

Continuous Probability Functions

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Outline

- **Continuous RVs, PDFs, CDFs**
- **Continuous Expectations**
- **Continuous Moments**
- **Continuous Uniform PDFs, CDFs, Matlab**
- **Continuous Normal PDFs, CDFs, Matlab**

Continuous RVs, PDFs, and CDFs

Assume that the continuous random variable (RV) can take on values

$$x \in [a, b]$$

then, the probability distribution function (PDF) is given by

$$f(x | \theta) \quad \text{defined for } x \in [a, b]$$

where x can be defined within an infinite interval

and θ are any parameters that the PDF depends on.

Continuous RVs, PDFs, and CDFs

Further, the cumulative distribution function (CDF) is given by

$$F(x | \theta) = \int_{t=-\infty}^x f(t | \theta) dt$$

Additionally, any PDF must satisfy

1) $0 \leq f(x | \theta)$

2) $\int_x^\infty f(x | \theta) = 1$.

Continuous Expectation

Given an arbitrary continuous probability distribution

$f(x|\theta)$, we want to compute quantitative population

summaries of it such as:

population mean, $\mu = \int_{x=-\infty}^{\infty} xf(x|\theta)dx$

population variance, $\sigma^2 = \int_{x=-\infty}^{\infty} (x - \mu)^2 f(x|\theta)dx$

population standard deviation, $\sigma = \sqrt{\sigma^2}$

Continuous Summaries

Given an arbitrary continuous probability distribution

$f(x|\theta)$, we want to compute quantitative population

summaries of it such as:

population median \tilde{x} , $\int_{x=-\infty}^{\tilde{x}} f(x|\theta)dx = \frac{1}{2}$

population mode \hat{x} , $\left. \frac{\partial}{\partial x} f(x|\theta) \right|_{\hat{x}} = 0$

Provided f is differentiable.
Max if 2nd derivative neg at point.
Check boundary points for max.

Continuous Expectation

These population numerical summaries are found by expectation

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

The mean is

$$\mu = E(X | \theta)$$

and the variance is

$$\sigma^2 = E[(X - \mu)^2 | \theta]$$

Continuous Expectation

Linearity Property:

$$E\left[\sum_{j=1}^n a_j g_j(X) \mid \theta\right] = \sum_{j=1}^n a_j E[g_j(x) \mid \theta]$$

i.e. $E[a_1 g_1(X) + a_2 g_2(X) \mid \theta] = a_1 E[g_1(x) \mid \theta] + a_2 E[g_2(x) \mid \theta]$

$$\begin{aligned} E[a_1 g_1(X) + a_2 g_2(X) \mid \theta] &= \int_{x=-\infty}^{\infty} [a_1 g_1(x) + a_2 g_2(x)] f(x \mid \theta) dx \\ &= a_1 \int_{x=-\infty}^{\infty} g_1(x) f(x \mid \theta) dx + a_2 \int_{x=-\infty}^{\infty} g_2(x) f(x \mid \theta) dx \\ &= a_1 E[g_1(X) \mid \theta] + a_2 E[g_2(X) \mid \theta] \end{aligned}$$

Continuous Moments

For each integer n , the n^{th} moment of X , μ'_n , is

$$\mu'_n = E(X^n | \theta) .$$

The n^{th} central moment of X , μ_n , is

$$\mu_n = E[(X - \mu)^n | \theta]$$

where $\mu'_1 = E(X | \theta) .$

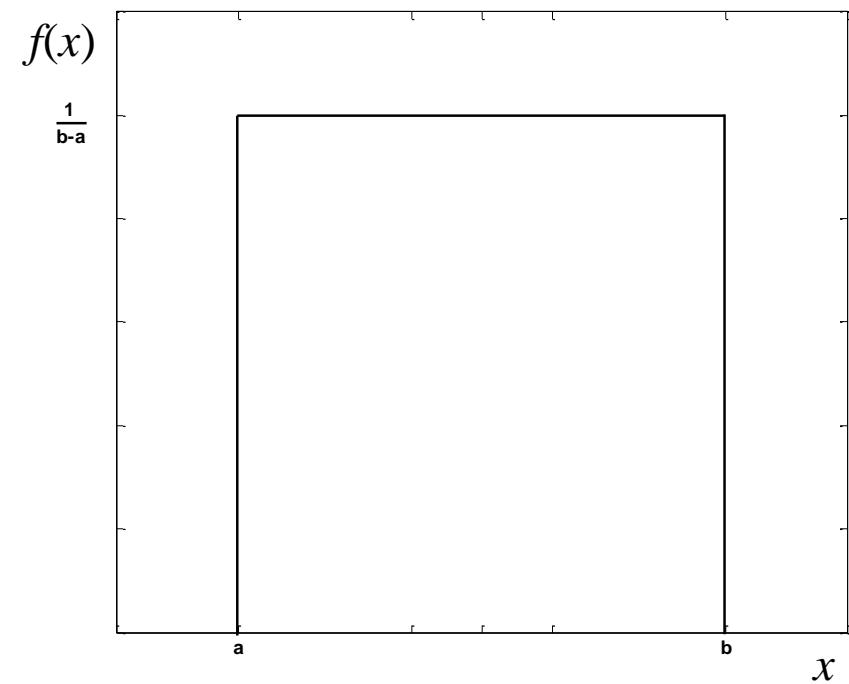
Continuous Distributions

Uniform:

A random variable x has a continuous uniform distribution, $x \sim \text{uniform}(a,b)$ if

$$f(x|a,b) = \frac{1}{b-a}, \quad x \in [a,b]$$

where, $a, b \in \mathbb{R}$, $a < b$.



