpha$$



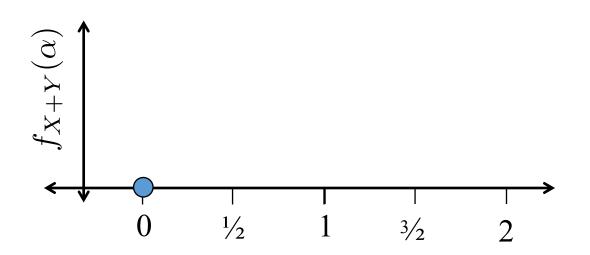
$$\alpha = \frac{1}{2}$$

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

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

$$f_{X+Y}(1/2) = \int_{k=-\infty}^{\infty} f(X=k)f(Y=1/2-k) dk$$
 $\alpha = 1/2$

$$0 < k < 1 \qquad -1/2 < k < 1/2$$



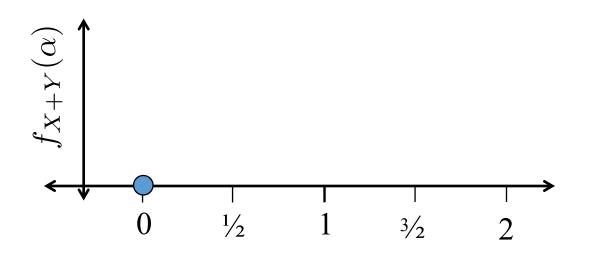
$$\alpha = \frac{1}{2}$$

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

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

$$f_{X+Y}(1/2) = \int_{k=0}^{1/2} f(X=k)f(Y=1/2-k) dk \qquad \alpha = 1/2$$

$$0 < k < 1$$
 $-1/2 < k < 1/2$



$$\alpha = \frac{1}{2}$$

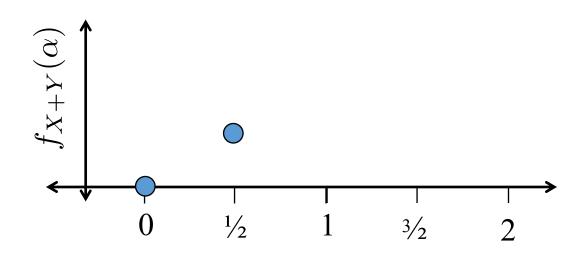
$$X \sim \mathrm{Uni}(0,1)$$
 $Y \sim \mathrm{Uni}(0,1)$
 X and Y are independent

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

$$f_{X+Y}(1/2) = \int_{k=0}^{1/2} 1 \ dk = 0.5$$

 $\alpha = 1/2$

$$0 < k < 1 \qquad -1/2 < k < 1/2$$

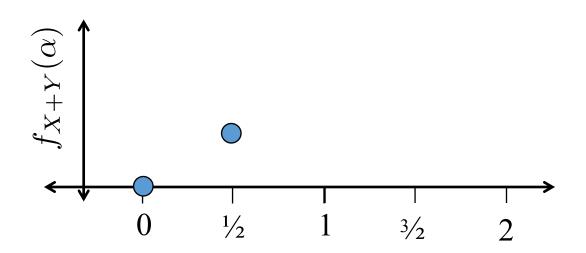


 $X \sim \mathrm{Uni}(0,1)$ $Y \sim \mathrm{Uni}(0,1)$ X and Y are independent

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

$$f_{X+Y}(\alpha) = \int_{k=-\infty}^{\infty} f(X=k)f(Y=\alpha-k) dk$$

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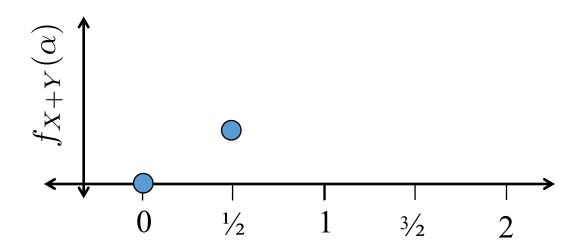


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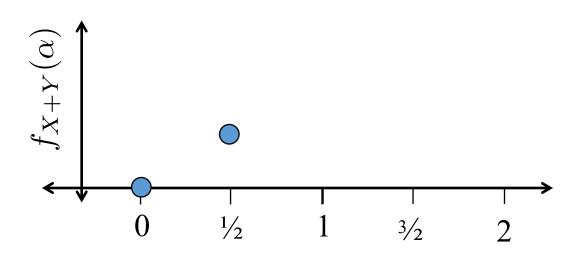
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 $\alpha - 1 < k < \alpha$

$$0 < k < \alpha$$



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

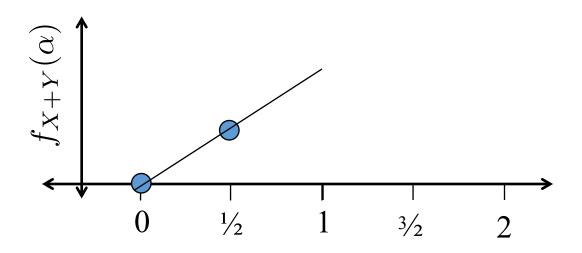
$$f_{X+Y}(\alpha)$$
?

$$f_{X+Y}(\alpha) = \int_{k=0}^{\alpha} 1 \ dk = \alpha$$

For these values of k, the densities are 1

$$0 < k < 1$$
 $\alpha - 1 < k < \alpha$

 $0 < k < \alpha$

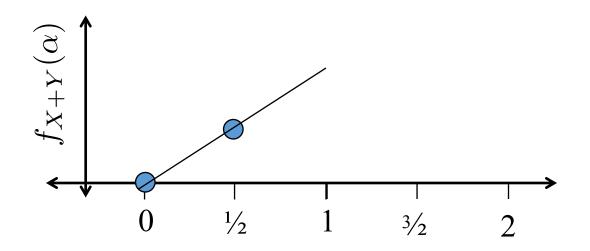


 $X \sim \mathrm{Uni}(0,1)$ $Y \sim \mathrm{Uni}(0,1)$ X and Y are independent

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

$$f_{X+Y}(\alpha) = \int_{k=-\infty}^{\infty} f(X=k)f(Y=\alpha-k) dk$$

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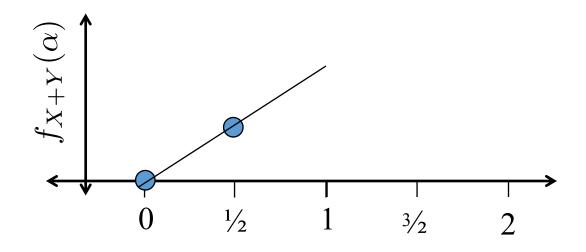


