

CS70: Jean Walrand: Lecture 34.

Continuous Probability 1

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1. Examples
2. Events
3. Continuous Random Variables

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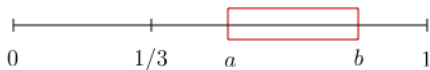
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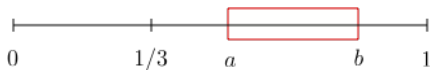
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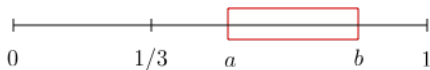


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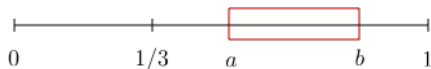


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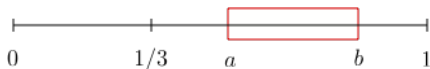
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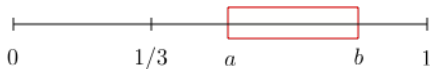
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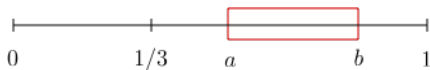
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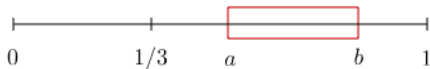
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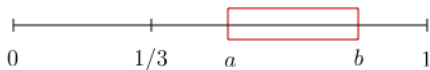
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Makes sense: $b - a$ is the fraction of $[0, 1]$ that $[a, b]$ covers.

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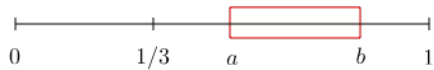
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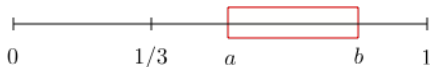
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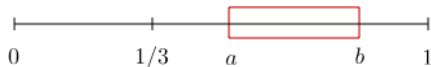


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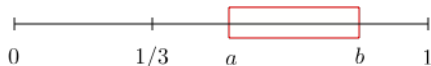
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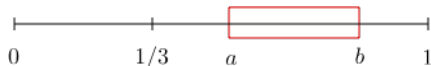
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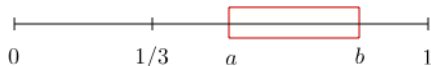
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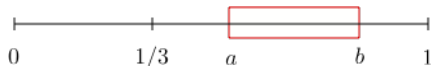
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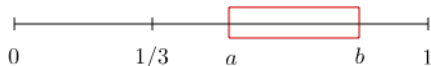
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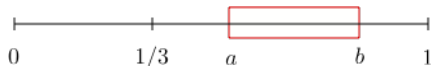
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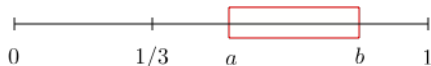
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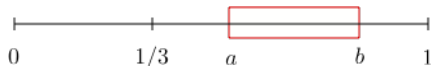
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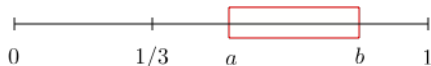
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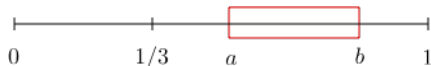
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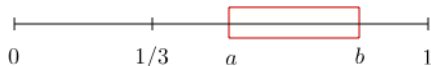
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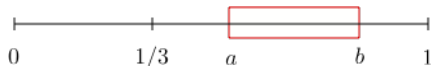
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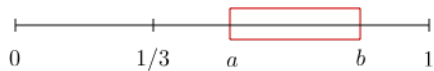
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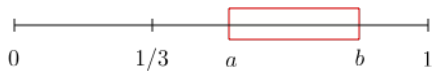
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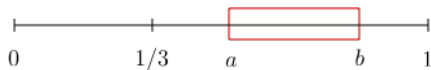


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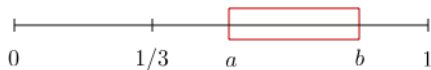
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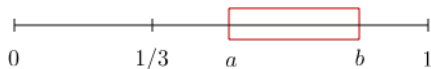
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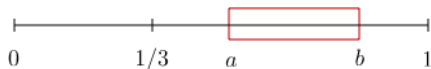
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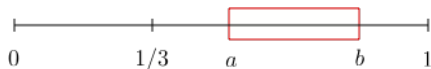
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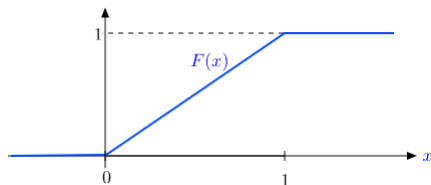


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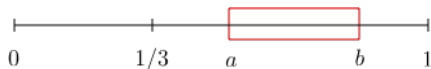
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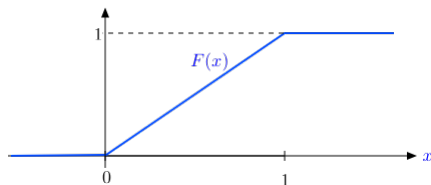
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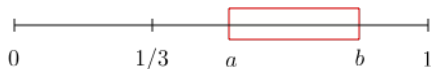


Note: $Pr[X \leq x] = x$ for $x \in [0, 1]$. Also, $Pr[X \leq x] = 0$ for $x < 0$ and $Pr[X \leq x] = 1$ for $x > 1$. Let us define $F(x) = Pr[X \leq x]$.

