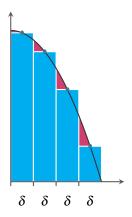


Fill it out!! tinyurl.com/cs70-survey

Calculus



Riemann Sum/Integral: $\int_a^b f(x) dx = \lim_{\delta \to 0} \sum_i \delta f(a_i)$ "Area is defined as rectangles and add up some thin ones."

Derivative (Rate of change): $F'(x) = \lim_{h \to 0} \frac{F(x+h) - F(x)}{h}$. "Rise over run of close together points."

Fundamental Theorem: $F(b) - F(a) = \int_a^b F'(x) dx$. "Area ($F(\cdot)$) under f(x) grows at x, F'(x), by f(x)" Thus F'(x) = f(x).

CS70: Continuous Probability.

Continuous Probability 1

- 1. Examples
- 2. Events
- 3. Continuous Random Variables

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:

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

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

Poll

$$F_X(x) = Pr[X \leq x]$$

$$f_X(x) = \lim_{\delta \to 0} \frac{\Pr[X \in (x, x + \delta)]}{\delta}$$

What is true? (A) $F_X(x) = \int_{-\infty}^{\infty} f_X(y) dy$ (B) $\int_{-\infty}^{\infty} f_X(x) = 1$ (C) $F_X(x) = \int_{-\infty}^{x} f(y) dy$. (D) $f(x) = F'_X(x)$. (E) $\int_{-\infty}^{\infty} F_X(x) dx = 1$. (F) $\int_{-\infty}^{\infty} xf(x) dx = \int_{-\infty}^{\infty} (1 - F(x)) dx$.

(A) False. limits wrong. (B) cuz probability distribution.(C) "sums up probability of rectangles", e.g. calculus.(D) calculus, fundamental theorem.

(F) is true since $\int_{-\infty}^{\infty} xf(x) dx = \int_{-\infty}^{\infty} F(x) dx = E[X]$.

Next lecture.

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

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

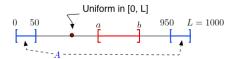
Intervals like $[a,b] \subseteq \Omega = [0,1]$ are **events.** More generally, events in this space are unions of intervals. Example: the event *A* - "within 0.2 of 0 or 1" is $A = [0,0.2] \cup [0.8,1]$. Thus,

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

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.



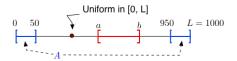
Note: A **radical** change in approach. **Finite prob. space:** $\Omega = \{1, 2, ..., N\}$, with $Pr[\omega] = p_{\omega}$. $\implies Pr[A] = \sum_{\omega \in A} p_{\omega}$ for $A \subset \Omega$.

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

 $Pr[\omega]$ is typically 0.

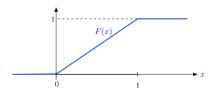
Instead, start with *Pr*[*A*] for some events *A*.

Event A = interval, or union of intervals.

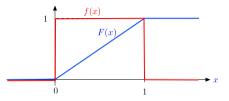


 $Pr[X \le x] = x$ for $x \in [0, 1]$. Also, $Pr[X \le x] = 0$ for x < 0. $Pr[X \le x] = 1$ for .2x > 1.

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



Then we have $Pr[X \in (a, b]] = Pr[X \le b] - Pr[X \le a] = F(b) - F(a)$. Thus, $F(\cdot)$ specifies the probability of all the events!



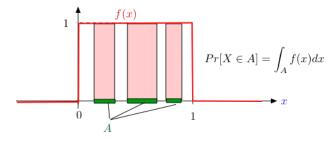
$$Pr[X \in (a,b]] = Pr[X \leq b] - Pr[X \leq a] = F(b) - F(a).$$

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:

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



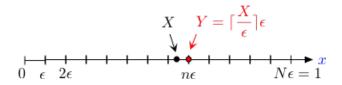
Think of f(x) as describing how one unit of probability is spread over [0,1]: uniformly!

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

Observe:

This makes the probability automatically additive.

• We need
$$f(x) \ge 0$$
 and $\int_{-\infty}^{\infty} f(x) dx = 1$.



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

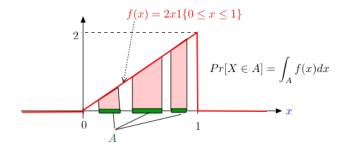
Define $Y = n\varepsilon$ if $(n-1)\varepsilon < X \le n\varepsilon$ for n = 1, ..., N.

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

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

Thus, X is 'almost discrete.'

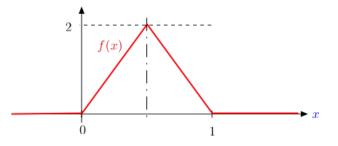
Calculus view: $Pr[Y = n\varepsilon]$ is area of rectangle in Riemann sum.