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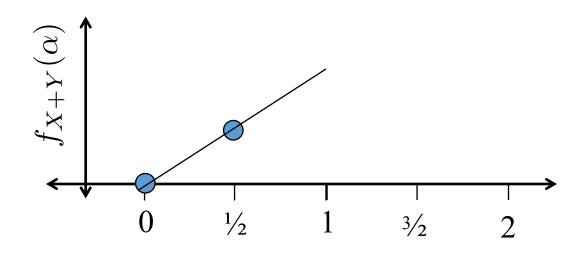
 $X \sim \mathrm{Uni}(0,1)$ $Y \sim \mathrm{Uni}(0,1)$ X and Y are independent

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

$$f_{X+Y}(\alpha) = \int_{k=\alpha-1}^{1} f(X=k)f(Y=\alpha-k) dk$$

$$0 < k < 1$$
 $\alpha - 1 < k < \alpha$

$$\alpha - 1 < k < 1$$



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

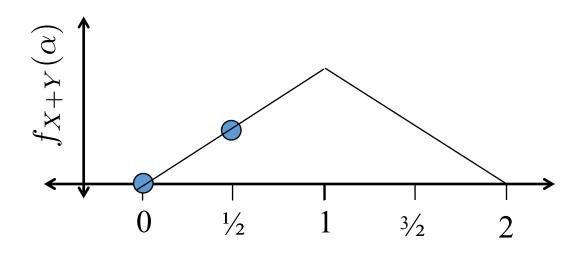
$$f_{X+Y}(\alpha)$$
?

$$f_{X+Y}(\alpha) = \int_{k=\alpha-1}^{1} 1 \ dk = 2 - \alpha$$

$$0 < k < 1$$
 $\alpha - 1 < k < \alpha$

$$\alpha - 1 < k < \alpha$$

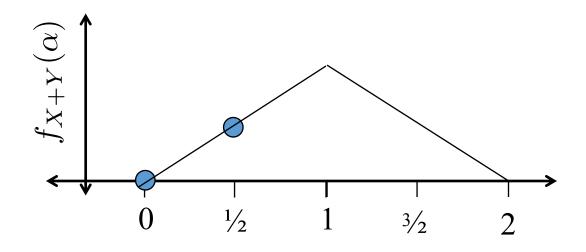
$$\alpha - 1 < k < 1$$



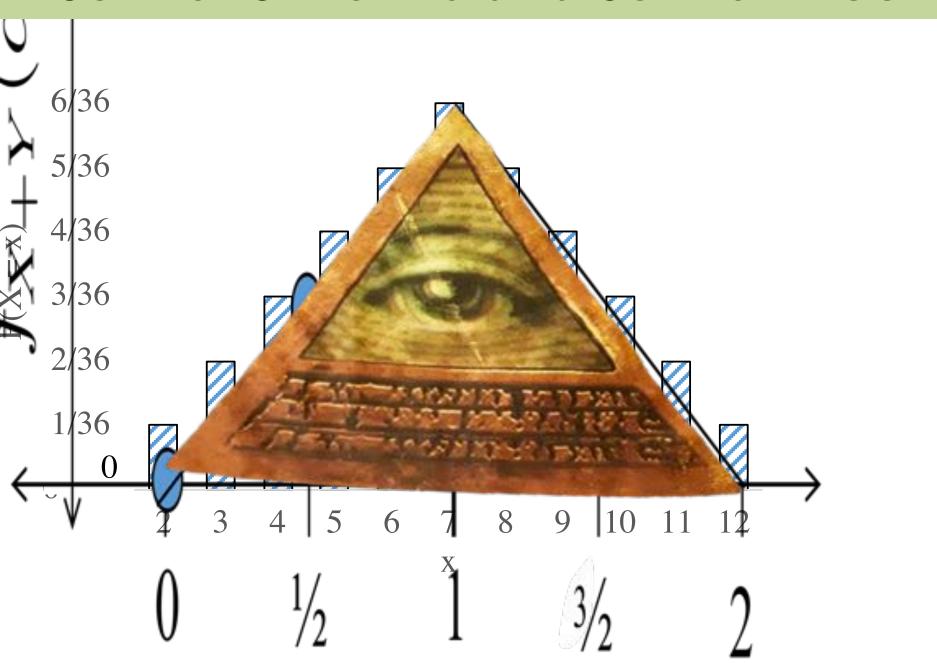
 $X \sim \mathrm{Uni}(0,1)$ $Y \sim \mathrm{Uni}(0,1)$ X and Y are independent

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

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



Sum of Uniforms and Sum of Dice



That was hard...

Ready for something easy and truly useful?

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_i ~ N(μ_i, σ_i²) 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
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 - Question: Probability of more than 40 infections?

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

$$X \sim N(\mu, \sigma^2)$$

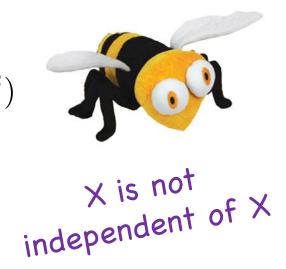
$$Y = X + X = 2 \cdot X$$

$$Y \sim N(2\mu, 4\sigma^2)$$

$$Y = X + X = 2 \cdot X$$

$$X + X \sim N(\mu + \mu, \sigma^2 + \sigma^2)$$

$$Y \sim N(2\mu, 2\sigma^2)$$
 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.