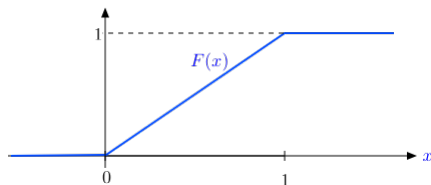


Then we have $Pr[X \in (a, b]] = Pr[X \leq b] - Pr[X \leq a]$

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

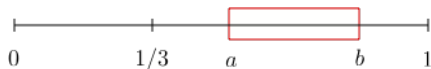


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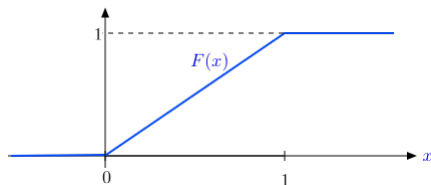


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Uniformly at Random in $[0, 1]$.



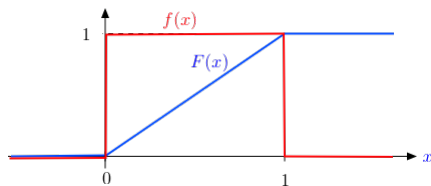
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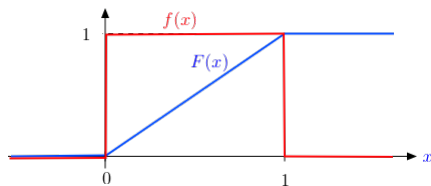
Thus, $F(\cdot)$ specifies the probability of all the events!

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



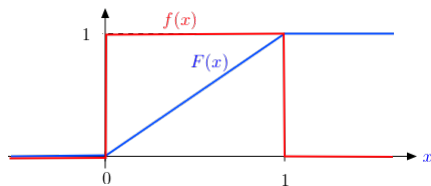
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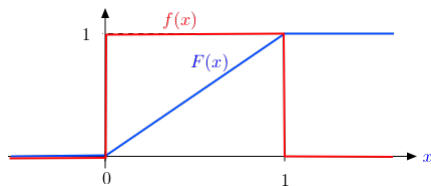
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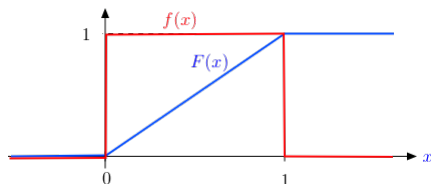
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$$Pr[X \in (a, b]] = Pr[X \leq b] - Pr[X \leq a] = F(b) - F(a).$$

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

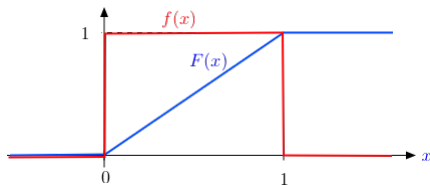


$$\Pr[X \in (a, b]] = \Pr[X \leq b] - \Pr[X \leq a] = F(b) - F(a).$$

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

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



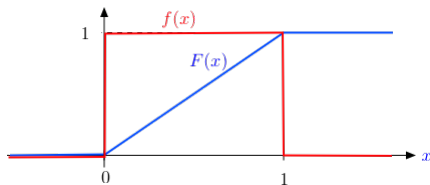
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Thus, the probability of an event is the integral of $f(x)$ over the event:

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



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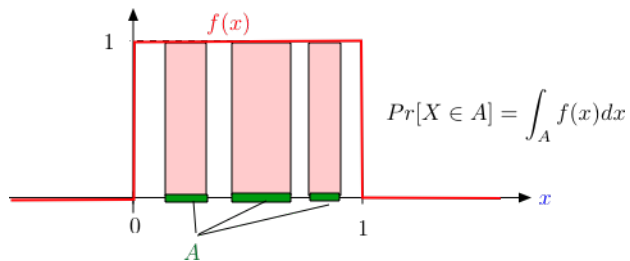
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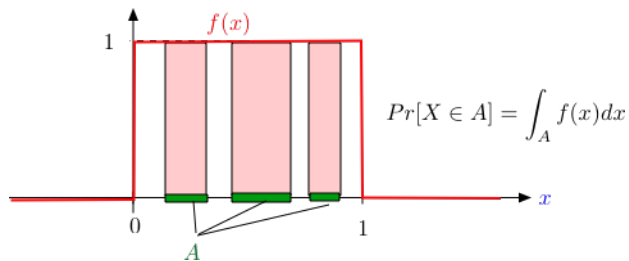
Thus, the probability of an event is the integral of $f(x)$ over the event:

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

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

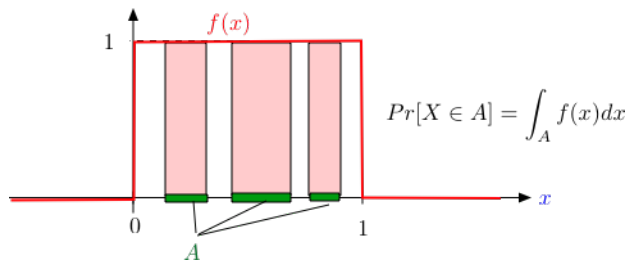


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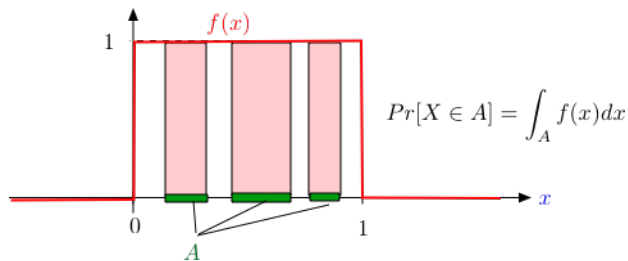
Think of $f(x)$ as describing how
one unit of probability is spread over $[0, 1]$:

