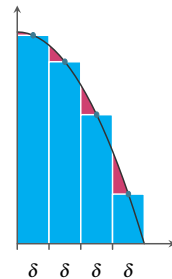


## Survey

Fill it out!!  
<https://forms.gle/XL79oruU8BHRQcaeA>

## Calculus



Riemann Sum/Integral:  $\int_a^b f(x) dx = \lim_{\delta \rightarrow 0} \sum_i \delta f(a_i)$

Derivative (Rate of change):

$$F'(x) = \lim_{h \rightarrow 0} \frac{F(x+h) - F(x)}{h}$$

Fundamental Theorem:  $F(b) - F(a) = \int_a^b F'(x) dx$ .

## CS70: Continuous Probability.

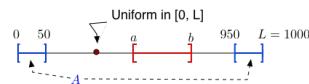
### Continuous Probability 1

1. Examples
2. Events
3. Continuous Random Variables

## Uniformly at Random in $[0, 1]$ .

Choose a real number  $X$ , uniformly at random in  $[0, 1]$ .

What is the probability that  $X$  is exactly equal to  $1/3$ ? Well, ..., 0.



What is the probability that  $X$  is exactly equal to 0.6? Again, 0.

In fact, for any  $x \in [0, 1]$ , one has  $Pr[X = x] = 0$ .

How should we then describe 'choosing uniformly at random in  $[0, 1]$ '?

Here is the way to do it:

$$Pr[X \in [a, b]] = b - a, \forall 0 \leq a \leq b \leq 1.$$

Makes sense:  $b - a$  is the fraction of  $[0, 1]$  that  $[a, b]$  covers.

## Uniformly at Random in $[0, 1]$ .

Let  $[a, b]$  denote the **event** that the point  $X$  is in the interval  $[a, b]$ .

$$Pr[[a, b]] = \frac{\text{length of } [a, b]}{\text{length of } [0, 1]} = \frac{b - a}{1} = b - a.$$

Intervals like  $[a, b] \subseteq \Omega = [0, 1]$  are **events**.

More generally, events in this space are **unions of intervals**.

Example: the event  $A$  - "within 0.2 of 0 or 1" is  $A = [0, 0.2] \cup [0.8, 1]$ .

Thus,

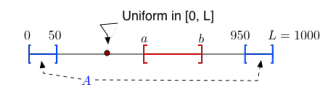
$$Pr[A] = Pr[[0, 0.2]] + Pr[[0.8, 1]] = 0.4.$$

More generally, if  $A_n$  are pairwise disjoint intervals in  $[0, 1]$ , then

$$Pr[\cup_n A_n] := \sum_n Pr[A_n].$$

Many subsets of  $[0, 1]$  are of this form. Thus, the probability of those sets is well defined. We call such sets **events**.

## Uniformly at Random in $[0, 1]$ .



Note: A **radical** change in approach.

**Finite prob. space:**  $\Omega = \{1, 2, \dots, N\}$ , with  $Pr[\omega] = p_\omega$ .

$$\implies Pr[A] = \sum_{\omega \in A} p_\omega \text{ for } A \subseteq \Omega.$$

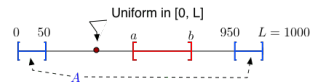
**Continuous space:** e.g.,  $\Omega = [0, 1]$ ,

**$Pr[\omega]$  is typically 0.**

Instead, start with  $Pr[A]$  for some events  $A$ .

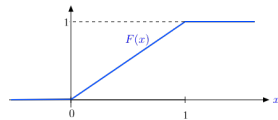
Event  $A$  = interval, or union of intervals.

### Uniformly at Random in $[0, 1]$ .



$Pr[X \leq x] = x$  for  $x \in [0, 1]$ . Also,  $Pr[X \leq x] = 0$  for  $x < 0$ .  
 $Pr[X \leq x] = 1$  for  $.2x > 1$ .