This figure shows a different choice of $f(x) \ge 0$ 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 $Pr[X \le x] = \int_{-\infty}^{x} f(u) du = x^2$ 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 $f(x) \ge 0$ 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 \Re

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 $a_1 < b_1 < a_2 < b_2 < \cdots < b_n$,

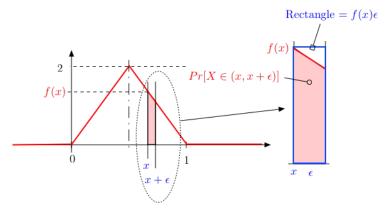
$$Pr[X \in (a_1, b_1] \cup (a_2, b_2] \cup (a_n, b_n]] \\= Pr[X \in (a_1, b_1]] + \dots + Pr[X \in (a_n, b_n]] \\= F(b_1) - F(a_1) + \dots + F(b_n) - F(a_n).$$

Let
$$f(x) = \frac{d}{dx}F(x)$$
. Then,
 $Pr[X \in (x, x + \varepsilon]] = F(x + \varepsilon) - F(x) \approx f(x)\varepsilon.$

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 $f_X(x)$.

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

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



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

Discrete Approximation

Fix
$$\varepsilon \ll 1$$
 and let $Y = n\varepsilon$ if $X \in (n\varepsilon, (n+1)\varepsilon]$.
Thus, $Pr[Y = n\varepsilon] = F_X((n+1)\varepsilon) - F_X(n\varepsilon)$.
Note that $|X - Y| \le \varepsilon$ and Y is a discrete random variable.
Also, if $f_X(x) = \frac{d}{dx}F_X(x)$, then $F_X(x + \varepsilon) - F_X(x) \approx f_X(x)\varepsilon$.

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

Thus, we can think of *X* of being almost discrete with $Pr[X = n\varepsilon] \approx f_X(n\varepsilon)\varepsilon$.

Example: CDF, pre-poll

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



What is probability of being within *y* of the center, for non-negative $y \le 1$?

(A) 1. (B) 0. (C) $\int_0^y (2\pi y) dy$ (D) y^2 .

(D) Next slide.

Example: CDF

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



Random Variable: *Y* distance from center. Probability within *y* of center:

$$Pr[Y \le y] = \frac{\text{area of small circle}}{\text{area of dartboard}}$$
$$= \frac{\pi y^2}{\pi} = y^2.$$

Hence,

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

Calculation of event with dartboard..

Probability between .5 and .6 of center? Recall CDF.

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

$$Pr[0.5 < Y \le 0.6] = Pr[Y \le 0.6] - Pr[Y \le 0.5]$$

= F_Y(0.6) - F_Y(0.5)
= .36 - .25
= .11

PDF.

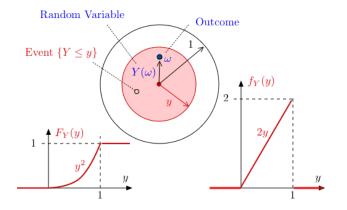
Example: "Dart" board. Recall that

$$F_{Y}(y) = \Pr[Y \le y] = \begin{cases} 0 & \text{for } y < 0\\ y^{2} & \text{for } 0 \le y \le 1\\ 1 & \text{for } y > 1 \end{cases}$$
$$f_{Y}(y) = F'_{Y}(y) = \begin{cases} 0 & \text{for } y < 0\\ 2y & \text{for } 0 \le y \le 1\\ 0 & \text{for } y > 1 \end{cases}$$

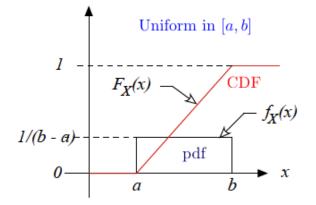
The cumulative distribution function (cdf) and probability distribution function (pdf) give full information.

Use whichever is convenient.

Target



U[*a*,*b*]



Exponential derivation:Poll.

$$Pr[X = i] = (1 - p)^{i-1}p.$$

Let $p = \lambda/n$. and $Y = X/n$.
What is true?

(A)
$$X \sim G(p)$$

(B) $Pr[X > i] = (1-p)^{i}$.
(C) $Pr[Y > i/n] = (1-\lambda/n)^{i}$.
(D) $Pr[Y > y] = (1-\lambda/n)^{ny}$.
(E) $\lim_{n\to\infty} (1-\lambda/n)^{ny} = e^{-\lambda y}$.

(A) True by definition.
(B)
$$Pr[X > i] = (1 - p)^i$$
 at least *i* coin flips fail.
(C) True, definition of *Y*
(D) True, $y = i/n$ means $i = ny$.
(E) $(1 - \lambda/n)^{ny} = ((1 - \lambda/n)^{n/\lambda})^{\lambda y}$
and $\lim_{n \to \infty} (1 - \frac{\lambda}{n})^{n/\lambda} = e$.

The limit as $n \to \infty$ of Y has $Pr[Y > y] = e^{-\lambda y}$.

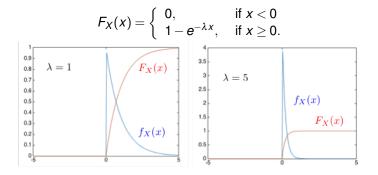
Pr[Y > y] is defined as "Survival function."