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

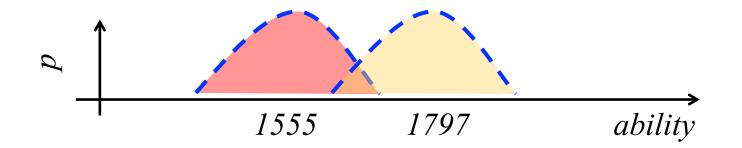
$$A_B \sim \mathcal{N}(1555, 200^2)$$
 $A_W \sim \mathcal{N}(1797, 200^2)$ $P(\text{Warriors win}) = P(A_W > A_B)$ $= P(A_W - A_B > 0)$

$$D = A_W - A_B$$

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

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

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



Dance of Covariance

Recall our Ebola Bats



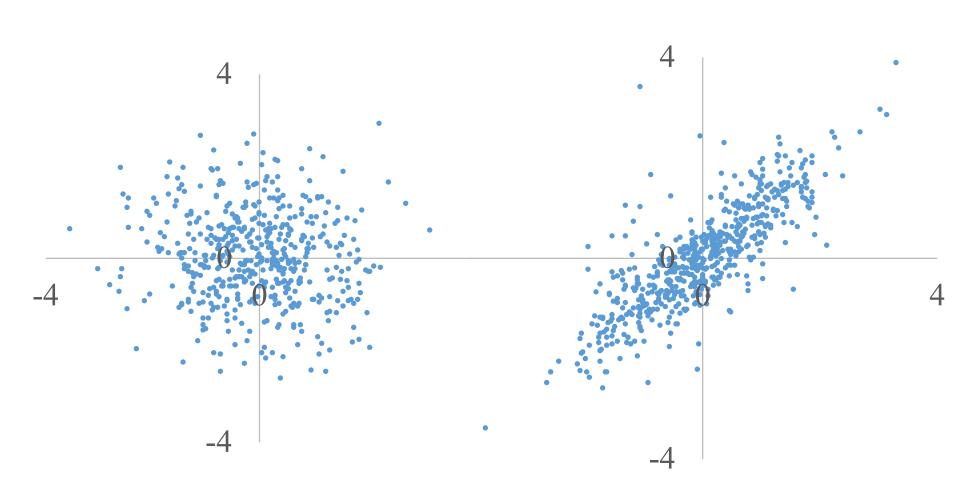
Bat Data

_						
	Gene1	Gene2	Gene3	Gene4	Gene5	Trait
	TRUE	FALSE	TRUE	TRUE	FALSE	FALSE
	FALSE	FALSE	TRUE	TRUE	TRUE	TRUE
	TRUE	FALSE	TRUE	FALSE	FALSE	FALSE
	TRUE	FALSE	TRUE	TRUE	TRUE	FALSE
	FALSE	TRUE	TRUE	TRUE	TRUE	TRUE
	FALSE	FALSE	FALSE	TRUE	FALSE	FALSE
	TRUE	FALSE	FALSE	TRUE	FALSE	FALSE
	TRUE	FALSE	FALSE	TRUE	FALSE	FALSE
	TRUE	FALSE	TRUE	FALSE	FALSE	FALSE
	FALSE	TRUE	FALSE	TRUE	FALSE	FALSE
	TRUE	TRUE	FALSE	TRUE	FALSE	FALSE
	TRUE	FALSE	FALSE	TRUE	FALSE	FALSE
	TRUE	FALSE	TRUE	TRUE	TRUE	FALSE
	FALSE	FALSE	TRUE	TRUE	FALSE	FALSE
	TRUE	FALSE	FALSE	TRUE	FALSE	FALSE
	TRUE	FALSE	FALSE	TRUE	FALSE	FALSE
			•	·•		
_	TRUE	FALSE	FALSE	TRUE	FALSE	FALSE

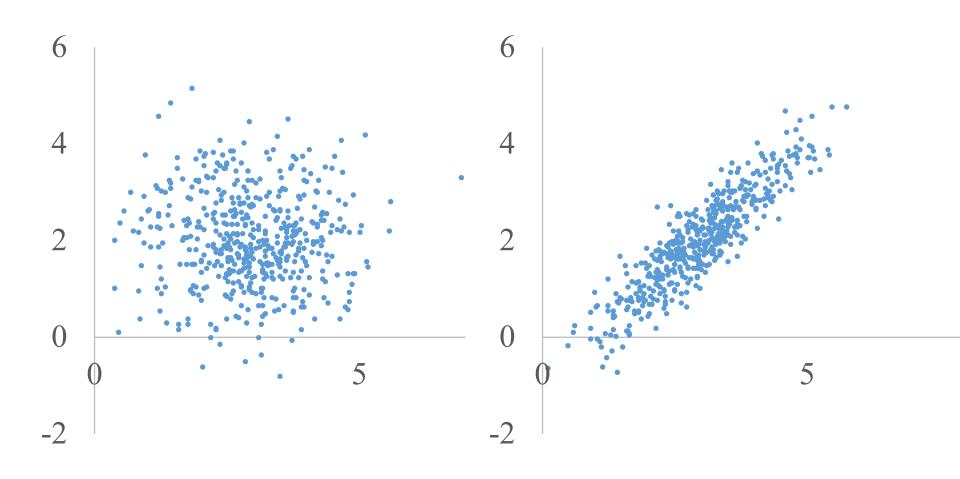
Expression Amount

Gene5	Trait
0.76	0.83
0.94	0.85
0.82	0.03
0.94	0.32
0.50	0.10
0.40	0.53
0.90	0.67
0.29	0.71
0.72	0.25
0.15	0.24
0.79	0.98
0.68	0.77
0.71	0.37
0.36	0.18
0.62	0.08
0.59	0.38
0.82	0.76