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



Think of $f(x)$ as describing how
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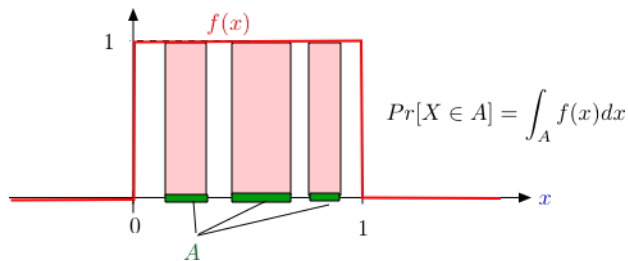
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Think of $f(x)$ as describing how
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Then $Pr[X \in A]$ is the probability mass over A .

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

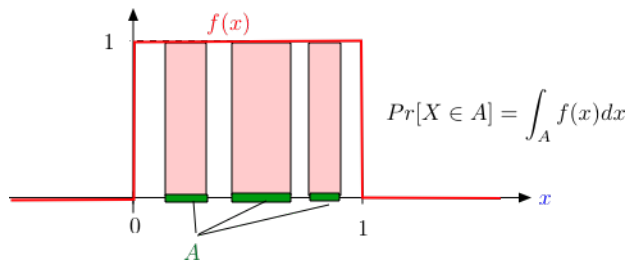


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Uniformly at Random in $[0, 1]$.



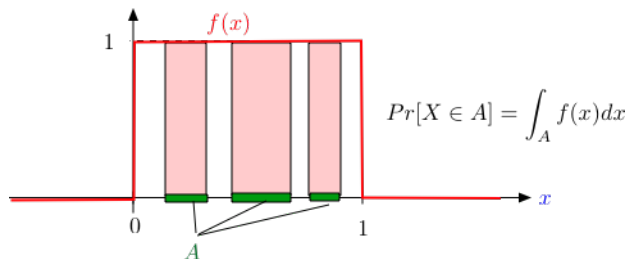
Think of $f(x)$ as describing how
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Then $Pr[X \in A]$ is the probability mass over A .

Observe:

- ▶ This makes the probability automatically additive.

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



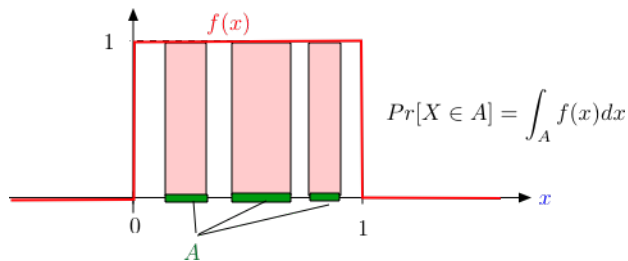
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Uniformly at Random in $[0, 1]$.



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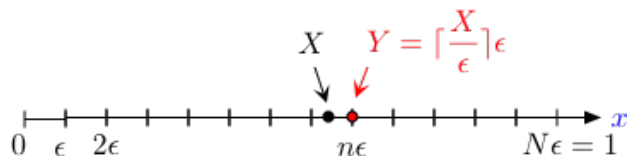
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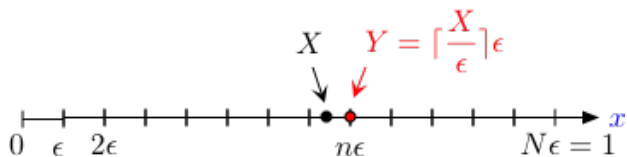
- ▶ This makes the probability automatically additive.
- ▶ We need $f(x) \geq 0$ and $\int_{-\infty}^{\infty} f(x) dx = 1$.

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

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

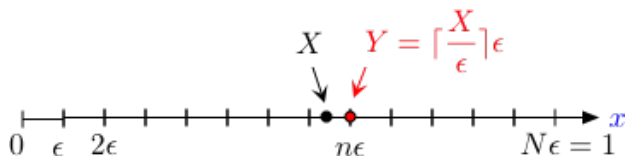


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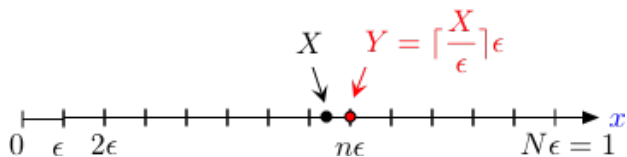
Discrete Approximation:

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



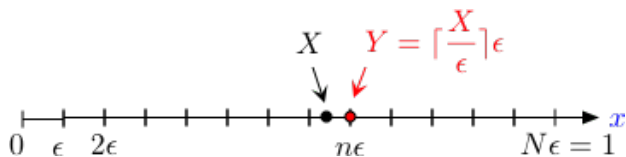
Discrete Approximation: Fix $N \gg 1$

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



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

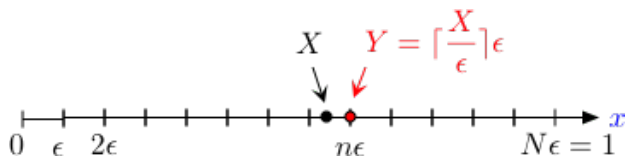
Uniformly at Random in $[0, 1]$.