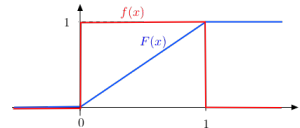
Define  $F(x) = Pr[X \leq x]$ .



Then we have  $Pr[X \in (a, b]] = Pr[X \leq b] - Pr[X \leq a] = F(b) - F(a)$ .

Thus,  $F(\cdot)$  specifies the probability of all the events!

### Uniformly at Random in $[0, 1]$ .



$$Pr[X \in (a, b]] = Pr[X \leq b] - Pr[X \leq a] = F(b) - F(a).$$

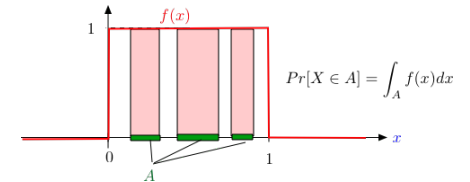
An alternative view is to define  $f(x) = \frac{d}{dx} F(x) = 1 \{x \in [0, 1]\}$ . Then

$$F(b) - F(a) = \int_a^b f(x) dx.$$

Thus, the probability of an event is the integral of  $f(x)$  over the event:

$$Pr[X \in A] = \int_A f(x) dx.$$

### Uniformly at Random in $[0, 1]$ .



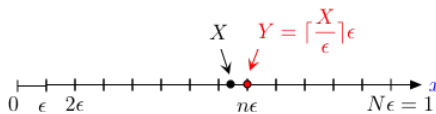
Think of  $f(x)$  as describing how one unit of probability is spread over  $[0, 1]$ : uniformly!

Then  $Pr[X \in A]$  is the probability mass over  $A$ .

Observe:

- This makes the probability automatically additive.
- We need  $f(x) \geq 0$  and  $\int_{-\infty}^{\infty} f(x) dx = 1$ .

### Uniformly at Random in $[0, 1]$ .



**Discrete Approximation:** Fix  $N \gg 1$  and let  $\varepsilon = 1/N$ .

Define  $Y = n\varepsilon$  if  $(n-1)\varepsilon < X \leq n\varepsilon$  for  $n = 1, \dots, N$ .

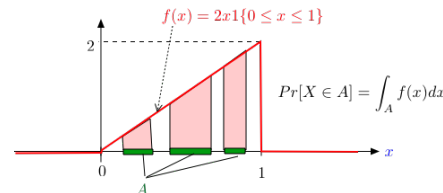
Then  $|X - Y| \leq \varepsilon$  and  $Y$  is discrete:  $Y \in \{\varepsilon, 2\varepsilon, \dots, N\varepsilon\}$ .

Also,  $Pr[Y = n\varepsilon] = \frac{1}{N}$  for  $n = 1, \dots, N$ .

Thus,  $X$  is 'almost discrete.'

Calculus view:  $Pr[Y = n\varepsilon]$  is area of rectangle in Riemann sum.

### Nonuniformly at Random in $[0, 1]$ .



This figure shows a different choice of  $f(x) \geq 0$  with  $\int_{-\infty}^{\infty} f(x) dx = 1$ .

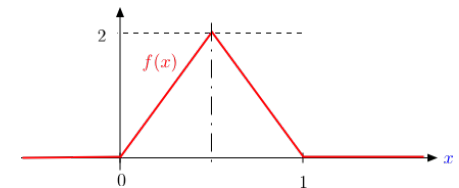
It defines another way of choosing  $X$  at random in  $[0, 1]$ .

Note that  $X$  is more likely to be closer to 1 than to 0.

One has  $Pr[X \leq x] = \int_{-\infty}^x f(u) du = x^2$  for  $x \in [0, 1]$ .

Also,  $Pr[X \in (x, x + \varepsilon)] = \int_x^{x+\varepsilon} f(u) du \approx f(x)\varepsilon$ .