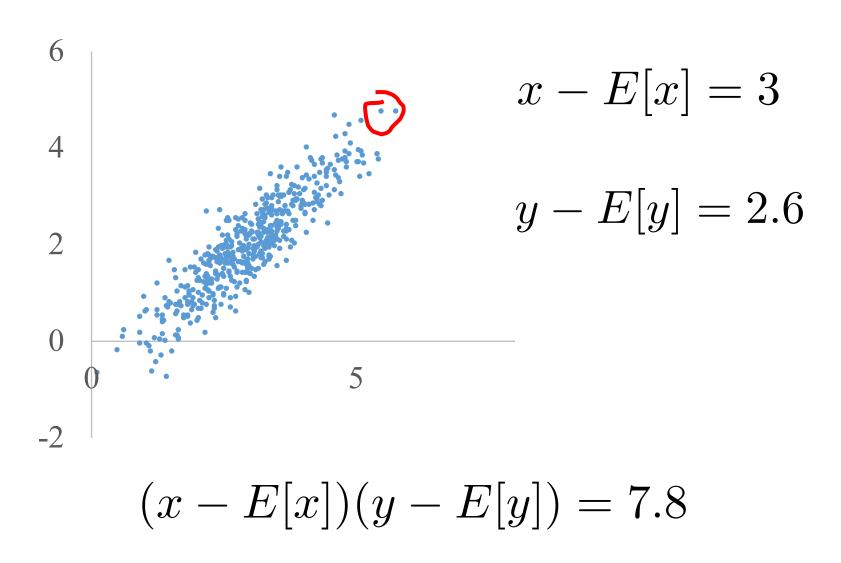
Spot The Difference



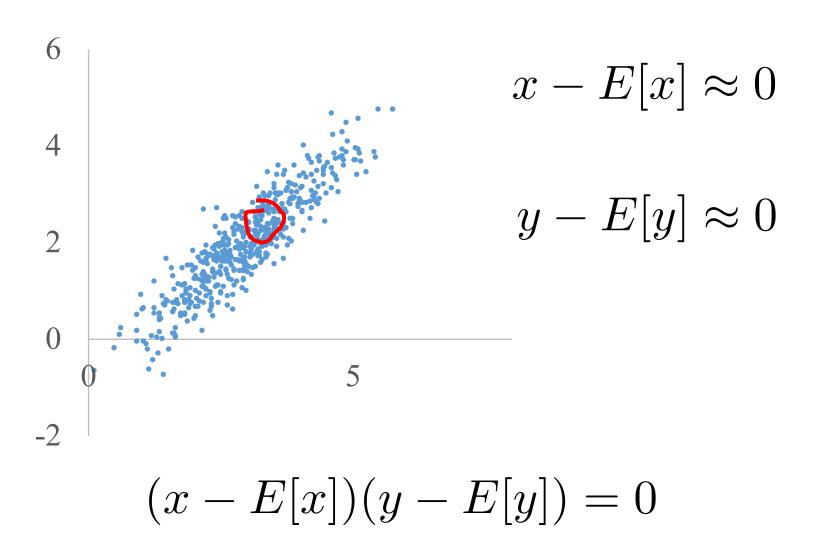
Spot The Difference



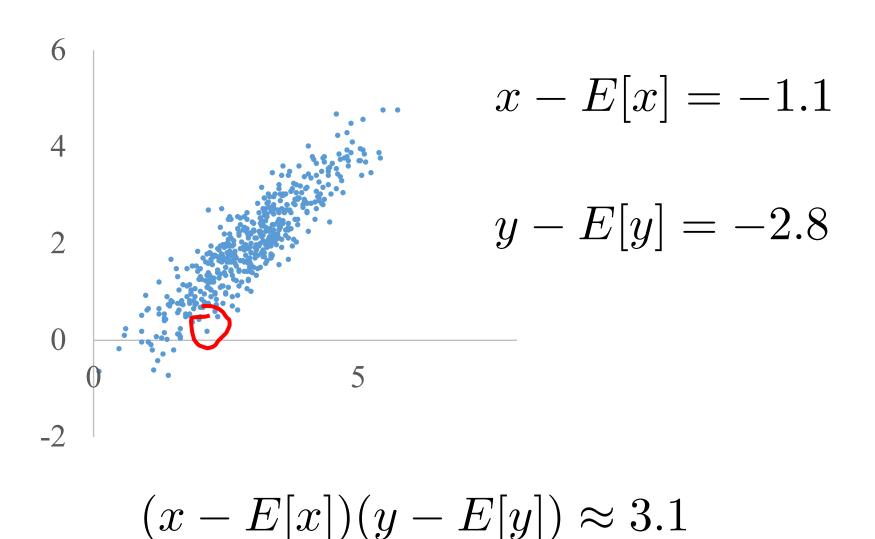
Vary Together



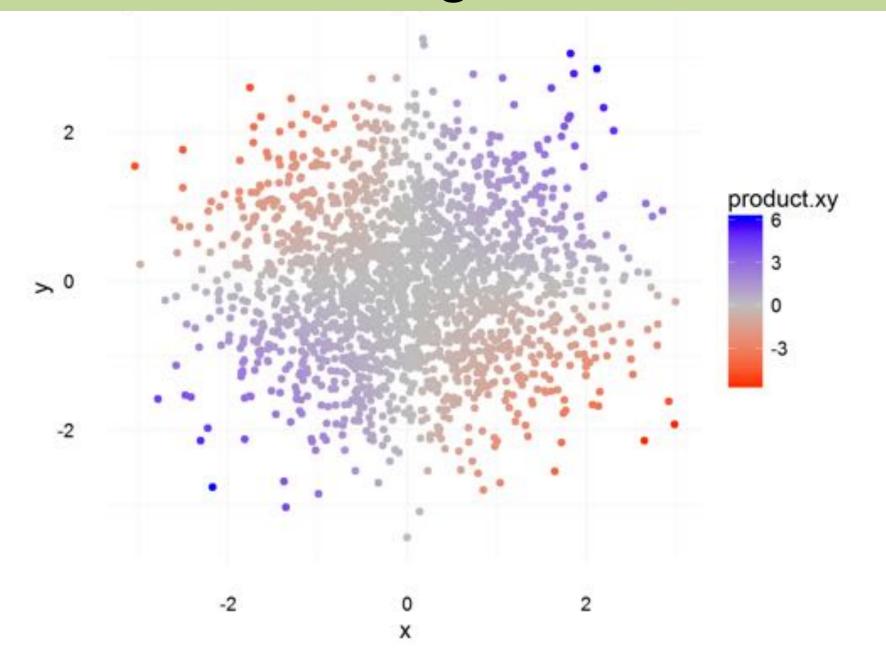
Vary Together



Vary Together



Understanding Covariance



The Dance of the Covariance

- Say X and Y are arbitrary random variables
- Covariance of X and Y:

$$Cov(X, Y) = E[(X - E[X])(Y - E[Y])]$$

X	У	(x - E[X])(y - E[Y])p(x,y)
Above mean	Above mean	Positive
Bellow mean	Bellow mean	Positive
Bellow mean	Above mean	Negative
Above mean	Bellow mean	Negative

