

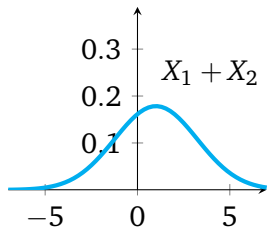
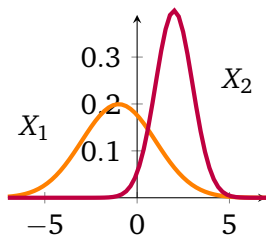
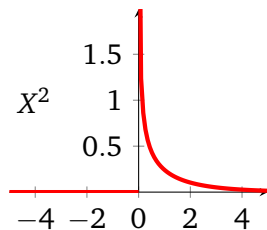
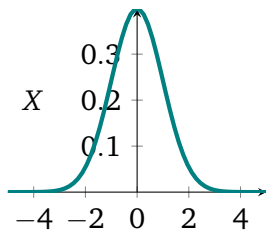
# ILLC Project Course in Statistical Learning Theory

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# Random Variables



# Expected Values

## Definition

The **mean** or **expected value** of a discrete variable  $X$  is

$$E[X] = p(x_1)x_1 + p(x_2)x_2 + p(x_3)x_3 + \dots$$

The mean of a continuous variable is the corresponding integral.

## Theorem

1.  $E[X + Y] = E[X] + E[Y]$ ;
2.  $E[c \cdot X] = c \cdot E[X]$  when  $c$  is a constant;
3.  $E[X \cdot Y] = E[X] \cdot E[Y]$  when  $X$  and  $Y$  are independent.

# Variance

## Problem

Which of the following is more probable?

1. 600 tails in 1,000 coin flips;
2. 600,000 tails in 1,000,000 coin flips.

## Problem

A dust particle starts at  $X_1 = 0$  and walks one step left or one step right in each time period. Approximately how far away from the origin is the particle going to be after  $t = 100$  time periods?

# Variance

$Z$	$z$			$E[Z]$
$X$	-4	0	12	
$X - 2$	-6	-2	10	
$ X - 2 $	6	2	10	
$(X - 2)^2$	36	4	100	
$\Pr(Z = z)$	1/4	1/2	1/4	

## Variance

$Z$	$z$			$E[Z]$
$X$	-4	0	12	2
$X - 2$	-6	-2	10	0
$ X - 2 $	6	2	10	5
$(X - 2)^2$	36	4	100	36
$\Pr(Z = z)$	1/4	1/2	1/4	

# Variance

## Definition

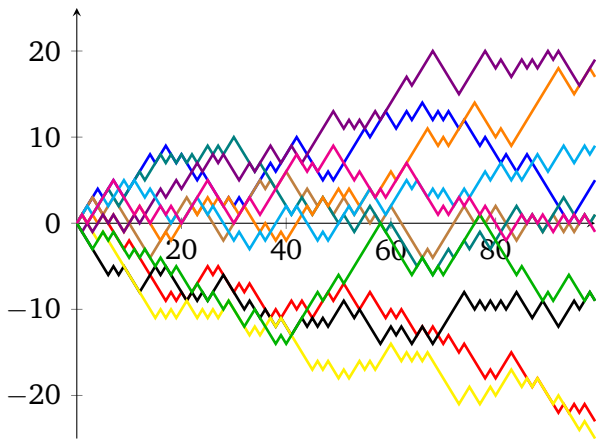
The **variance** of a random variable  $X$  is

$$\mathbf{Var}[X] = E[(X - E[X])^2].$$

## Theorem

1.  $\mathbf{Var}[X] = E[X^2] - E[X]^2$ ;
2.  $\mathbf{Var}[c \cdot X] = c^2 \cdot \mathbf{Var}[X]$  when  $c$  is a constant;
3.  $\mathbf{Var}[X + Y] = \mathbf{Var}[X] + \mathbf{Var}[Y]$  when  $X$  and  $Y$  are independent.

