

# Probability and Statistics

Nels Grevstad

Metropolitan State University of Denver

*ngrevsta@msudenver.edu*

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# Topics

## 1 Cumulative Distribution Functions

# Objectives

## Objectives:

- Obtain the cumulative distribution function from a probability density function.
- Use a cumulative distribution function to find probabilities.
- Obtain the probability density function from a cumulative distribution function.

# Cumulative Distribution Functions: Continuous (4.2)

- The ***cumulative distribution function*** (or ***cdf***) of a random variable  $X$ , denoted  $F(x)$ , is defined for all  $x$  as

$$F(x) = P(X \leq x).$$

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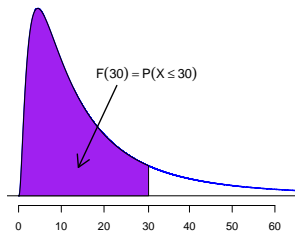
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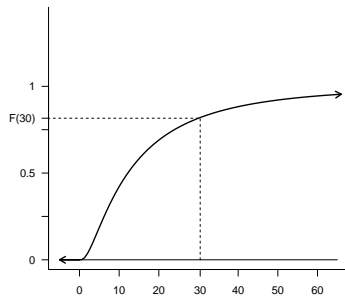
- If  $X$  is **continuous** with **pdf**  $f(x)$ , then

$$F(x) = \int_{-\infty}^x f(y) dy.$$

Pdf of X



Cdf of X





## Example

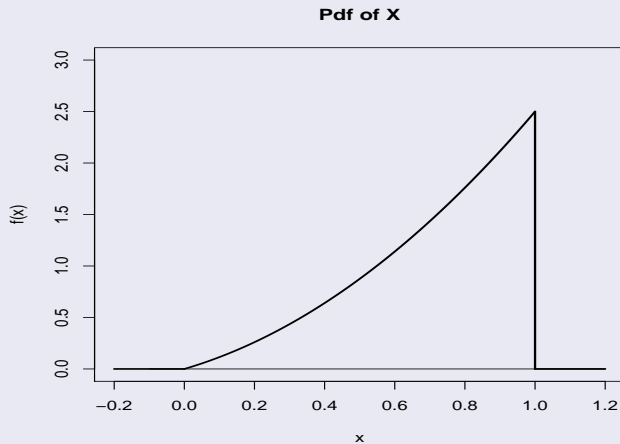
Suppose (again) that  $X$  is the gain in a certain investment, in thousands of dollars, and has **pdf**

$$f(x) = \begin{cases} \frac{1}{2}(3x^2 + 2x) & \text{for } 0 \leq x \leq 1 \\ 0 & \text{otherwise} \end{cases}$$

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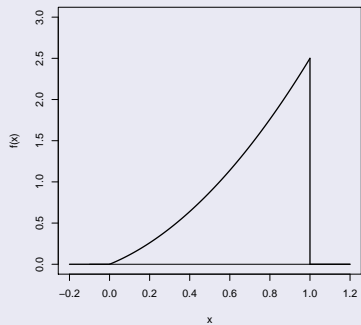
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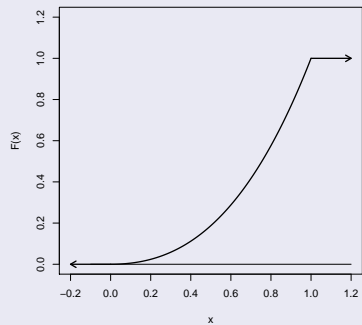
Thus the **cdf** of  $X$  is

$$F(x) = \begin{cases} 0 & \text{for } x < 0 \\ \frac{1}{2}(x^3 + x^2) & \text{for } 0 \leq x \leq 1 \\ 1 & \text{for } x > 1 \end{cases}$$