The Dance of the Covariance

- Say X and Y are arbitrary random variables
- Covariance of X and Y:

$$Cov(X, Y) = E[(X - E[X])(Y - E[Y])]$$

Equivalently:

$$Cov(X,Y) = E[XY - E[X]Y - XE[Y] + E[Y]E[X]]$$

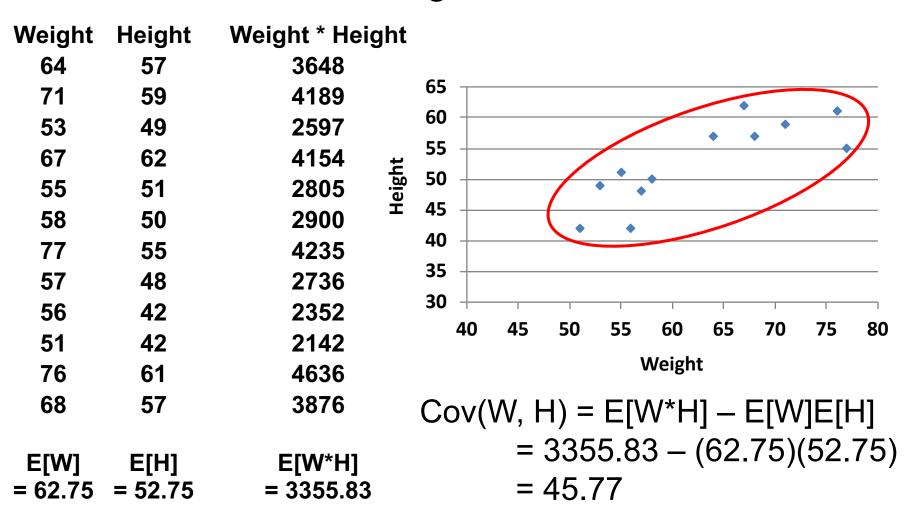
$$= E[XY] - E[X]E[Y] - E[X]E[Y] + E[X]E[Y]$$

$$= E[XY] - E[X]E[Y]$$

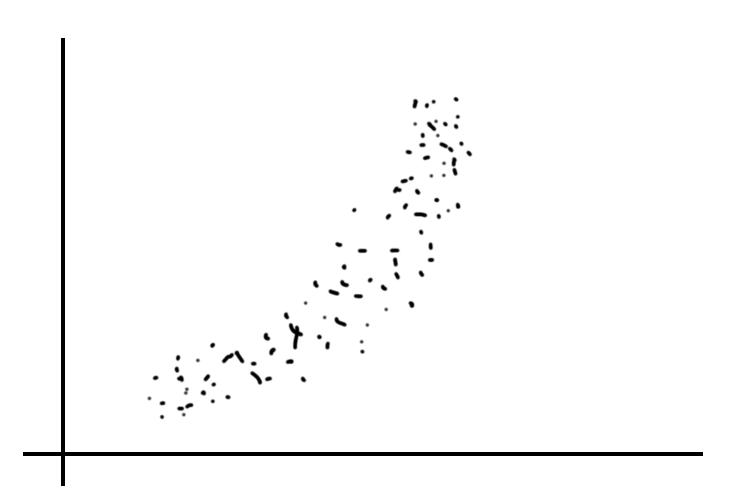
- X and Y independent, E[XY] = E[X]E[Y] → Cov(X,Y) = 0
- But Cov(X,Y) = 0 does <u>not</u> imply X and Y independent!