## Variance



$$\text{Var}[X_1 + \dots + X_t] = \text{Var}[X_1] + \dots + \text{Var}[X_t]$$

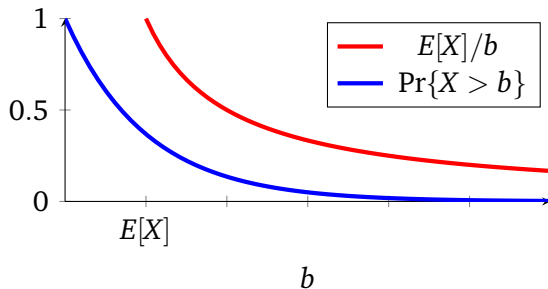


# The Markov Bound

## Theorem

Suppose that  $\Pr\{X \geq 0\} = 1$ ; then

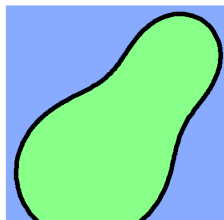
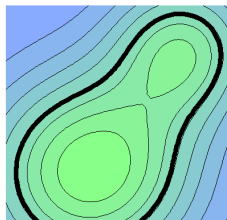
$$\Pr(X > b) \leq \frac{E[X]}{b}$$



# The Markov Bound

$$B = \begin{cases} b & (X > b) \\ 0 & (X \leq b) \end{cases}$$

$X$	0	1	2	3	4
$B$	0	0	0	2	2
$\text{Pr}$	.25	.10	.25	.30	.10



$$E[X] \geq E[B] = b \text{Pr}\{X > b\}.$$

# The Weak Law of Large Numbers

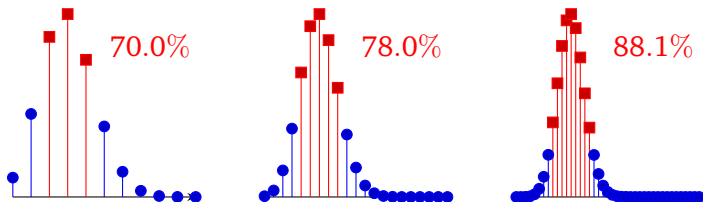
## Theorem (Tchebichef, 1867)

Let

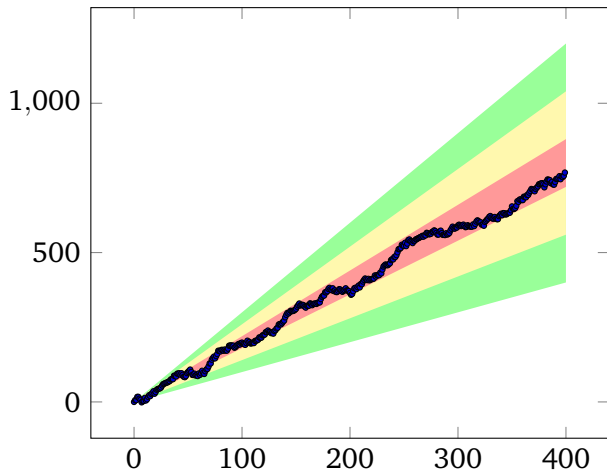
$$S_t = X_1 + X_2 + \cdots + X_t$$

be a sum of independent and identically distributed variables with a shared mean  $E[X]$  and variance  $\text{Var}[X]$ . Then

$$\Pr \left\{ \left| \frac{S_t}{t} - E[X] \right| > \varepsilon \right\} \leq \frac{\text{Var}[X]}{t\varepsilon^2}.$$



# The Weak Law of Large Numbers



$$(E[X] - \varepsilon)t \leq S_t \leq (E[X] + \varepsilon)t$$

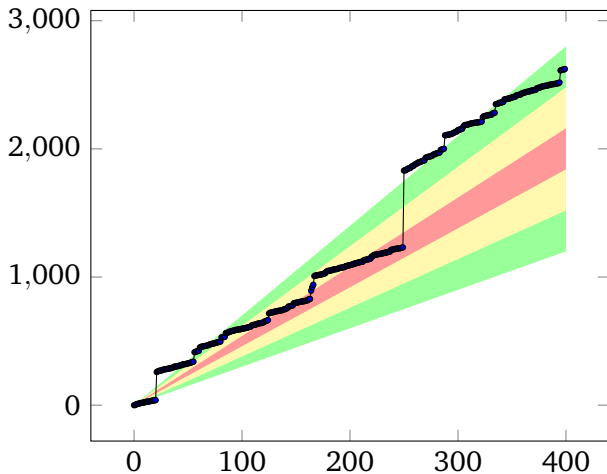
# The Weak Law of Large Numbers

## Problem

How many times do you have to throw a loaded die in order for the average result to deviate from the expected result by less than  $\varepsilon = 1.50$  with a probability of less than  $\alpha = .05$ ?

$$\Pr \left\{ \left| \frac{S_t}{t} - E[X] \right| > \varepsilon \right\} \leq \frac{\text{Var}[X]}{t\varepsilon^2}.$$

# The Weak Law of Large Numbers



$$(E[X] - \epsilon)t \leq S_t \leq (E[X] + \epsilon)t$$