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

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

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

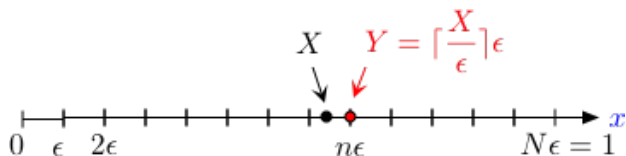


Discrete Approximation: Fix $N \gg 1$ and let $\epsilon = 1/N$.

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

Then $|X - Y| \leq \epsilon$

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

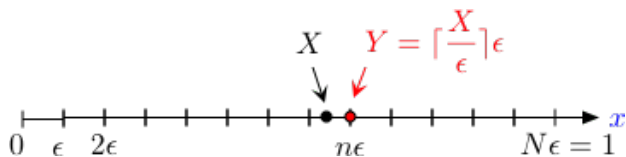


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Then $|X - Y| \leq \epsilon$ and Y is discrete:

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

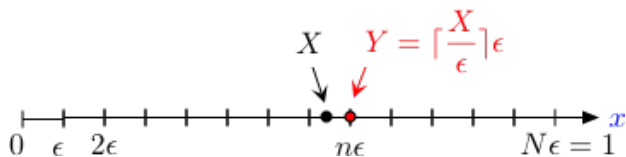


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Then $|X - Y| \leq \epsilon$ and Y is discrete: $Y \in \{\epsilon, 2\epsilon, \dots, N\epsilon\}$.

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



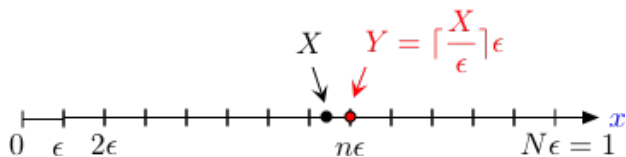
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Also, $\Pr[Y = n\epsilon] = \frac{1}{N}$ for $n = 1, \dots, N$.

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



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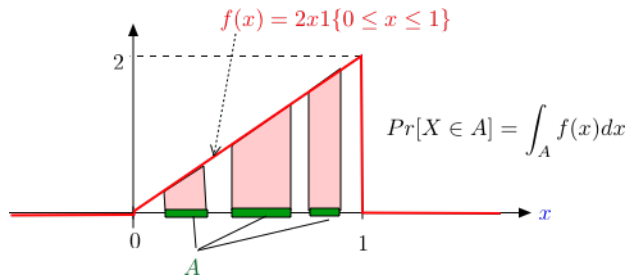
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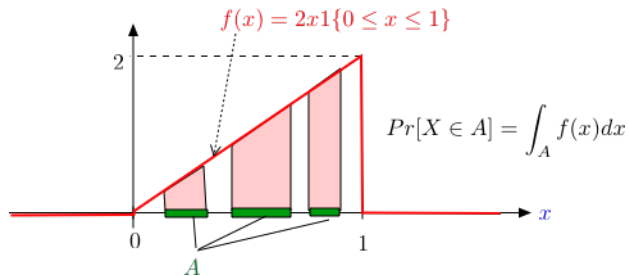
Thus, X is 'almost discrete.'

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

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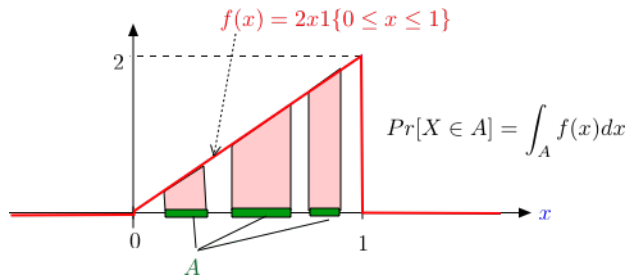


Nonuniformly at Random in $[0, 1]$.



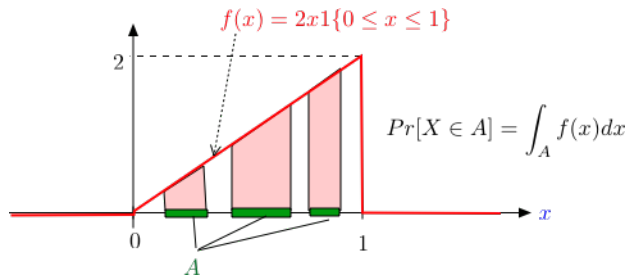
This figure shows a different choice of $f(x) \geq 0$ with $\int_{-\infty}^{\infty} f(x) dx = 1$.

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



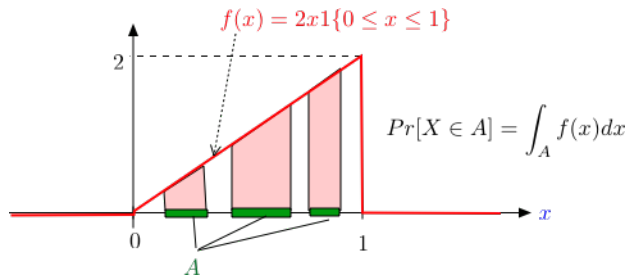
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Nonuniformly at Random in $[0, 1]$.