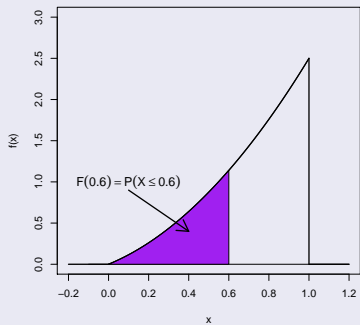
Pdf of X



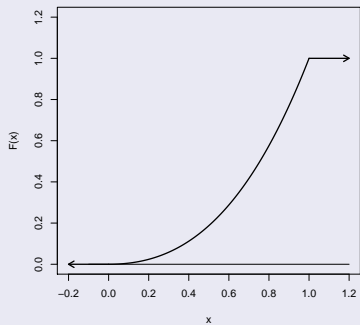
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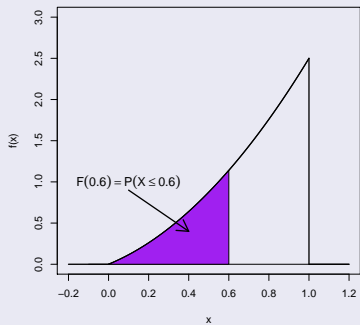
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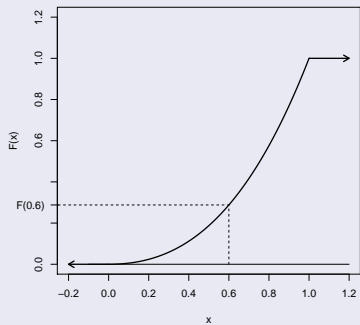
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$$\begin{aligned}P(X \leq 0.6) &= F(0.6) \\ &= \frac{1}{2}(0.6^3 + 0.6^2) \\ &= \mathbf{0.288}.\end{aligned}$$

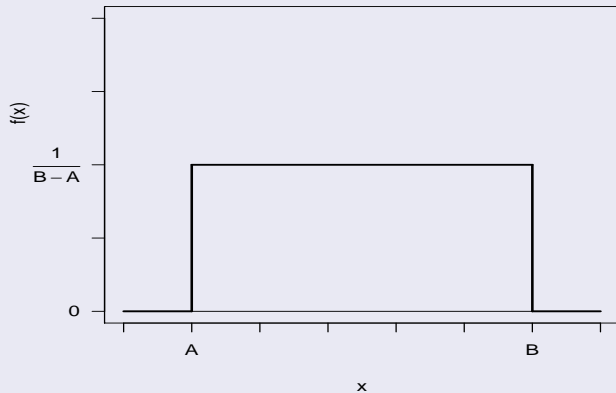


### Example

Suppose  $X$ , the thickness of a certain metal sheet, follows a **uniform distribution** over the range  $A$  to  $B$ .

The **pdf** is below.

## Uniform(A, B) Pdf



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because all of the area under the **pdf** is accumulated to the left of  $B$ .

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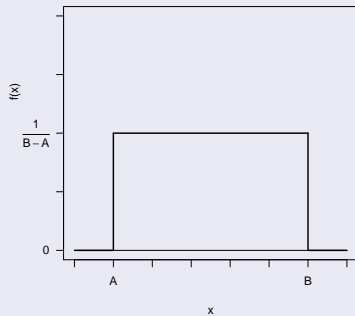
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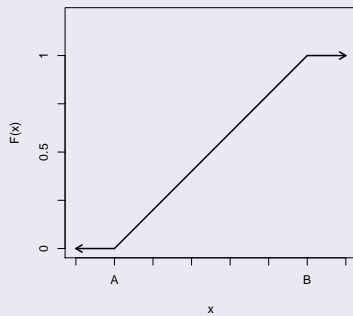
Thus the **cdf** of a **uniform** random variable is

$$F(x) = \begin{cases} 0 & \text{for } x < A \\ \frac{x-A}{B-A} & \text{for } A \leq x \leq B \\ 1 & \text{for } x > B \end{cases}$$

Uniform(A, B) Pdf



Uniform(A, B) Cdf

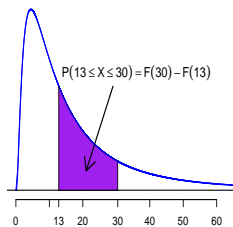


## Proposition

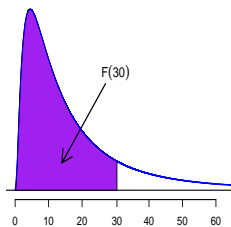
Suppose  $X$  is a random variable with **cdf**  $F(x)$ . Then for any numbers  $a$  and  $b$ , with  $a \leq b$ ,

