#### Economics 103 – Statistics for Economists

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Lecture # 11

#### Continuous RVs – What Changes?

- Probability Density Functions replace Probability Mass Functions (aka Probability Distributions)
- 2. Integrals Replace Sums

Everything Else is Essentially Unchanged!

# What is the probability of "Yellow?"

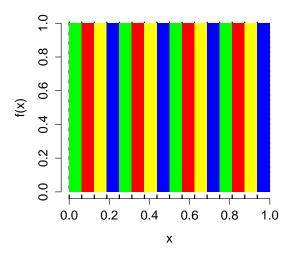




# What is the probability that the spinner lands in any particular place?



# From Twister to Density - Probability as Area



#### Continuous Random Variables

For continuous RVs, probability is a matter of finding the area of *intervals*. Individual *points* have *zero* probability.

# Probability Density Function (PDF)

For a continuous random variable X,

$$P(a \le X \le b) = \int_a^b f(x) \ dx$$

where f(x) is the probability density function for X.

#### Extremely Important

For any realization x,  $P(X = x) = 0 \neq f(x)$ !

#### Properties of PDFs

$$1. \int_{-\infty}^{\infty} f(x) \ dx = 1$$

- 2.  $f(x) \ge 0$  for all x
- 3. f(x) is not a probability and can be greater than one!

4. 
$$P(X \le x_0) = F(x_0) = \int_{-\infty}^{x_0} f(x) dx$$

# Simplest Possible Continuous RV: Uniform(0,1)

You'll look at a generalization, Uniform(a, b) for homework.

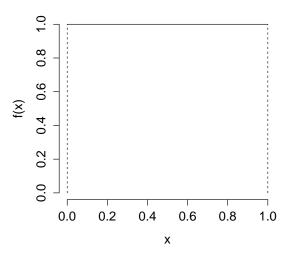
#### $X \sim \text{Uniform}(0,1)$

A Uniform(0,1) RV is equally likely to take on *any value* in the range [0,1] and never takes on a value outside this range.

#### Uniform PDF

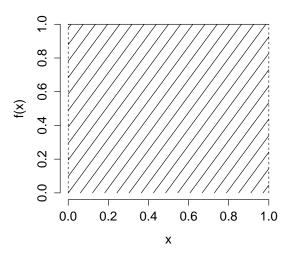
f(x) = 1 for  $0 \le x \le 1$ , zero elsewhere.

# $\mathsf{Uniform}(0,1)\;\mathsf{PDF}$

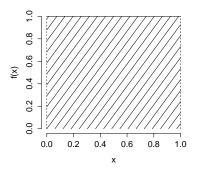








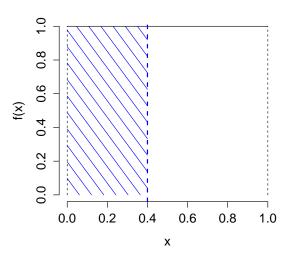
#### What is the area of the shaded region?



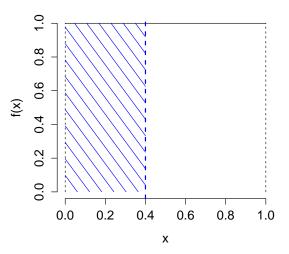
$$\int_{-\infty}^{\infty} f(x) \ dx = \int_{0}^{1} 1 \ dx = x|_{0}^{1} = 1 - 0 = 1$$







# $F(0.4) = P(X \le 0.4) = 0.4$



#### Relationship between PDF and CDF

Integrate pdf  $\rightarrow$  CDF

$$F(x_0) = P(X \le x_0) = \int_{-\infty}^{x_0} f(x) dx$$

Differentiate CDF  $\rightarrow$  pdf

$$f(x) = \frac{d}{dx}F(x)$$

This is just the First Fundamental Theorem of Calculus.

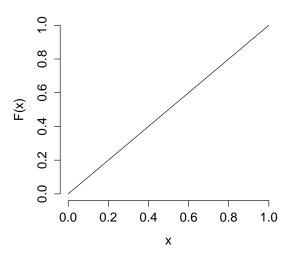
#### Example: Uniform(0,1) RV

Integrate the pdf, f(x) = 1, to get the CDF

$$F(x_0) = \int_{-\infty}^{x_0} f(x) \ dx = \int_0^{x_0} 1 \ dx = x|_0^{x_0} = x_0 - 0 = x_0$$

$$F(x_0) = \begin{cases} 0, x_0 < 0 \\ x_0, 0 \le x_0 \le 1 \\ 1, x_0 > 1 \end{cases}$$