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

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



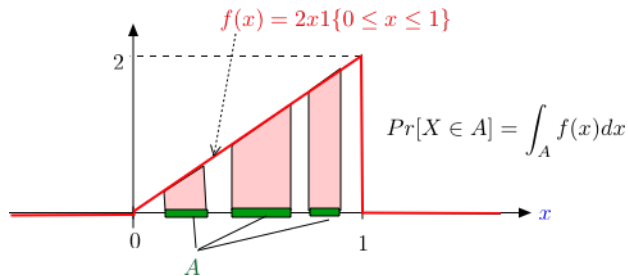
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Nonuniformly at Random in $[0, 1]$.



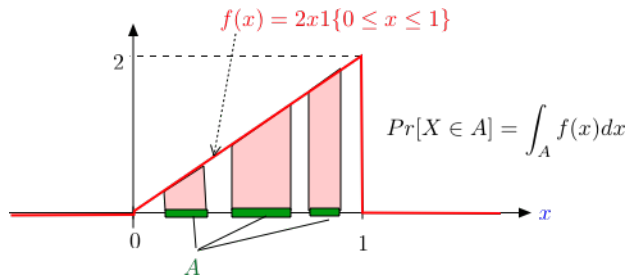
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One has $Pr[X \leq x] = \int_{-\infty}^x f(u) du = x^2$

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



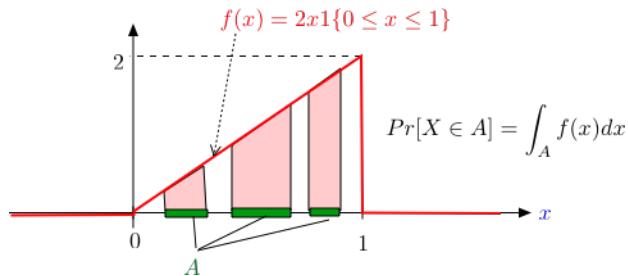
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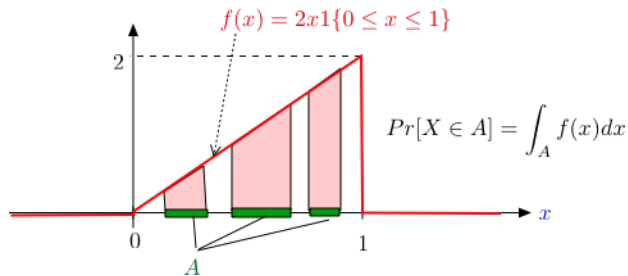
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Also, $Pr[X \in (x, x + \varepsilon)] = \int_x^{x+\varepsilon} f(u) du$

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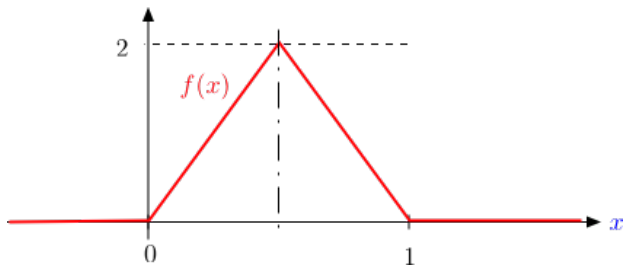
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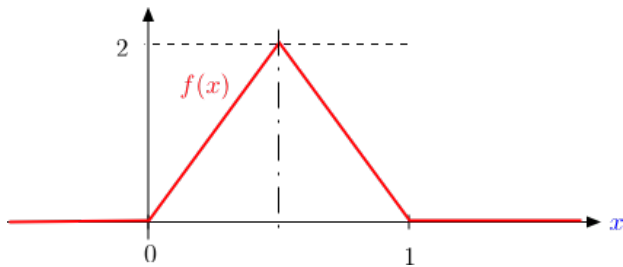
Also, $Pr[X \in (x, x + \varepsilon)] = \int_x^{x+\varepsilon} f(u) du \approx f(x)\varepsilon$.

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

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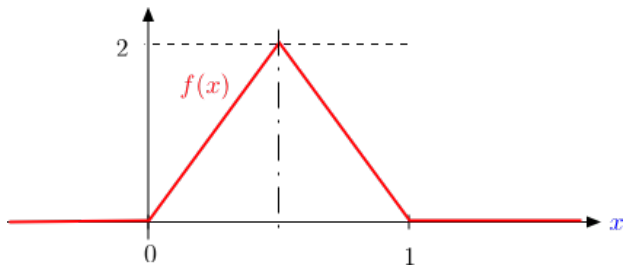


Another Nonuniform Choice at Random in $[0, 1]$.