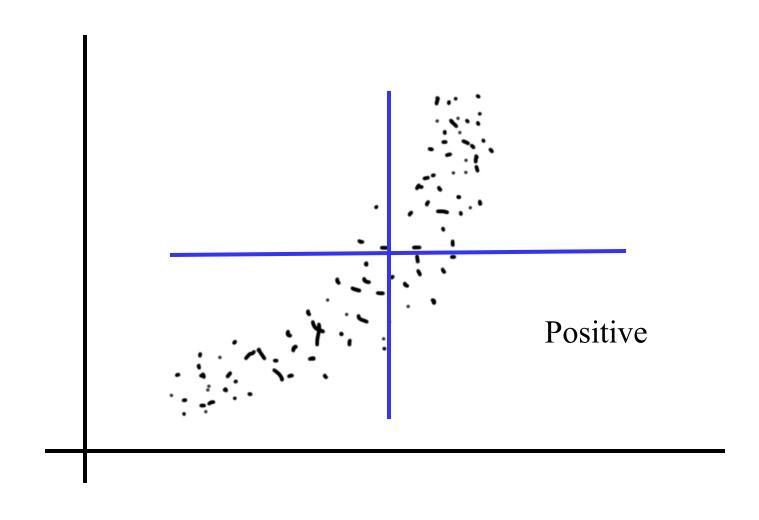
Covariance and Data

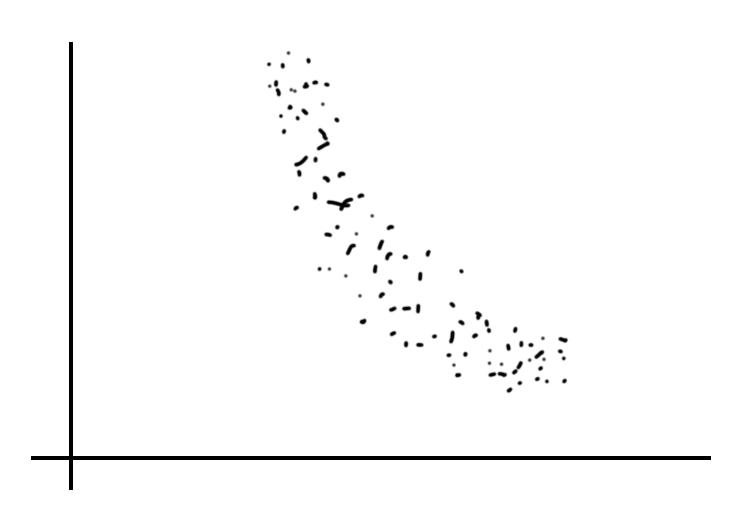
Consider the following data:

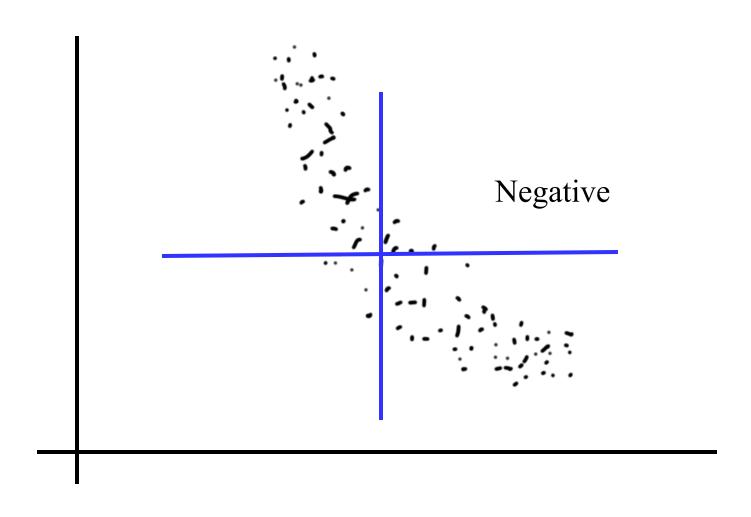


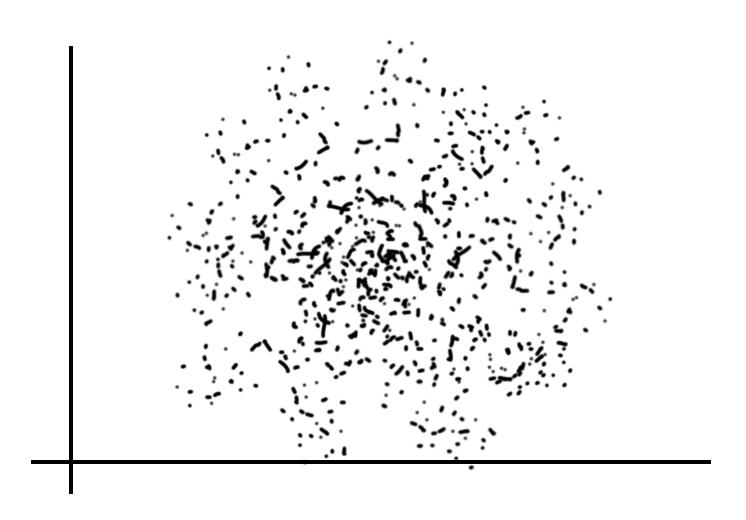
Poll: (a) positive, (b) negative, (c) zero

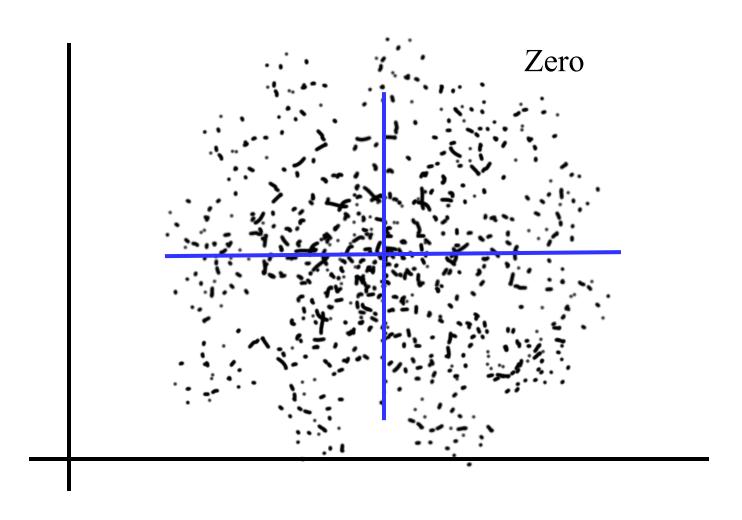












Independence and Covariance

X and Y are random variables with PMF:

YX	-1	0	1	$p_{Y}(y)$	
0	1/3	0	1/3	2/3	$Y = \begin{cases} 0 \end{cases}$
1	0	1/3	0	1/3	1 - 1
$p_X(x)$	1/3	1/3	1/3	1	

if $X \neq 0$

otherwise

•
$$E[X] = -1(1/3) + 0(1/3) + 1(1/3) = 0$$

•
$$E[Y] = 0(2/3) + 1(1/3) = 1/3$$

- Since XY = 0, E[XY] = 0
- Cov(X, Y) = E[XY] E[X]E[Y] = 0 0 = 0
- But, X and Y are clearly dependent!