Continuous Distributions

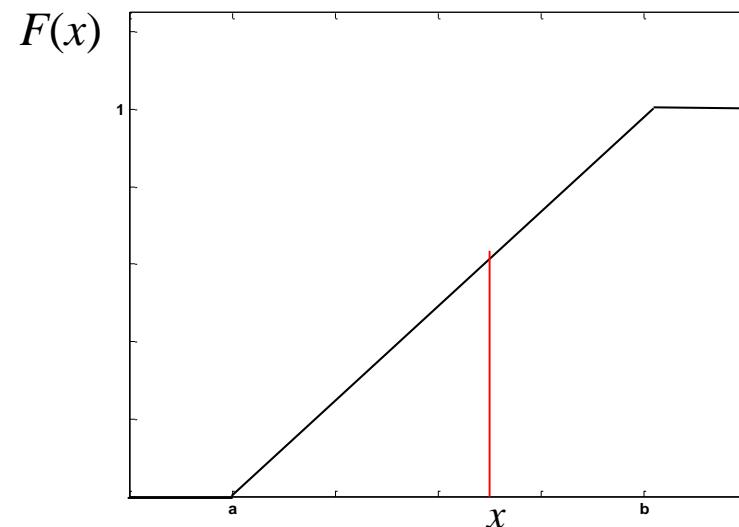
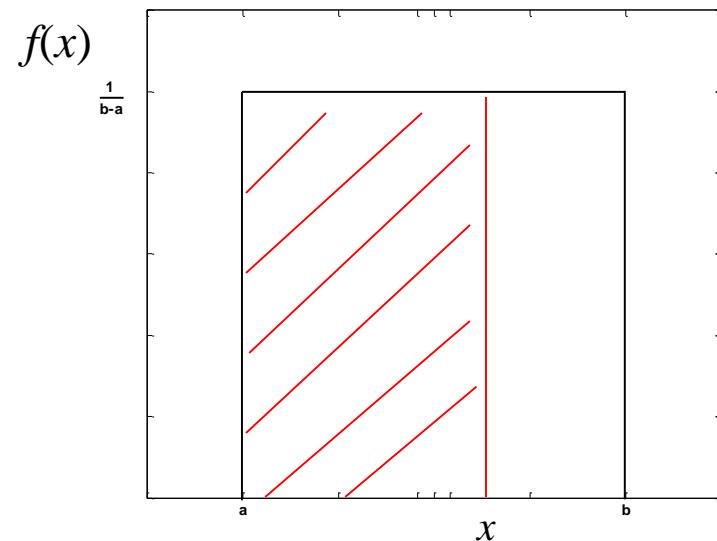
Uniform:

The CDF of the continuous uniform distribution is

$$F(x | \theta) = \int_{t=-\infty}^x f(t | \theta) dt$$

$$F(x | a, b) = \int_{t=a}^x \frac{1}{b-a} dt$$

$$= \begin{cases} 0 & x < a \\ \frac{x-a}{b-a} & x \in [a, b] \\ 1 & x > b \end{cases}$$

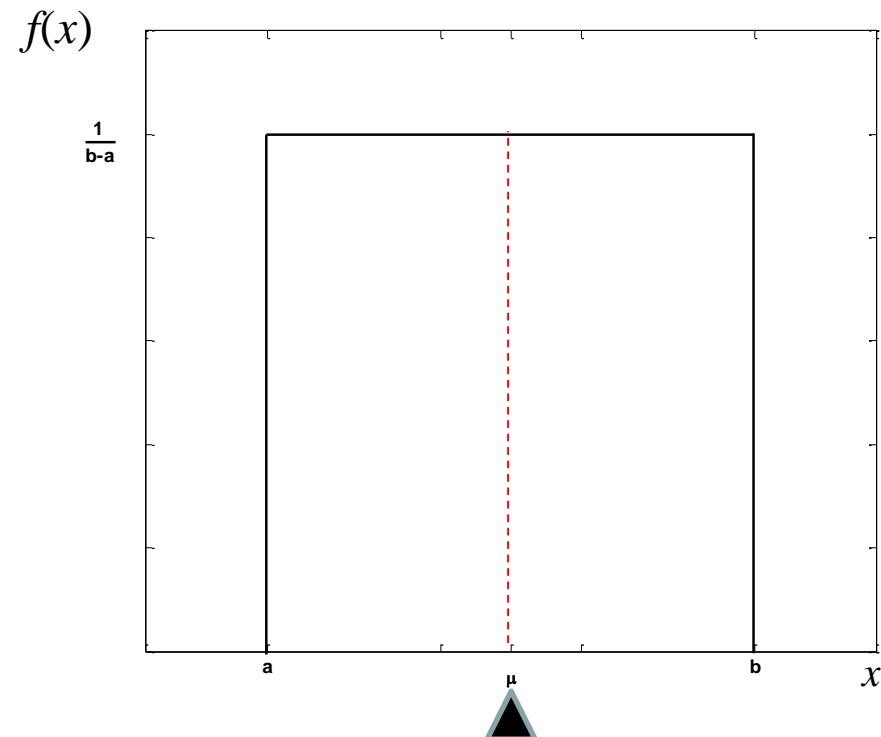


Continuous Distributions

Uniform:

It can be shown that

$$\begin{aligned}\mu &= \int_x xf(x | \theta) dx \\ &= \int_{x=a}^b x \frac{1}{b-a} dx \\ &= \frac{b+a}{2}\end{aligned}$$



Continuous Distributions

Uniform:

It can be shown that

median

$$\int_{x=a}^{\tilde{x}} f(x | \theta) dx = \frac{1}{2}$$

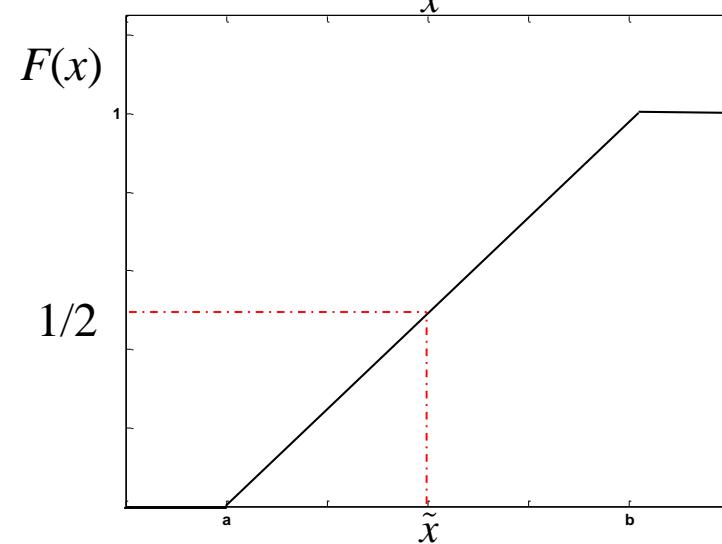
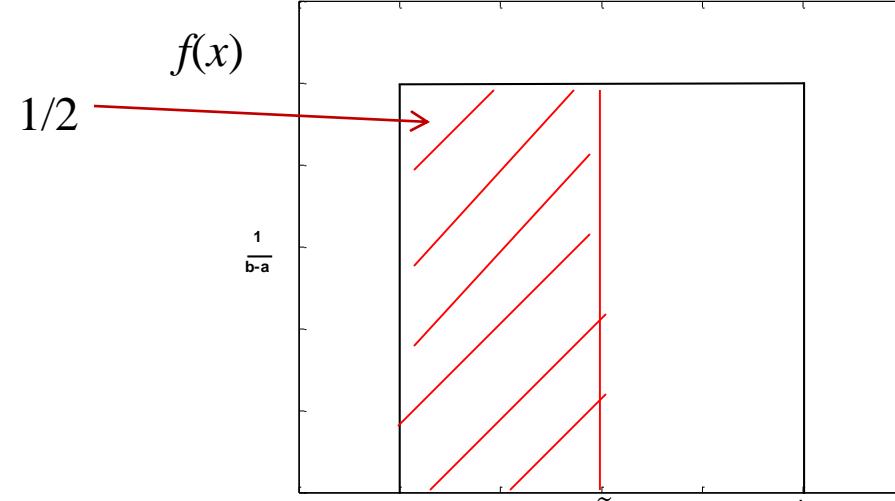
$$\tilde{x} = \frac{b+a}{2}$$

mode

$$\left. \frac{\partial}{\partial x} f(x | \theta) \right|_{\hat{x}} = 0$$

\hat{x} = any value $\in [a, b]$

flat distribution

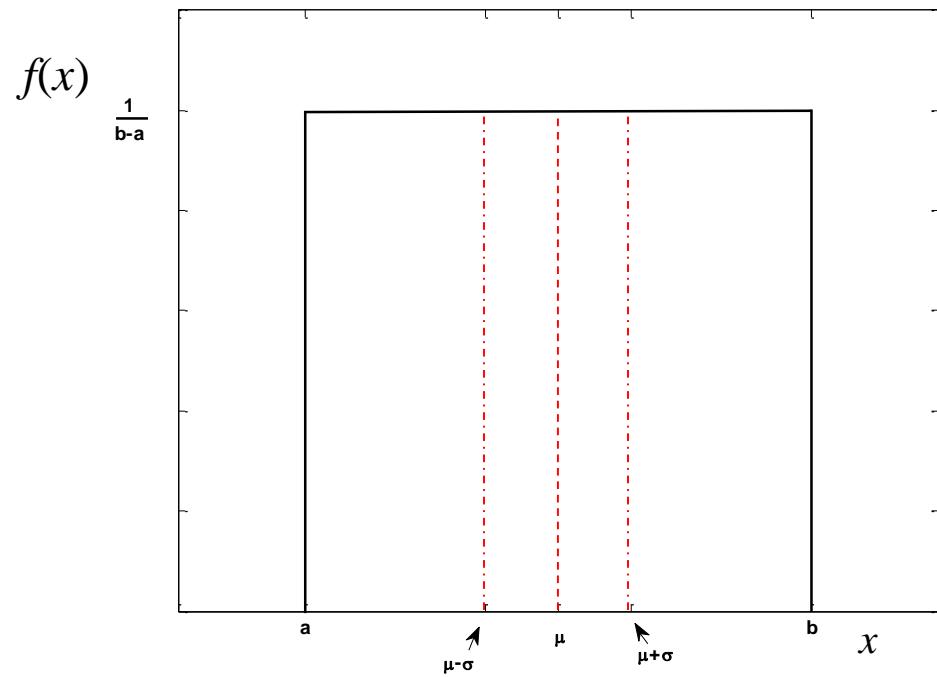


Continuous Distributions

Uniform:

that

$$\begin{aligned}\sigma^2 &= \int_x (x - \mu)^2 f(x | \theta) dx \\ &= \int_{x=a}^b (x - \mu)^2 \frac{1}{b-a} dx \\ &= \frac{(b-a)^2}{12} \\ \sigma &= \frac{(b-a)}{\sqrt{12}}\end{aligned}$$