### Another Nonuniform Choice at Random in $[0, 1]$ .



This figure shows yet a different choice of  $f(x) \geq 0$  with  $\int_{-\infty}^{\infty} f(x) dx = 1$ .

It defines another way of choosing  $X$  at random in  $[0, 1]$ .

Note that  $X$  is more likely to be closer to  $1/2$  than to 0 or 1.

For instance,  $Pr[X \in [0, 1/3]] = \int_0^{1/3} 4x dx = 2[x^2]_0^{1/3} = \frac{2}{9}$ .

Thus,  $Pr[X \in [0, 1/3]] = Pr[X \in [2/3, 1]] = \frac{2}{9}$  and  $Pr[X \in [1/3, 2/3]] = \frac{5}{9}$ .

## General Random Choice in $\mathfrak{R}$

Let  $F(x)$  be a nondecreasing function with  $F(-\infty) = 0$  and  $F(+\infty) = 1$ .

Define  $X$  by  $Pr[X \in (a, b]] = F(b) - F(a)$  for  $a < b$ . Also, for  $a_1 < b_1 < a_2 < b_2 < \dots < b_n$ ,

$$\begin{aligned} Pr[X \in (a_1, b_1] \cup (a_2, b_2] \cup \dots \cup (a_n, b_n]] \\ = Pr[X \in (a_1, b_1]] + \dots + Pr[X \in (a_n, b_n]] \\ = F(b_1) - F(a_1) + \dots + F(b_n) - F(a_n). \end{aligned}$$

Let  $f(x) = \frac{d}{dx} F(x)$ . Then,

$$Pr[X \in (x, x + \epsilon]] = F(x + \epsilon) - F(x) \approx f(x)\epsilon.$$

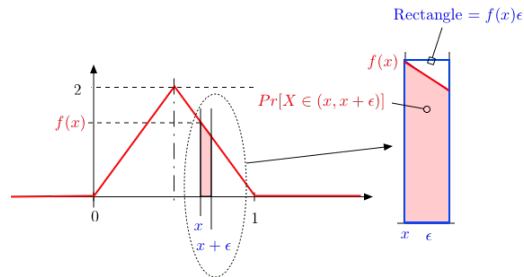
$F(x)$  is **cumulative distribution function (cdf)** of  $X$

$f(x)$  is the **probability density function (pdf)** of  $X$ .

When  $F$  and  $f$  correspond RV  $X$ :  $F_X(x)$  and  $f_X(x)$ .

## $Pr[X \in (x, x + \epsilon)]$

An illustration of  $Pr[X \in (x, x + \epsilon)] \approx f_X(x)\epsilon$ :



Thus, the pdf is the 'local probability by unit length.'

It is the 'probability density.'

## Discrete Approximation

Fix  $\epsilon \ll 1$  and let  $Y = n\epsilon$  if  $X \in (n\epsilon, (n+1)\epsilon]$ .

Thus,  $Pr[Y = n\epsilon] = F_X((n+1)\epsilon) - F_X(n\epsilon)$ .

Note that  $|X - Y| \leq \epsilon$  and  $Y$  is a discrete random variable.

Also, if  $f_X(x) = \frac{d}{dx} F_X(x)$ , then  $F_X(x + \epsilon) - F_X(x) \approx f_X(x)\epsilon$ .

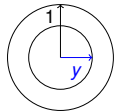
Hence,  $Pr[Y = n\epsilon] \approx f_X(n\epsilon)\epsilon$ .

Thus, we can think of  $X$  as being almost discrete with

$$Pr[X = n\epsilon] \approx f_X(n\epsilon)\epsilon.$$

## Example: CDF

Example: hitting random location on gas tank.  
Random location on circle.



Random Variable:  $Y$  distance from center.