Properties of Covariance

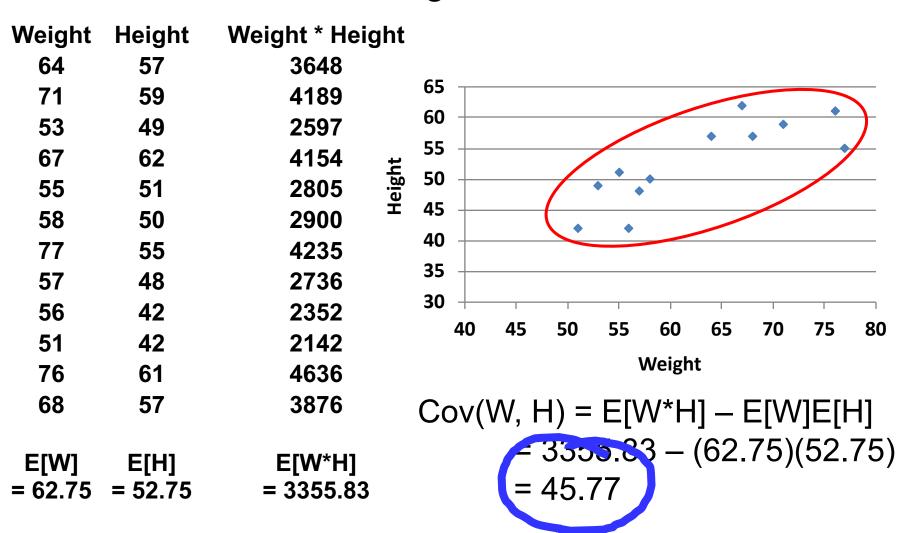
- Say X and Y are arbitrary random variables
 - Cov(X,Y) = Cov(Y,X)
 - $Cov(X, X) = E[X^2] E[X]E[X] = Var(X)$
 - Cov(aX + b, Y) = aCov(X, Y)
- Covariance of sums of random variables
 - $X_1, X_2, ..., X_n$ and $Y_1, Y_2, ..., Y_m$ are random variables

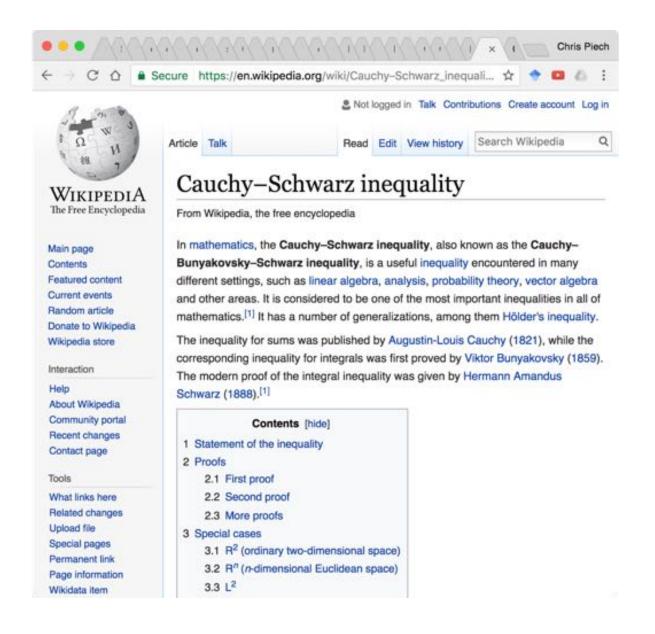
•
$$\operatorname{Cov}\left(\sum_{i=1}^{n} X_{i}, \sum_{j=1}^{m} Y_{j}\right) = \sum_{i=1}^{n} \sum_{j=1}^{m} \operatorname{Cov}(X_{i}, Y_{j})$$

Correlation

What is Wrong With This?

Consider the following data:





$-\mathrm{Std}(X)\mathrm{Std}(Y) \le \mathrm{Cov}(X,Y) \le \mathrm{Std}(X)\mathrm{Std}(Y)$

Viva La Correlatión

- Say X and Y are arbitrary random variables
 - Correlation of X and Y, denoted $\rho(X, Y)$:

$$\rho(X,Y) = \frac{\text{Cov}(X,Y)}{\sqrt{\text{Var}(X)\text{Var}(Y)}}$$

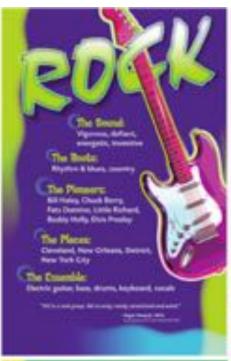
- Note: $-1 \le \rho(X, Y) \le 1$
- Correlation measures <u>linearity</u> between X and Y
- $\rho(X, Y) = 1$ $\Rightarrow Y = aX + b$ where $a = \sigma_y/\sigma_x$
- $\rho(X, Y) = -1$ $\Rightarrow Y = aX + b$ where $a = -\sigma_y/\sigma_x$
- $\rho(X, Y) = 0$ \Rightarrow absence of <u>linear</u> relationship
 - But, X and Y can still be related in some other way!
- If $\rho(X, Y) = 0$, we say X and Y are "uncorrelated"
 - Note: Independence implies uncorrelated, but <u>not</u> vice versa!

Viva La Correlatión

- Say X and Y are arbitrary random variables
 - Correlation of X and Y, denoted $\rho(X, Y)$:

$$\rho(X,Y) = \frac{\text{Cov}(X,Y)}{\sqrt{\text{Var}(X)\text{Var}(Y)}}$$

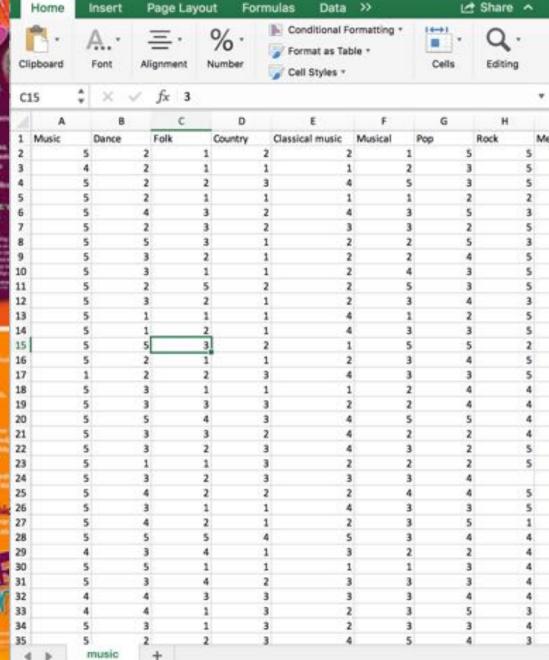
Say Y = cX. Correlation should be 1.