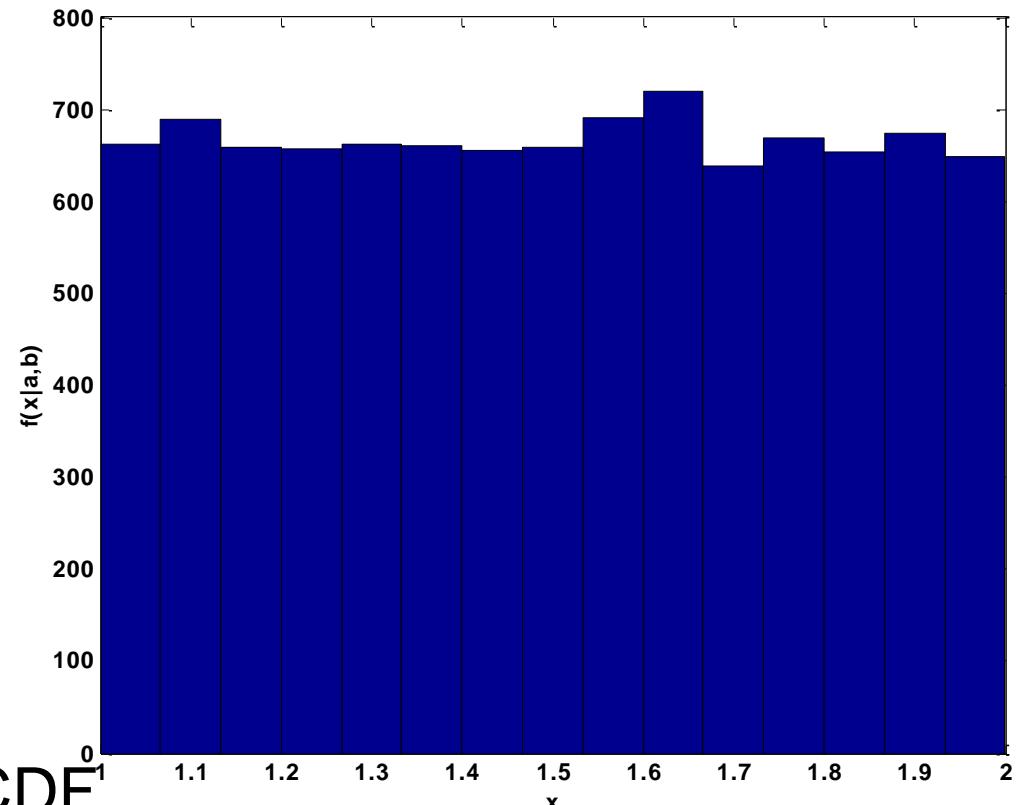
Continuous Distributions

Uniform:

$$f(x | a, b) = \frac{1}{b - a}$$

```
a=1; b=2; num=10^4;
x=a+(b-a)*rand(num,1);
mean(x)
var(x)
hist(x,15)
```

True	Simulated
μ	1.5
σ^2	0.0833

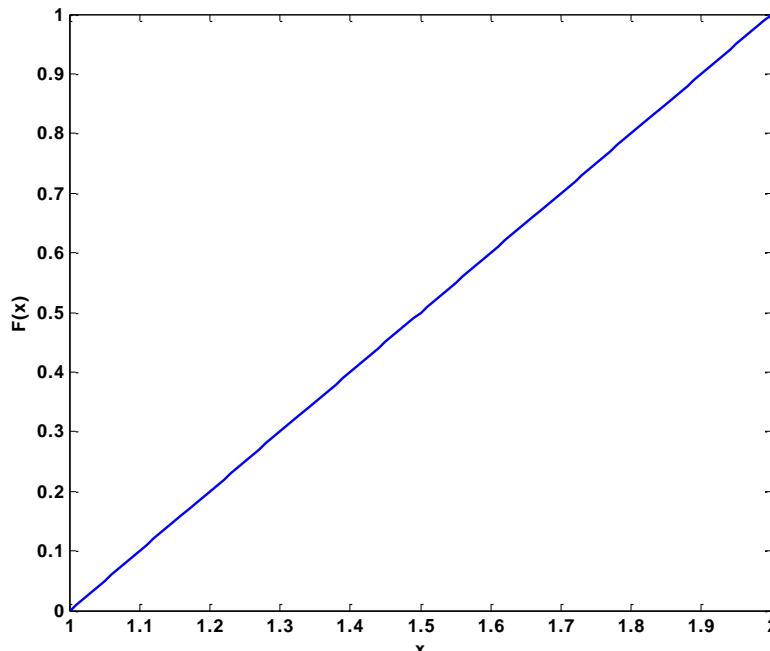


Can also find and plot ECDF.

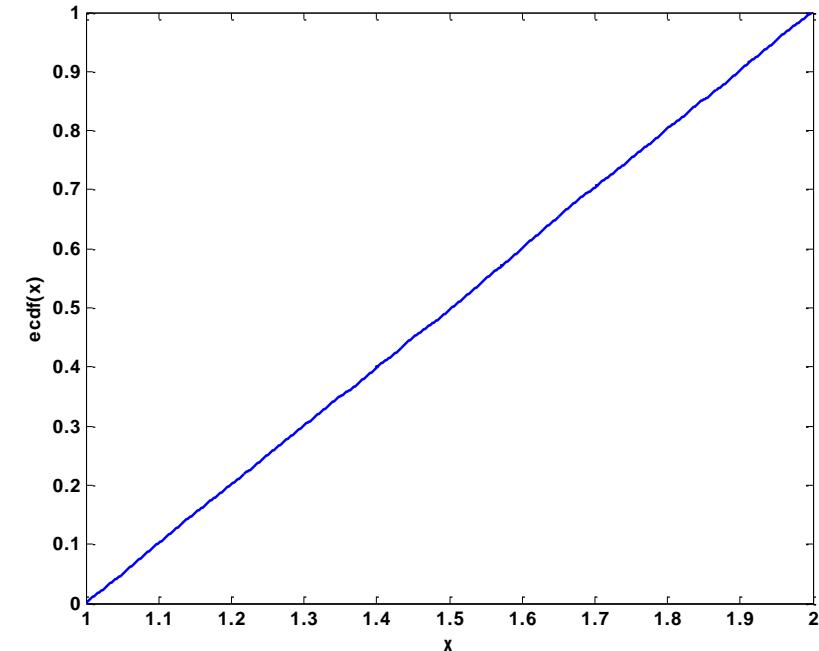
Continuous Distributions

Uniform:

$$f(x | a, b) = \frac{1}{b - a}$$



```
a=1;;b=2;;
y=cdf('unif',(1:.01:2),a,b)
plot((1:.01:2),y)
```