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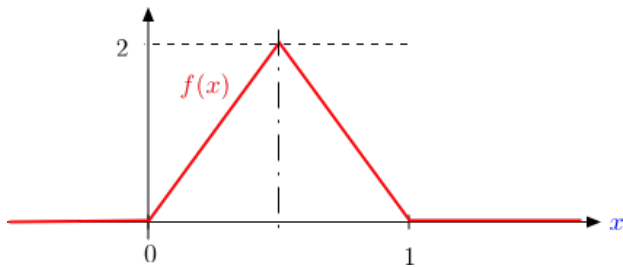
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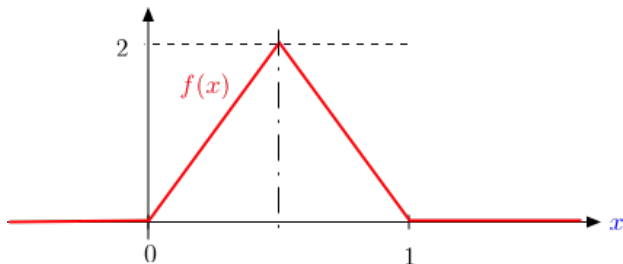


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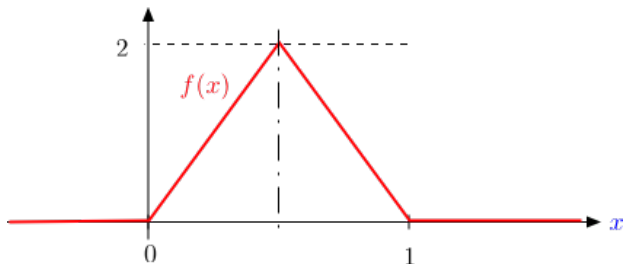
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For instance, $Pr[X \in [0, 1/3]] =$

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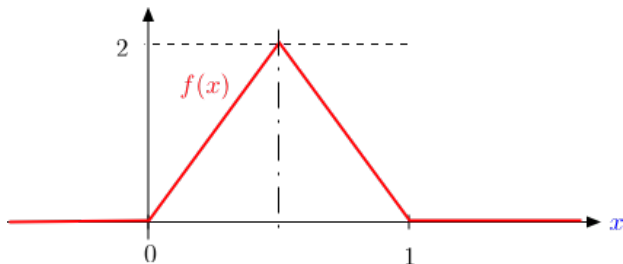
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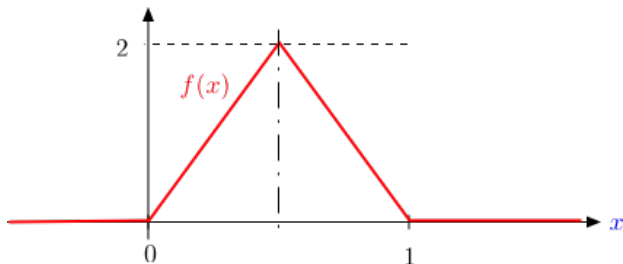
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For instance, $Pr[X \in [0, 1/3]] = \int_0^{1/3} 4x dx = 2[x^2]_0^{1/3} = \frac{2}{9}$.

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



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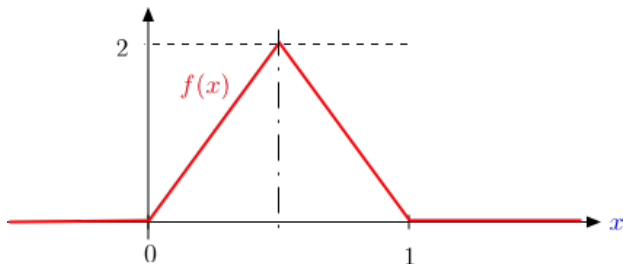
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Thus, $Pr[X \in [0, 1/3]] = Pr[X \in [2/3, 1]] = \frac{2}{9}$

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



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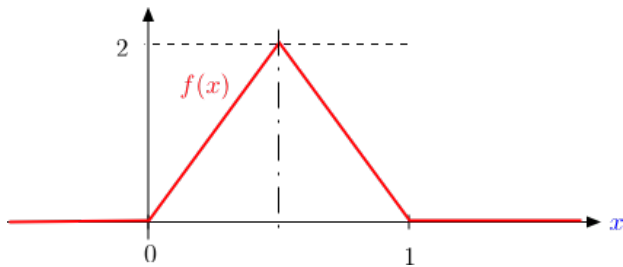
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Thus, $Pr[X \in [0, 1/3]] = Pr[X \in [2/3, 1]] = \frac{2}{9}$ and

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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 $[0, 1]$.

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To indicate that F and f correspond to the RV X , we will write them $F_X(x)$ and $f_X(x)$.