# Uniform(0,1) CDF



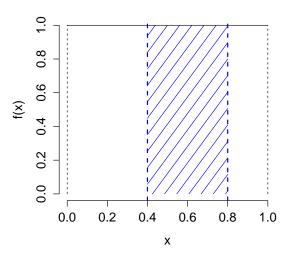
# Example: Uniform(0,1) RV

Differentiate the CDF,  $F(x_0) = x_0$ , to get the pdf

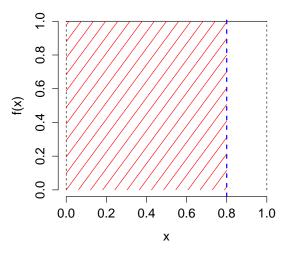
$$\frac{d}{dx}F(x)=1=f(x)$$



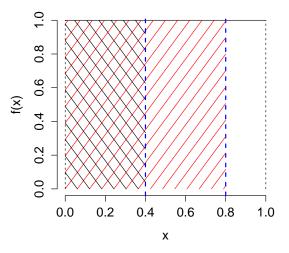




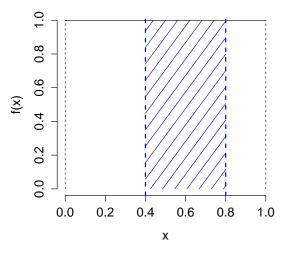
# $F(0.8) = P(X \le 0.8)$



# F(0.8) - F(0.4) = ?



$$F(0.8) - F(0.4) = P(0.4 \le X \le 0.8) = 0.4$$



#### Key Idea: Probability of Interval for Continuous RV

$$P(a \le X \le b) = \int_a^b f(x) \ dx = F(b) - F(a)$$

This is just the Second Fundamental Theorem of Calculus.

#### Expected Value for Continuous RVs

$$\int_{-\infty}^{\infty} x f(x) \ dx$$

Remember: Integrals Replace Sums!

# Example: Uniform(0,1) Random Variable



$$E[X] = \int_{-\infty}^{\infty} x f(x) \ dx =$$

# Example: Uniform(0,1) Random Variable



$$E[X] = \int_{-\infty}^{\infty} xf(x) dx = \int_{0}^{1} x \cdot 1 dx$$
$$= \frac{x^{2}}{2} \Big|_{0}^{1} = 1/2 - 0 = 1/2$$

#### Expected Value of a Function of a Continuous RV

$$E[g(X)] = \int_{-\infty}^{\infty} g(x)f(x) dx$$

# Example: Uniform(0,1) Random Variable



$$E[X^2] = \int_{-\infty}^{\infty} x^2 f(x) \ dx$$

# Example: Uniform(0,1) Random Variable



$$E[X^{2}] = \int_{-\infty}^{\infty} x^{2} f(x) dx = \int_{0}^{1} x^{2} \cdot 1 dx$$
$$= \frac{x^{3}}{3} \Big|_{0}^{1} = 1/3$$

#### What about all those rules for expected value?

- ► The only difference between expectation for continuous versus discrete is how we do the *calculation*.
- Sum for discrete; integral for continuous.
- All properties of expected value continue to hold!
- Includes linearity, shortcut for variance, etc.

#### Variance of Continuous RV

$$Var(X) = \int_{-\infty}^{\infty} (x - \mu)^2 f(x) \ dx$$

where

$$\mu = E[X] = \int_{-\infty}^{\infty} x f(x) \ dx$$

Shortcut formula still holds for continuous RVs!

$$Var(X) = E[X^2] - (E[X])^2$$

# Example: Uniform(0,1) RV

$$Var(X) = E[(X - E[X])^2] = E[X^2] - (E[X])^2$$
  
= 1/3 - (1/2)^2  
= 1/12  
 $\approx 0.083$ 

#### Much More Complicated Without the Shortcut Formula!

$$Var(X) = E\left[(X - E[X])^2\right] = \int_{-\infty}^{\infty} (x - \mu)^2 f(x) dx$$

$$= \int_0^1 (x - 1/2)^2 \cdot 1 dx = \int_0^1 (x^2 - x + 1/4) dx$$

$$= \left(\frac{x^3}{3} - \frac{x^2}{2} + \frac{x}{4}\right)\Big|_0^1 = 1/3 - 1/2 + 1/4$$

$$= 4/12 - 6/12 + 3/12 = 1/12$$

# We're Won't Say More About These, But Just So You're Aware of Them...

#### Joint Density

$$P(a \le X \le b \cap c \le Y \le d) = \int_{c}^{d} \int_{a}^{b} f(x, y) \, dxdy$$

#### Marginal Densities

$$f_X(x) = \int_{-\infty}^{\infty} f(x, y) \ dy, \qquad f_Y(y) = \int_{-\infty}^{\infty} f(x, y) \ dx$$

Independence in Terms of Joint and Marginal Densities

$$f_{XY}(x, y) = f_X(x)f_Y(y)$$

Conditional Density

$$f_{Y|X} = f_{XY}(x,y)/f_X(x)$$

#### So where does that leave us?

#### What We've Accomplished

We've covered all the basic properties of RVs on this Handout.

#### Where are we headed next?

Next up is the most important RV of all: the normal RV. After that it's time to do some statistics!

#### How should you be studying?

If you *master* the material on RVs (both continuous and discrete) and in particular the normal RV the rest of the semester will seem easy. If you don't, you're in for a rough time. . .