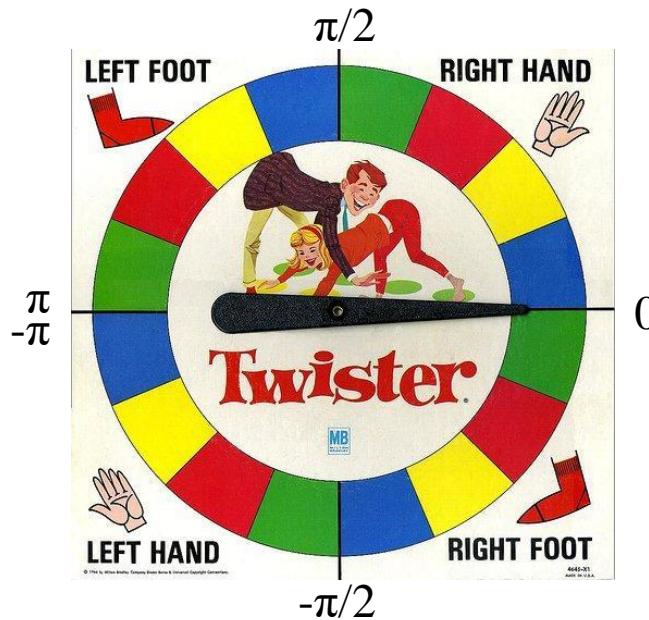
```
[F,xx]=ecdf(x);
stairs(xx,F,'LineWidth',2)
```

Continuous Distributions

Uniform:

$$f(x | a, b) = \frac{1}{b - a}$$

We are playing a board game. See spinner below. Assume that resulting angle x from a spin is uniformly distributed. i.e. Let $x \sim \text{uniform}(a = -\pi, b = \pi)$.



Find $P(\text{red})$.

FYI there are circular distributions!

Continuous Distributions

Normal:

A random variable x has a continuous normal distribution, $x \sim \text{normal}(\mu, \sigma^2)$ if

$$f(x | \mu, \sigma^2) = \frac{e^{-\frac{(x-\mu)^2}{2\sigma^2}}}{\sqrt{2\pi\sigma^2}}, \quad x \in \mathbb{R}$$

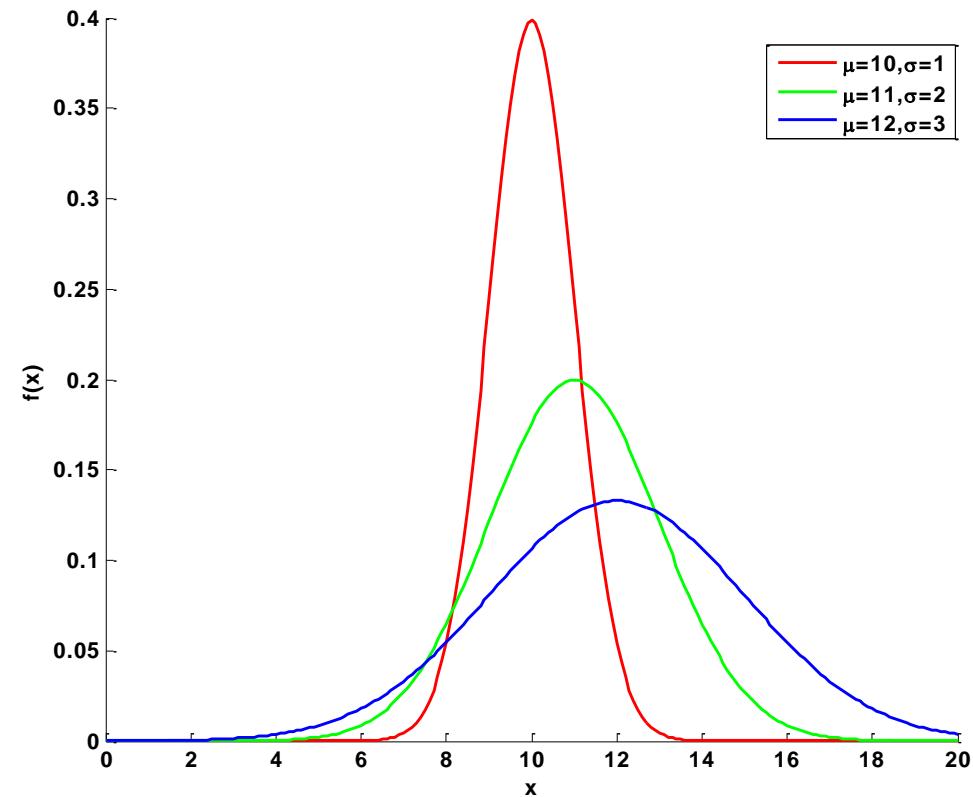
where, $-\infty < \mu < \infty$, and $0 < \sigma$.