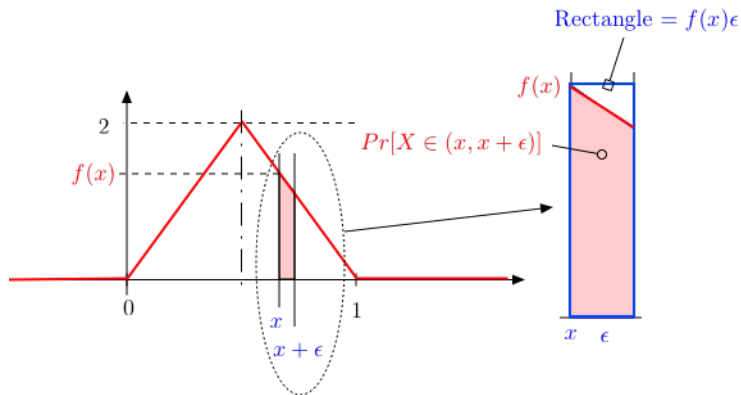
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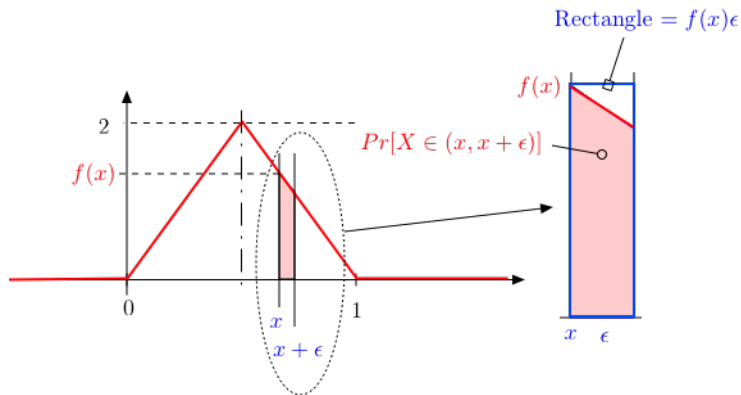
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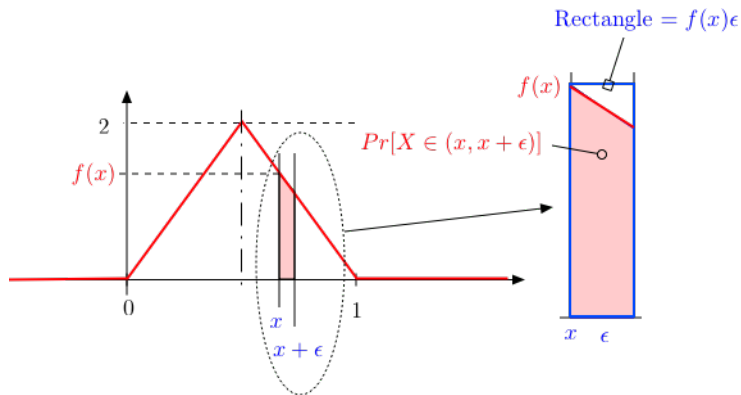
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Example: CDF

Example: hitting random location on gas tank.

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Random location on circle.

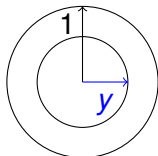
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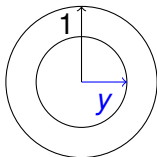
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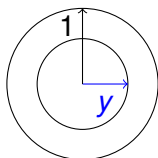
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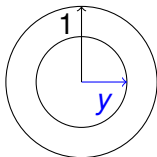
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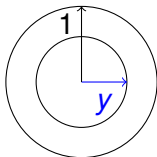
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Calculation of event with dartboard..

Probability between .5 and .6 of center?

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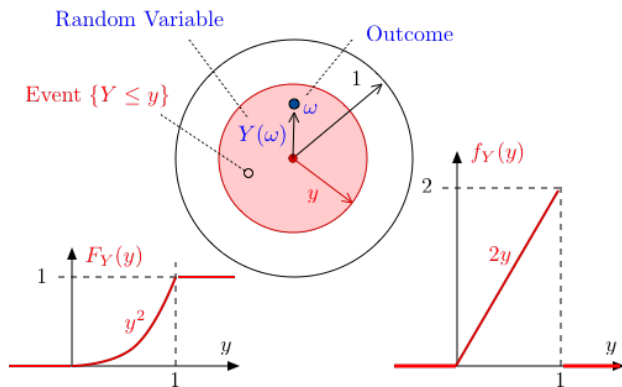
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Use whichever is convenient.

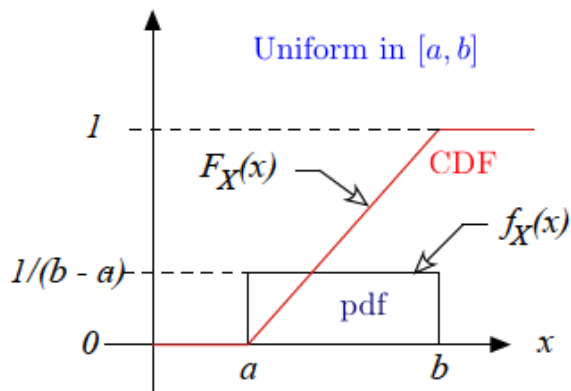
Target

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$U[a, b]$

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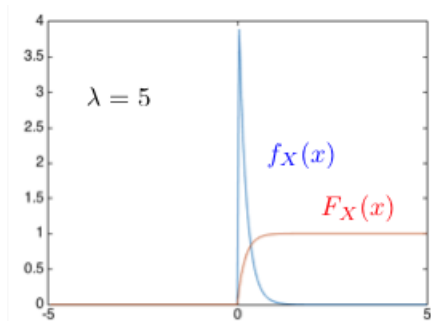
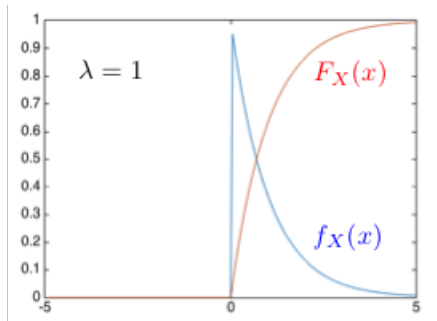
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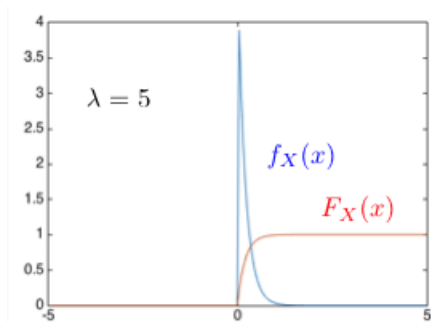
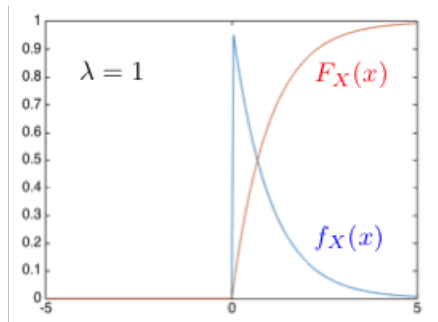