Probability within  $y$  of center:

$$\begin{aligned} Pr[Y \leq y] &= \frac{\text{area of small circle}}{\text{area of dartboard}} \\ &= \frac{\pi y^2}{\pi} = y^2. \end{aligned}$$

Hence,

$$F_Y(y) = Pr[Y \leq y] = \begin{cases} 0 & \text{for } y < 0 \\ y^2 & \text{for } 0 \leq y \leq 1 \\ 1 & \text{for } y > 1 \end{cases}$$

## Calculation of event with dartboard..

Probability between .5 and .6 of center?

Recall CDF.

$$F_Y(y) = Pr[Y \leq y] = \begin{cases} 0 & \text{for } y < 0 \\ y^2 & \text{for } 0 \leq y \leq 1 \\ 1 & \text{for } y > 1 \end{cases}$$

$$\begin{aligned} Pr[0.5 < Y \leq 0.6] &= Pr[Y \leq 0.6] - Pr[Y \leq 0.5] \\ &= F_Y(0.6) - F_Y(0.5) \\ &= .36 - .25 \\ &= .11 \end{aligned}$$

## PDF.

Example: "Dart" board.

Recall that

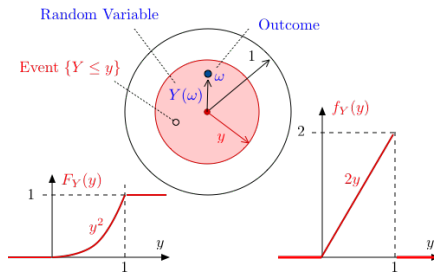
$$F_Y(y) = Pr[Y \leq y] = \begin{cases} 0 & \text{for } y < 0 \\ y^2 & \text{for } 0 \leq y \leq 1 \\ 1 & \text{for } y > 1 \end{cases}$$

$$f_Y(y) = F_Y'(y) = \begin{cases} 0 & \text{for } y < 0 \\ 2y & \text{for } 0 \leq y \leq 1 \\ 0 & \text{for } y > 1 \end{cases}$$

The cumulative distribution function (cdf) and probability distribution function (pdf) give full information.

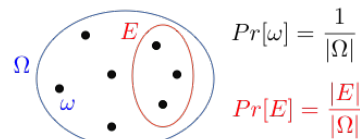
Use whichever is convenient.

## Target



## $U[a, b]$

### Uniform Probability Space



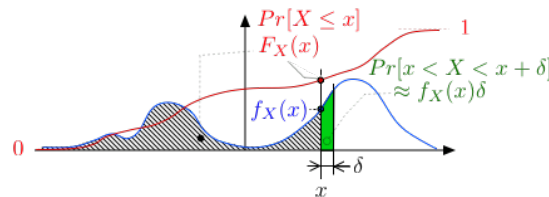
## Continuous Random Variables

Continuous random variable  $X$ , specified by

1.  $F_X(x) = \Pr[X \leq x]$  for all  $x$ .  
**Cumulative Distribution Function (cdf).**  
 $\Pr[a < X \leq b] = F_X(b) - F_X(a)$   
 1.1  $0 \leq F_X(x) \leq 1$  for all  $x \in \mathfrak{R}$ .  
 1.2  $F_X(x) \leq F_X(y)$  if  $x \leq y$ .
2. Or  $f_X(x)$ , where  $F_X(x) = \int_{-\infty}^x f_X(u) du$  or  $f_X(x) = \frac{d(F_X(x))}{dx}$ .  
**Probability Density Function (pdf).**  
 $\Pr[a < X \leq b] = \int_a^b f_X(x) dx = F_X(b) - F_X(a)$   
 2.1  $f_X(x) \geq 0$  for all  $x \in \mathfrak{R}$ .  
 2.2  $\int_{-\infty}^{\infty} f_X(x) dx = 1$ .