Continuous Distributions

Normal:

```
x=(0:.1:20)';
mu=[10,11,12];, sigma=[1,2,3];
figure(1)
hold on
for count=1:length(sigma)
    y = normpdf(x,mu(count),sigma(count));
    if count==1
        plot(x,y,'r','LineWidth',2)
    elseif count==2
        plot(x,y,'g','LineWidth',2)
    elseif count==3
        plot(x,y,'b','LineWidth',2)
    end
end
xlim([0 20])
```

$$f(x | \mu, \sigma^2) = \frac{e^{-\frac{(x-\mu)^2}{2\sigma^2}}}{\sqrt{2\pi\sigma^2}}$$



Continuous Distributions

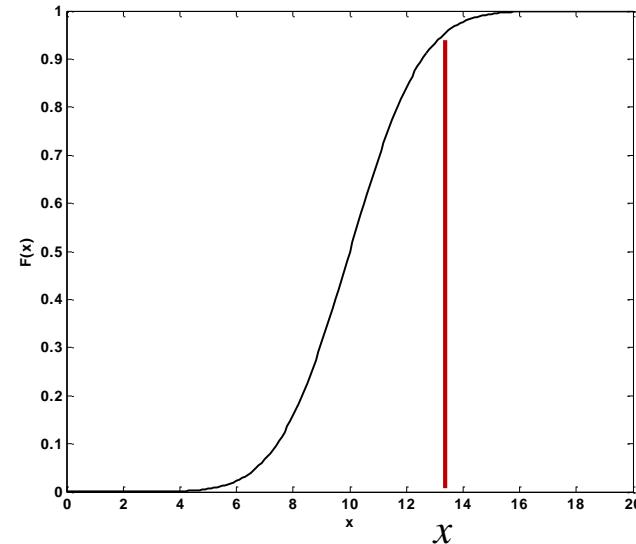
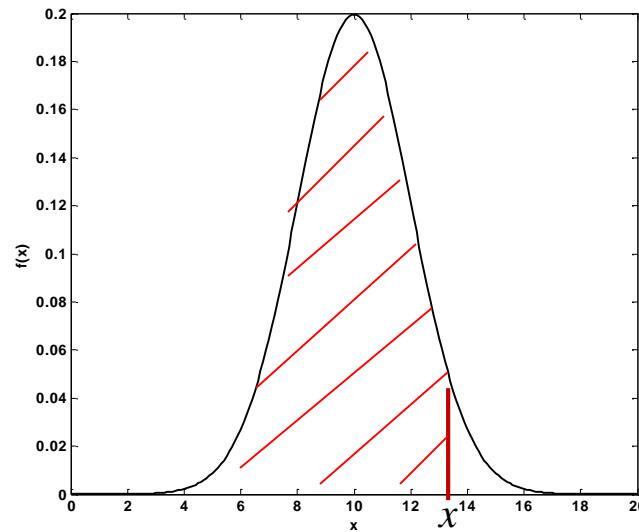
Normal:

The CDF of the continuous normal distribution is

$$F(x | \theta) = \int_{t=-\infty}^x f(t | \theta) dt$$

$$F(x | \mu, \sigma^2) = \int_{t=-\infty}^x \frac{e^{-\frac{(t-\mu)^2}{2\sigma^2}}}{\sqrt{2\pi\sigma^2}} dt$$

No closed form analytic solution.

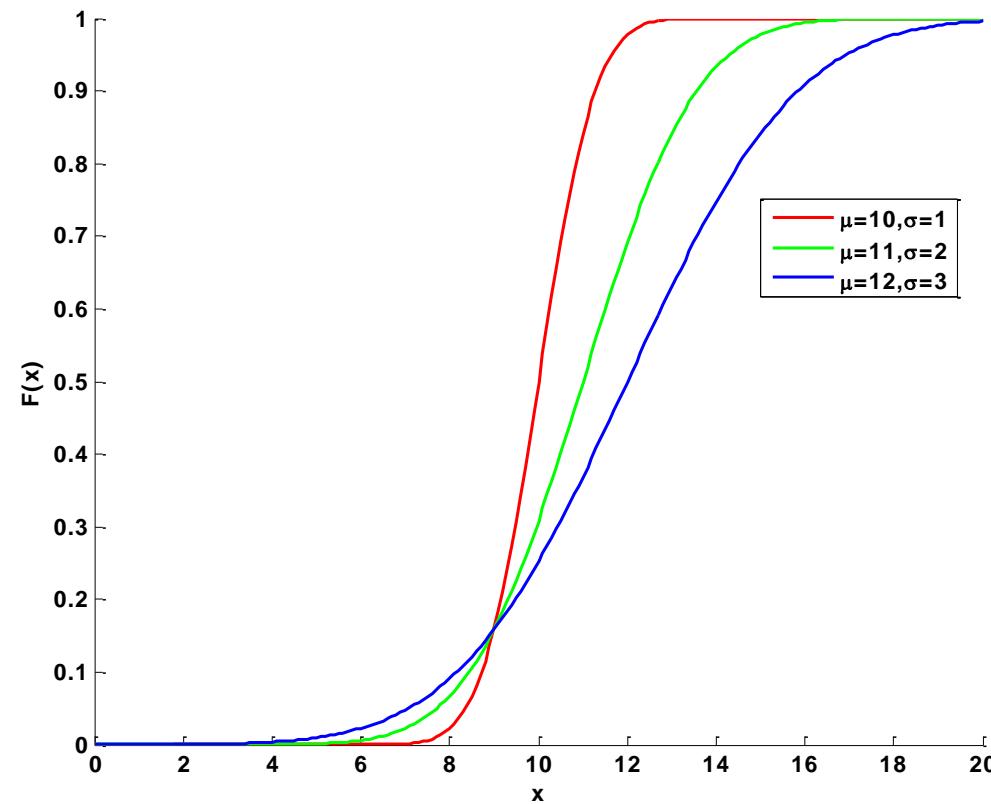


Continuous Distributions

Normal:

```
x=(0:.1:20)';
mu=[10,11,12];, sigma=[1,2,3];
figure(1)
hold on
for count=1:length(sigma)
    y = normcdf(x,mu(count),sigma(count));
    if count==1
        plot(x,y,'r','LineWidth',2)
    elseif count==2
        plot(x,y,'g','LineWidth',2)
    elseif count==3
        plot(x,y,'b','LineWidth',2)
    end
end
xlim([0 20])
```

$$f(x | \mu, \sigma^2) = \frac{e^{-\frac{(x-\mu)^2}{2\sigma^2}}}{\sqrt{2\pi\sigma^2}}$$