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Note that $Pr[X > t] = e^{-\lambda t}$ for $t > 0$.

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Continuous random variable X , specified by

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1.1 $0 \leq F_X(x) \leq 1$ for all $x \in \mathfrak{R}$.

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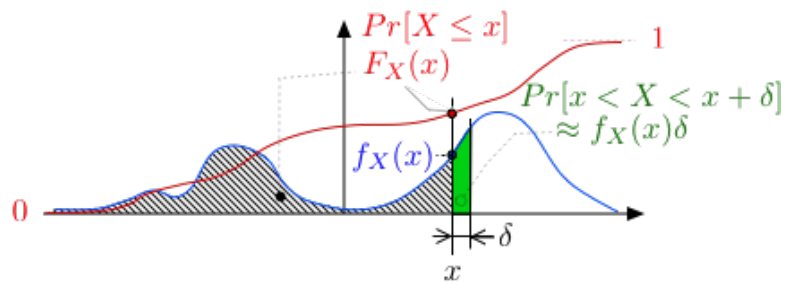
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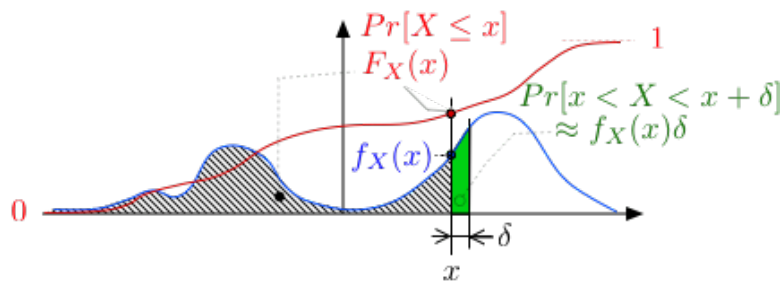
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Recall that $Pr[X \in (x, x + \delta)] \approx f_X(x)\delta$. Think of X taking discrete values $n\delta$ for $n = \dots, -2, -1, 0, 1, 2, \dots$ with $Pr[X = n\delta] = f_X(n\delta)\delta$.

A Picture

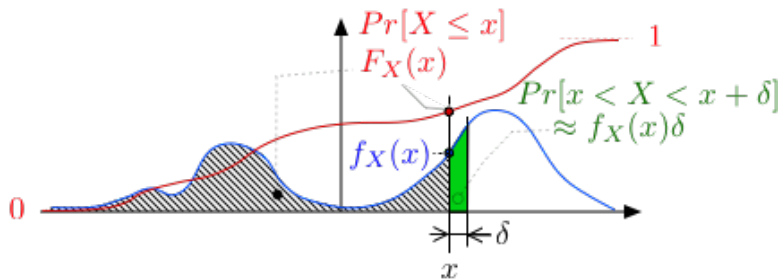


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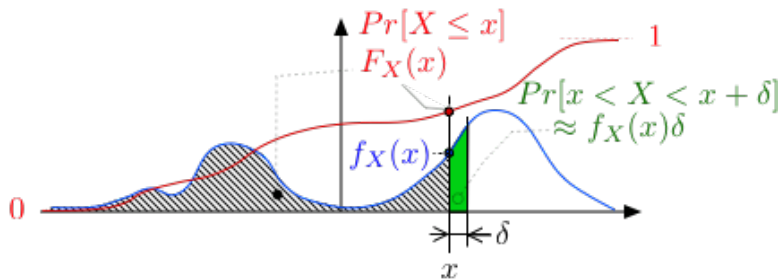
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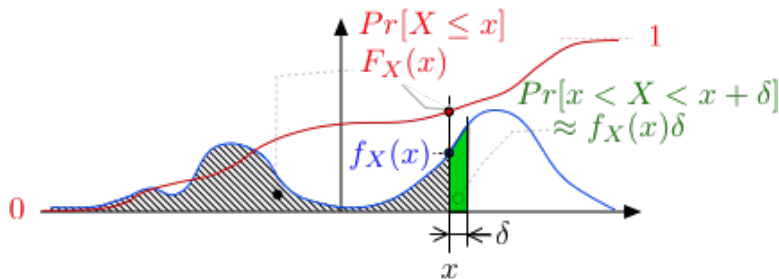
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5. Target: $f_X(x) = 2x1\{0 \leq x \leq 1\}$; $F_X(x) = x^2$ for $0 \leq x \leq 1$.