Ready



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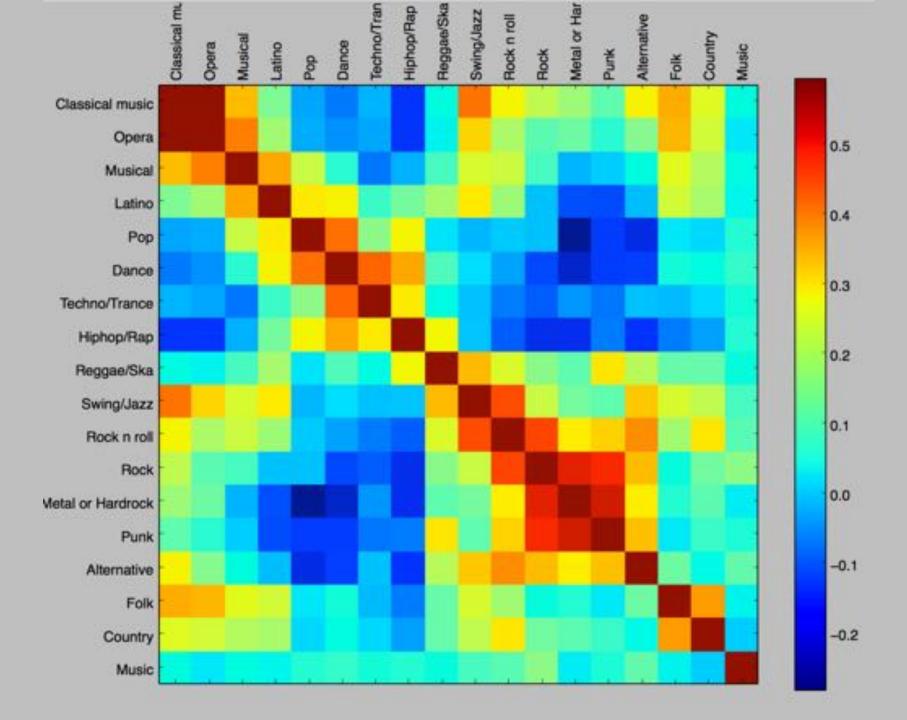
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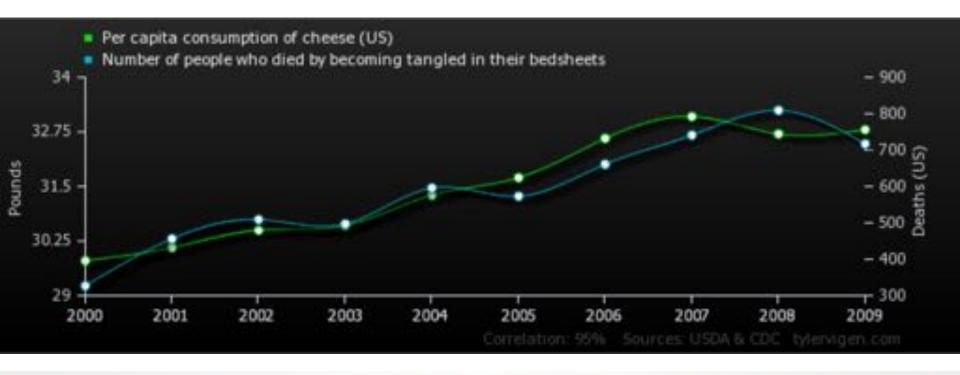
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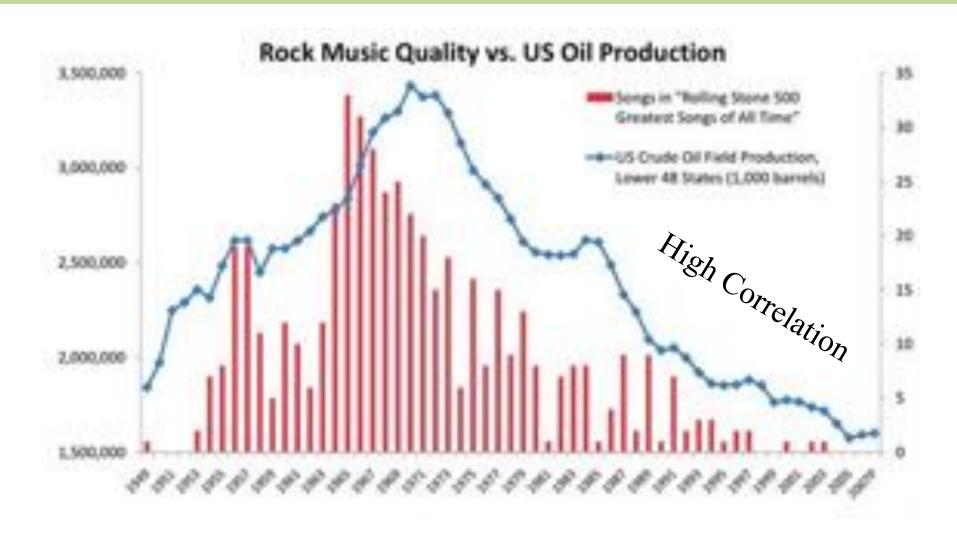
Tell your friends!



	2000	2001	2002	2003	2004	2005	2006	2007	2008	2009
Per capita consumption of cheese (US) Pounds (USDA)	29.8	30.1	30.5	30.6	31.3	31.7	32.6	33.1	32.7	32.8
Number of people who died by becoming tangled in their bedsheets Deaths (US) (CDC)		456	509	497	596	573	661	741	809	717

Correlation: 0.947091

Rock Music Vs Oil?



Hubbert Peak Theory

http://www.aei.org/publication/blog/

Divorce Vs Butter?