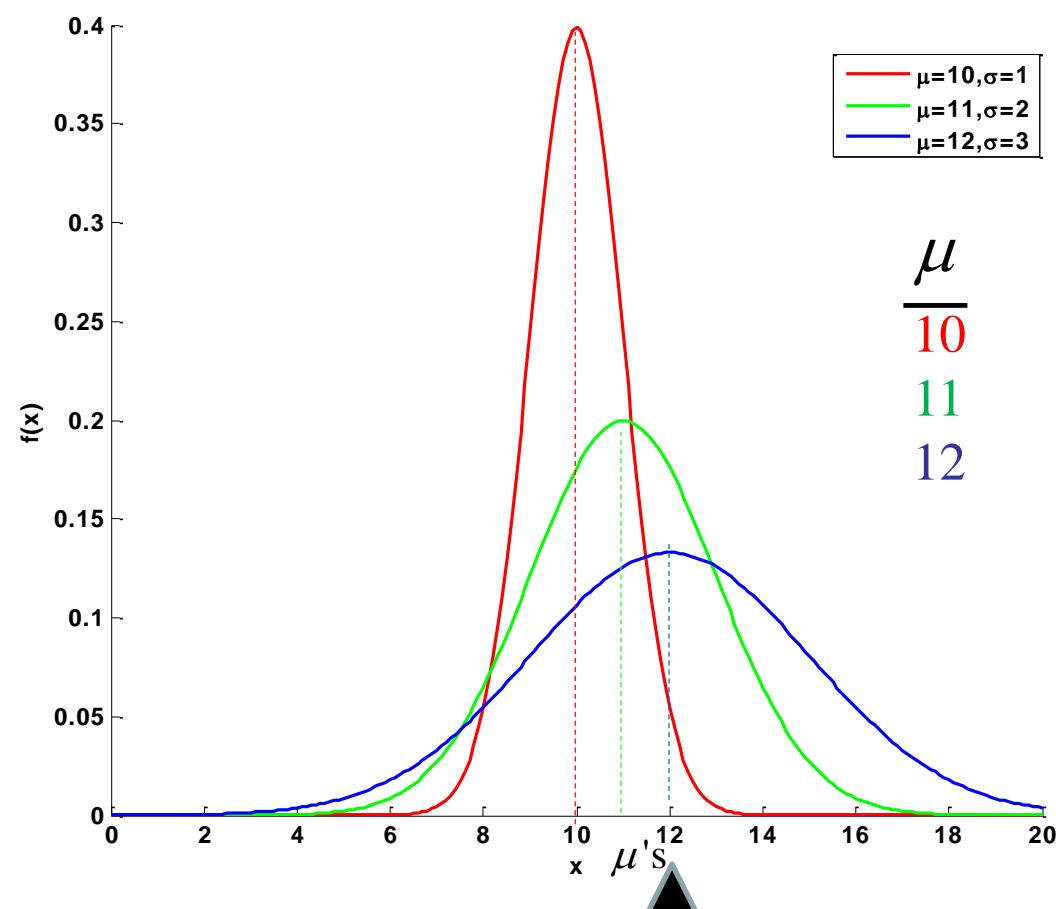


Continuous Distributions

Normal:

It can be shown that

$$\begin{aligned}\mu &= \int_x xf(x | \theta)dx \\ &= \int_{x=-\infty}^{\infty} x \frac{e^{-\frac{(x-\mu)^2}{2\sigma^2}}}{\sqrt{2\pi\sigma^2}} dx\end{aligned}$$



Continuous Distributions

Normal:

It can be shown that

median

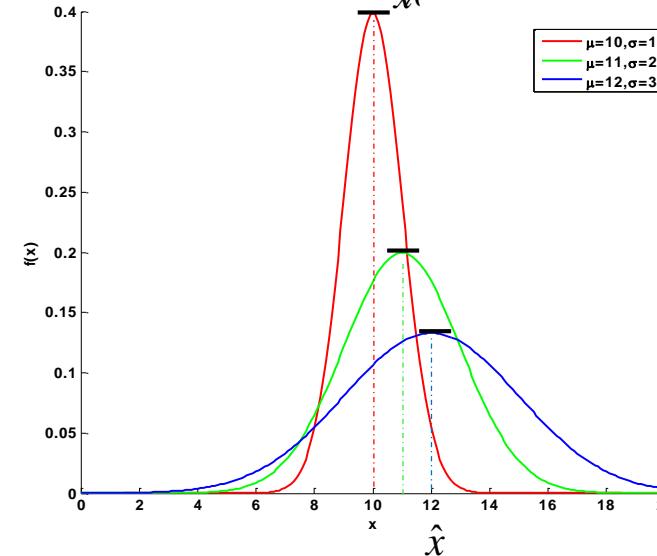
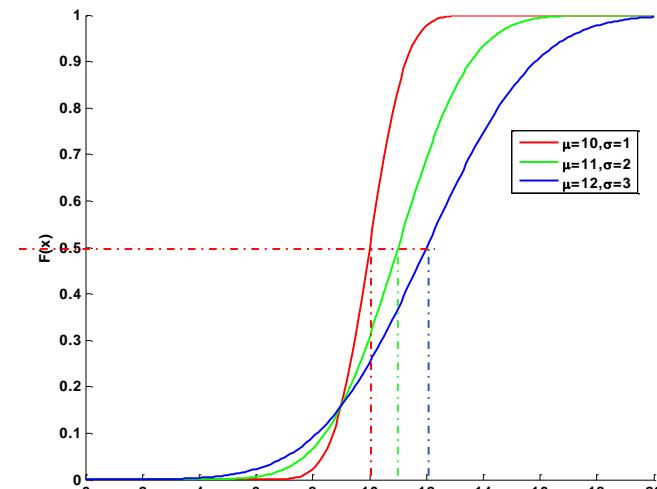
$$\int_{x=-\infty}^{\tilde{x}} f(x | \theta) dx = \frac{1}{2}$$