Recall that  $\Pr[X \in (x, x + \delta)] \approx f_X(x)\delta$ .  
 $X$  "takes" value  $n\delta$ , for  $n \in \mathbb{Z}$ , with  $\Pr[X = n\delta] = f_X(n\delta)\delta$

## A Picture



The pdf  $f_X(x)$  is a nonnegative function that integrates to 1.  
 The cdf  $F_X(x)$  is the integral of  $f_X$ .

$$\Pr[x < X < x + \delta] \approx f_X(x)\delta$$

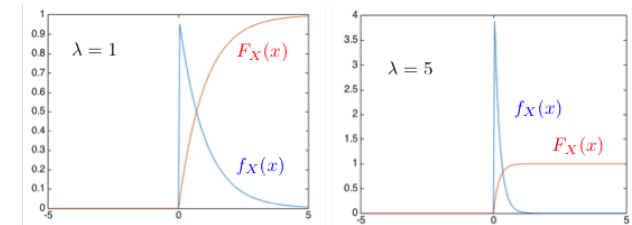
$$\Pr[X \leq x] = F_X(x) = \int_{-\infty}^x f_X(u) du$$

## $\text{Expo}(\lambda)$

The exponential distribution with parameter  $\lambda > 0$  is defined by

$$f_X(x) = \lambda e^{-\lambda x} \mathbf{1}\{x \geq 0\}$$

$$F_X(x) = \begin{cases} 0, & \text{if } x < 0 \\ 1 - e^{-\lambda x}, & \text{if } x \geq 0. \end{cases}$$



Note that  $\Pr[X > t] = e^{-\lambda t}$  for  $t > 0$ .

## Multiple Continuous Random Variables

One defines a pair  $(X, Y)$  of continuous RVs by specifying  $f_{X,Y}(x, y)$  for  $x, y \in \mathfrak{R}$  where

$$f_{X,Y}(x, y) dx dy = \Pr[X \in (x, x + dx), Y \in (y, y + dy)].$$

The function  $f_{X,Y}(x, y)$  is called the **joint pdf** of  $X$  and  $Y$ .

**Example:** Choose a point  $(X, Y)$  uniformly in the set  $A \subset \mathfrak{R}^2$ . Then

$$f_{X,Y}(x, y) = \frac{1}{|A|} \mathbf{1}\{(x, y) \in A\}$$

where  $|A|$  is the area of  $A$ .

**Interpretation.** Think of  $(X, Y)$  as being discrete on a grid with mesh size  $\varepsilon$  and  $\Pr[X = m\varepsilon, Y = n\varepsilon] = f_{X,Y}(m\varepsilon, n\varepsilon)\varepsilon^2$ .

Recall Marginal Distribution:

$$\Pr[X = x] = \sum_y \Pr[X = x, Y = y].$$

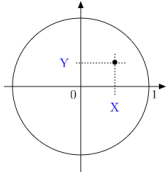
Similarly:

$$f_X(x) = \int f_{X,Y}(x, y) dy.$$

Sum "goes to" integral.

### Example of Continuous (X, Y)

Pick a point (X, Y) uniformly in the unit circle.



Thus,  $f_{X,Y}(x,y) = \frac{1}{\pi} \mathbf{1}\{x^2 + y^2 \leq 1\}$ .

Consequently,

$$Pr[X > 0, Y > 0] = \frac{1}{4}$$

$$Pr[X < 0, Y > 0] = \frac{1}{4}$$

$$Pr[X^2 + Y^2 \leq r^2] = \frac{\pi r^2}{\pi} = r^2$$

$$Pr[X > Y] = \frac{1}{2}$$

### Conditional density.

Conditional Density:  $f_{X|Y}(x,y)$ .

Conditional Probability:  $Pr[X \in A | Y \in B] = \frac{Pr[X \in A, Y \in B]}{Pr[Y \in B]}$

$$Pr[X \in [x, x+dx] | Y \in [y, y+dy]] = \frac{f_{X,Y}(x,y) dx dy}{f_Y dy}$$

$$f_{X|Y}(x,y) = \frac{f_{X,Y}(x,y)}{f_Y(y)} = \frac{f_{X,Y}(x,y)}{\int_{-\infty}^{+\infty} f_{X,Y}(x,y) dy}$$

Corollary: For independent random variables,  $f_{X|Y}(x,y) = f_X(x)$ .