$Expo(\lambda)$

"Limit of geometric."

From last slide:
$$S(t) = Pr[X > t] = e^{-\lambda t}$$
 for $t > 0$.
Note: $f_X(x) = F'(t) = (1 - S(t))' = -S'(t)$.

The exponential distribution with parameter $\lambda > 0$ is defined by $f_X(x) = \lambda e^{-\lambda x} \mathbf{1}\{x \ge 0\}$



Continuous Random Variables

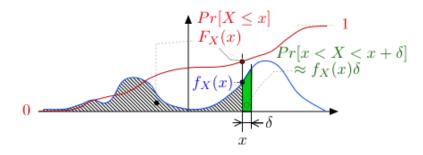
Continuous random variable X, specified by

1.
$$F_X(x) = Pr[X \le x]$$
 for all x .
Cumulative Distribution Function (cdf).
 $Pr[a < X \le b] = F_X(b) - F_X(a)$
1.1 $0 \le F_X(x) \le 1$ for all $x \in \Re$.
1.2 $F_X(x) \le F_X(y)$ if $x \le y$.

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).** $Pr[a < X \le b] = \int_a^b f_X(x) dx = F_X(b) - F_X(a)$ 2.1 $f_X(x) \ge 0$ for all $x \in \Re$. 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 $n \in Z$, 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 .

$$Pr[x < X < x + \delta] \approx f_X(x)\delta$$
$$Pr[X \le x] = F_x(x) = \int_{-\infty}^x f_X(u)du$$

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

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

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 $A \subset \Re^2$. 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 ε and $Pr[X = m\varepsilon, Y = n\varepsilon] = f_{X,Y}(m\varepsilon, n\varepsilon)\varepsilon^2$.

Recall Marginal Distribution:

$$\Pr[X=x] = \sum_{y} \Pr[X=x, Y=y].$$

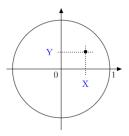
Similarly:

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

Sum "goes to" integral.

Example of Continuous (X, Y)

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



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

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

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

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

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

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].

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

$$\approx f_X(x) \ dx \times f_Y(y) \ dy$$

$$= f_X(x)f_Y(y) \ dxdy.$$

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

Mutual Independence.

Definition: Continuous RVs X_1, \ldots, X_n are mutually independent if

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

Theorem: Continuous RVs X_1, \ldots, X_n are mutually independent if and only if

$$f_{\mathbf{X}}(x_1,\ldots,x_n)=f_{X_1}(x_1)\cdots f_{X_n}(x_n).$$

Proof: As in the discrete case.

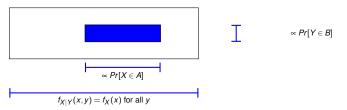
Conditional density.

Conditional Density: $f_{X|Y}(x, y)$. Conditional Probability: $Pr[X \in A | Y \in B] = \frac{Pr[X \in A, Y \in B]}{Pr[Y \in B]}$ $Pr[X \in [x, x + dx] | Y \in [y, y + dy]] = \frac{f_{X,Y}(x,y)dxdy}{f_Y dy}$ $f_{X|Y}(x, y) = \frac{f_{X,Y}(x,y)}{f_Y(y)} = \frac{f_{X,Y}(x,y)}{\int_{-\infty}^{+\infty} f_{X,Y}(x,y)dx}$

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

Independent Random Variables?

Uniform on a rectangle? Independent?



Also: $Pr[X \in A, Y \in B] \propto$ Area of rectangle $\propto Pr[X \in A] \times Pr[Y \in B]$. Independent!

Uniform on a circle? Independent?

 $f_{X|Y}(x,5)$ $f_{X|Y}(x,0)$ Not independent!

Summary

Continuous Probability 1

1. pdf:
$$Pr[X \in (x, x + \delta]] = f_X(x)\delta$$
.

- 2. CDF: $Pr[X \le x] = F_X(x) = \int_{-\infty}^x f_X(y) dy$.
- 3. U[a,b]: $f_X(x) = \frac{1}{b-a} 1\{a \le x \le b\}; F_X(x) = \frac{x-a}{b-a} \text{ for } a \le x \le b.$
- 4. $Expo(\lambda)$: $f_X(x) = \lambda \exp\{-\lambda x\} \mathbb{1}\{x \ge 0\}; F_X(x) = \mathbb{1} - \exp\{-\lambda x\} \text{ for } x \le 0.$
- 5. Target: $f_X(x) = 2x1\{0 \le x \le 1\}$; $F_X(x) = x^2$ for $0 \le x \le 1$.
- 6. Joint pdf: $Pr[X \in (x, x + \delta), Y = (y, y + \delta)) = f_{X,Y}(x, y)\delta^2$.
 - 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 similar to discrete RVs (break into intervals.)
- Think that $X \approx x$ with probability $f_X(x)\varepsilon$
- Sums become integrals,