

# MATH 180A: INTRO TO PROBABILITY (FOR DATA SCIENCE)

[www.math.ucsd.edu/~tkemp/180A](http://www.math.ucsd.edu/~tkemp/180A)

Today: § 8.3-8.4, 9.2

Next: § 9.3

Homework 7 Due **TONIGHT!**

Lab 7 Due Wednesday, Dec 4.

Homework 8 Due Friday, Dec 6.

Final Exam: Monday, Dec 9, 11:30a-2:30p  
in **REC GYM**

# Flashback to Math 20C

Given two vectors  $\underline{v}, \underline{w}$  in  $\mathbb{R}^n$ , their dot product is

$$\underline{v} \cdot \underline{w} =$$

It is a positive bilinear form:

(1)  $(a\underline{u} + b\underline{v}) \cdot \underline{w}$

(2)  $\underline{v} \cdot (a\underline{u} + b\underline{w})$

(3)  $\underline{v} \cdot \underline{v}$

The length<sup>2</sup> of a vector is  $\|\underline{v}\|^2$

Two vectors are orthogonal if

Cauchy-Schwarz Inequality:  $|\underline{v} \cdot \underline{w}|$

# The Geometry of Random Variables

Dot product  $\rightsquigarrow$  Covariance  $\text{Cov}(X, Y) = \mathbb{E}((X - \mathbb{E}(X))(Y - \mathbb{E}(Y)))$

$\Downarrow$

(Almost) positive bilinear form

(1)  $\text{Cov}(aX_1 + bX_2, Y)$

(2)  $\text{Cov}(X, aY_1 + bY_2)$

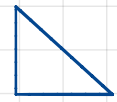
(3)  $\text{Cov}(X, X) = \text{Var}(X)$

Orthogonal  $\rightsquigarrow$   $\text{Cov}(X, Y) = 0$  uncorrelated

Length<sup>2</sup>  $\rightsquigarrow$   $\text{Var}(X)$

"Angles"  $\rightsquigarrow$

Ex.  $(X, Y)$  uniformly distributed on the triangle



$$\{(x, y) : x, y \geq 0, x + y \leq 1\}$$

# Moment Generating Function Revisited

8.3

Suppose  $X, Y$  are independent, and  $M_X, M_Y < \infty$  on an interval containing 0. Then

$$M_{X+Y}(t)$$

E.g.  $X \sim \text{Poisson}(\lambda), Y \sim \text{Poisson}(\mu)$  independent  
 $M_X(t) = e^{\lambda(e^t-1)} \quad M_Y(t) = e^{\mu(e^t-1)}$

Eg.  $X \sim \mathcal{N}(\mu_1, \sigma_1^2)$     $Y \sim \mathcal{N}(\mu_2, \sigma_2^2)$    independent

## Reminder: Chebyshev's Inequality

For any random variable  $X$  with finite

$$\mathbb{E}(X) = \mu \quad \text{Var}(X) = \sigma^2$$

$$\mathbb{P}(|X - \mu| \geq k\sigma) \leq \frac{1}{k^2}$$

We proved this using the fact that  $\mathbb{E}$  is monotone:

$$X \leq Y \Rightarrow \mathbb{E}(X) \leq \mathbb{E}(Y)$$

E.g. Ramen Menya Ultra has, on average, 1000 customers/day, with a standard deviation of 15. Estimate the probability that today they will have between 956 and 1044 customers.

9.1-

9.2

## (Weak) Law of Large Numbers

Let  $X_1, X_2, X_3, \dots, X_n, \dots$  be an infinite sequence of iid. random variables, each with  $\mathbb{E}(X_j) = \mu$  and  $\text{Var}(X_j) = \sigma^2$  finite.

Let  $S_n = X_1 + X_2 + \dots + X_n$ . Then for any fixed  $\varepsilon > 0$ ,



E.g. The Large Hadron Collider was built to detect and measure the mass of the Higgs Boson. Call the mass  $M$ .

For theoretical reasons, it is known that  $M \leq 1.78 \times 10^{-23} \text{ g}$ .

How many trials do the LHC physicists need to do to estimate the correct mass (via sample mean) within  $10^{-24} \text{ g}$ , with probability  $\geq 95\%$ ?

## Strong Law of Large Numbers

Let  $X_1, X_2, X_3, \dots, X_n, \dots$  be an infinite sequence of iid. random variables each with  $\mathbb{E}(X_j) = \mu$ .

Let  $\bar{X}_n = \frac{X_1 + X_2 + \dots + X_n}{n}$ . Then