### Independent Continuous Random Variables

**Definition:** Continuous RVs X and Y independent if and only if

$$Pr[X \in A, Y \in B] = Pr[X \in A] Pr[Y \in B], \forall A, B.$$

**Theorem:** Continuous RVs X and Y independent if and only if

$$f_{X,Y}(x,y) = f_X(x) f_Y(y).$$

Note:  $f_X(x)$  ( $f_Y(y)$ ) is (marginal) distribution of X (Y).

**Proof:** Intervals:  $A = [x, x+dx]$ ,  $B = [y, y+dy]$ .

$$\begin{aligned} Pr[X \in A, Y \in B] &= Pr[X \in A] \times Pr[Y \in B] \\ &\approx f_X(x) dx \times f_Y(y) dy \\ &= f_X(x) f_Y(y) dx dy. \end{aligned}$$

Thus,  $f_{X,Y}(x,y) = f_X(x) f_Y(y)$ .

□

### Mutual Independence.

**Definition:** Continuous RVs  $X_1, \dots, X_n$  are mutually independent if

$$Pr[X_1 \in A_1, \dots, X_n \in A_n] = Pr[X_1 \in A_1] \cdots Pr[X_n \in A_n], \forall A_1, \dots, A_n.$$

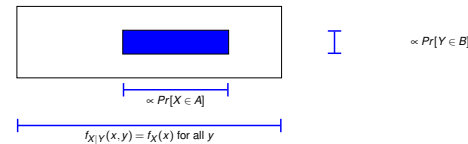
**Theorem:** Continuous RVs  $X_1, \dots, X_n$  are mutually independent if and only if

$$f_X(x_1, \dots, x_n) = f_{X_1}(x_1) \cdots f_{X_n}(x_n).$$

**Proof:** As in the discrete case.

### Independent Random Variables?

Uniform on a rectangle? Independent?



Also:  $Pr[X \in A, Y \in B] \propto \text{Area of rectangle} \propto Pr[X \in A] \times Pr[Y \in B]$ .

Independent!

Uniform on a circle? Independent?



### Summary

#### Continuous Probability 1

1. pdf:  $Pr[X \in (x, x+\delta)] = f_X(x)\delta$ .
2. CDF:  $Pr[X \leq x] = F_X(x) = \int_{-\infty}^x f_X(y) dy$ .
3.  $U[a,b]$ :  $f_X(x) = \frac{1}{b-a} \mathbf{1}\{a \leq x \leq b\}$ ;  $F_X(x) = \frac{x-a}{b-a}$  for  $a \leq x \leq b$ .
4.  $\text{Expo}(\lambda)$ :  $f_X(x) = \lambda \exp\{-\lambda x\} \mathbf{1}\{x \geq 0\}$ ;  $F_X(x) = 1 - \exp\{-\lambda x\}$  for  $x \geq 0$ .
5. Target:  $f_X(x) = 2x \mathbf{1}\{0 \leq x \leq 1\}$ ;  $F_X(x) = x^2$  for  $0 \leq x \leq 1$ .
6. Joint pdf:  $Pr[X \in (x, x+\delta), Y \in (y, y+\delta)] = f_{X,Y}(x,y)\delta^2$ .
  - 6.1 Conditional Distribution:  $f_{X|Y}(x,y) = \frac{f_{X,Y}(x,y)}{f_Y(y)}$ .
  - 6.2 Independence:  $f_{X|Y}(x,y) = f_X(x)$

## Summary

### Continuous Probability

- ▶ Continuous RVs are essentially the same as discrete RVs
- ▶ Think that  $X \approx x$  with probability  $f_X(x)\varepsilon$
- ▶ Sums become integrals, ....
- ▶ The exponential distribution is magical: memoryless.