1.  $P(X > a) = 1 - F(a)$
2.  $P(a < X \leq b) = F(b) - F(a)$

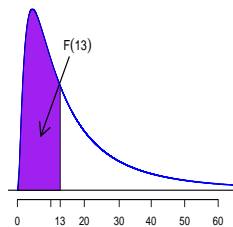
Right Skewed PDF



Right Skewed PDF



Right Skewed PDF



## Example

Suppose (again) that  $X$  is the gain in a certain investment, in thousands of dollars.

We found that the **cdf** of  $X$  is

$$F(x) = \begin{cases} 0 & \text{for } x < 0 \\ \frac{1}{2}(x^3 + x^2) & \text{for } 0 \leq x \leq 1 \\ 1 & \text{for } x > 1 \end{cases}$$

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$$\begin{aligned}P(X > 0.6) &= 1 - F(0.6) \\&= 1 - \frac{1}{2}(0.6^3 + 0.6^2) \\&= 1 - 0.288 \\&= \mathbf{0.712}.\end{aligned}$$

The probability  $P(0.4 \leq X \leq 0.8)$  that the investment gain will be **between \$400 and \$800** is

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- We obtain the **pdf**  $f(x)$  by taking the **derivative** of the **cdf**  $F(x)$ .

### Proposition

Suppose  $X$  is a continuous random variable with **cdf**  $F(x)$  and **pdf**  $f(x)$ . Then at every value  $x$  for which the derivative  $F'(x)$  exists,

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This is a consequence of the Fundamental Theorem of Calculus, which says

$$\frac{d}{dx} \int_{-\infty}^x f(y) dy = f(x).$$

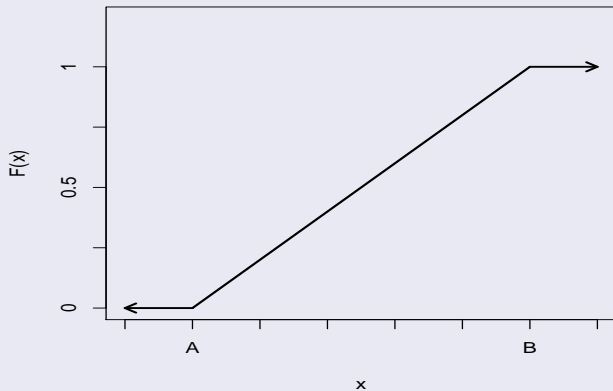
## Example

Recall that the **cdf** of a **uniform** random variable over the range  $A$  to  $B$  is

$$F(x) = \begin{cases} 0 & \text{for } x < A \\ \frac{x-A}{B-A} & \text{for } A \leq x \leq B \\ 1 & \text{for } x > B \end{cases}$$



Uniform(A, B) Cdf



The **pdf** is obtained from the **cdf** by taking its derivative:

$$f(x) = F'(x) = \begin{cases} \frac{d}{dx}(0) & \text{for } x < A \\ \frac{d}{dx}\left(\frac{x-A}{B-A}\right) & \text{for } A < x < B \\ \frac{d}{dx}(1) & \text{for } x > B \end{cases}$$

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# Cumulative Distribution Functions: Discrete (3.2)

- Recall that the ***cumulative distribution function*** (or ***cdf***) of a random variable  $X$ , denoted  $F(x)$ , is defined for all  $x$  as

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- In the **discrete** case,  $F(x)$  is a **step function**, with a step of size  $p(x)$  at each of the possible values  $x$  of the random variable  $X$ .

## Example

Suppose the **pmf** of a **discrete** random variable  $X$  is

$x$	1	1.5	2	4
$p(x)$	0.1	0.2	0.4	0.3

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$$F(x) = \begin{cases} 0 & \text{for } x < 1 \end{cases}$$



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A graph of  $F(x)$  is shown on the next slide.

Cumulative Distribution Function of  $F(x)$ 