$$\tilde{x} = \mu$$

mode

$$\left. \frac{\partial}{\partial x} f(x | \theta) \right|_{\hat{x}} = 0$$

$$\hat{x} = \mu$$

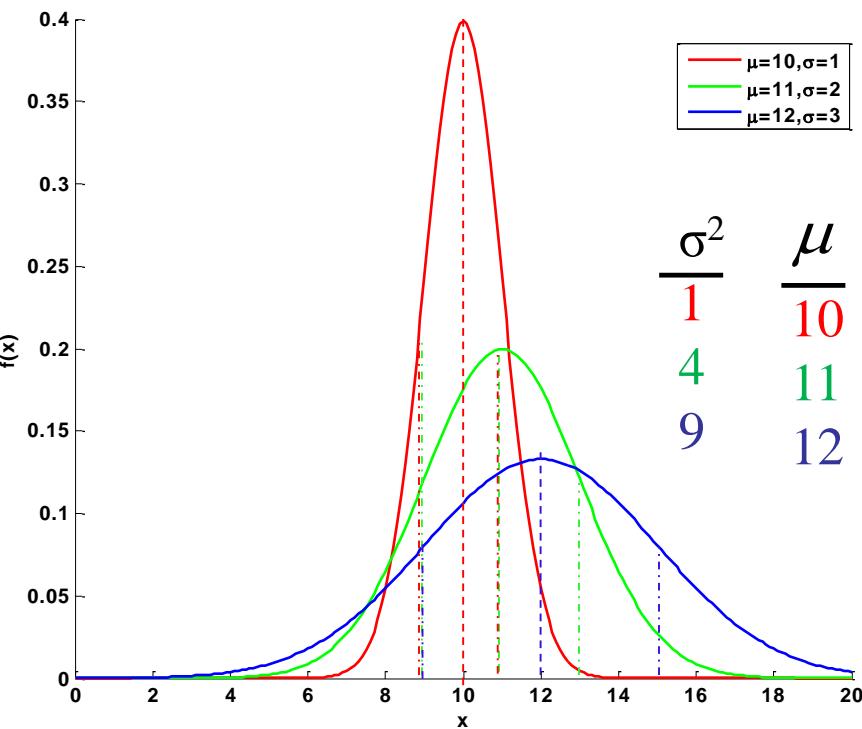


Continuous Distributions

Normal:

that

$$\begin{aligned}\sigma^2 &= \int_x (x - \mu)^2 f(x | \theta) dx \\ &= \int_{x=-\infty}^{\infty} (x - \mu)^2 \frac{e^{-\frac{(x-\mu)^2}{2\sigma^2}}}{\sqrt{2\pi\sigma^2}} dx\end{aligned}$$



Continuous Distributions

Normal:

$$f(x | \mu, \sigma^2) = \frac{e^{-\frac{(x-\mu)^2}{2\sigma^2}}}{\sqrt{2\pi\sigma^2}}$$

```
mu=11; sigma=2; num=10^4;
```

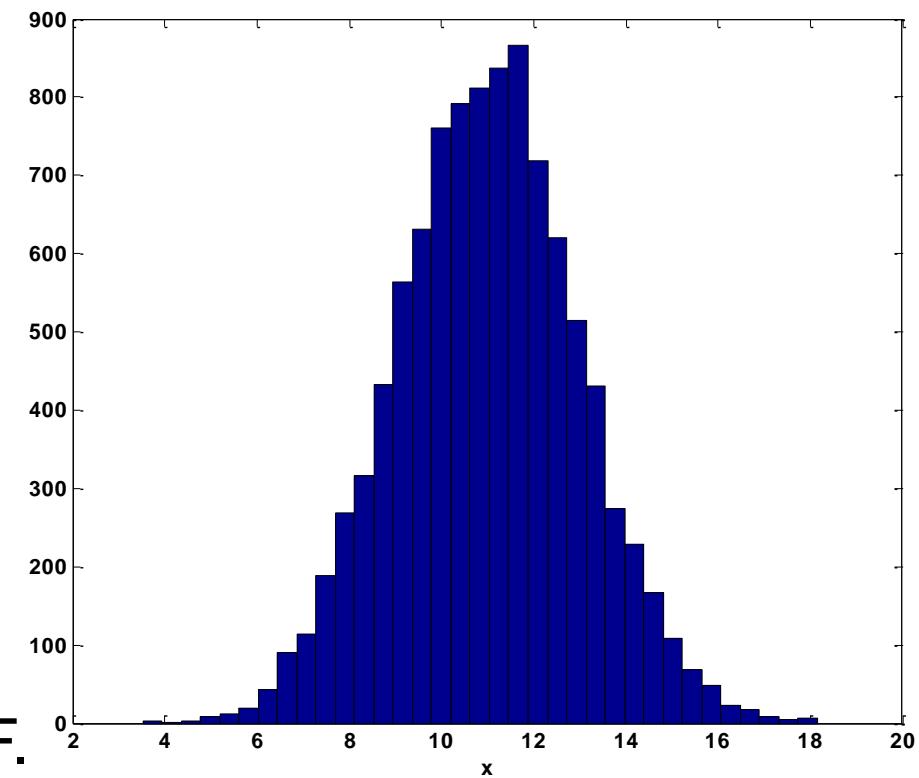
```
x=normrnd(mu,sigma,num,1);
```

```
mean(x)
```

```
var(x)
```

```
hist(x,35)
```

True	Simulated
μ	11
σ^2	4



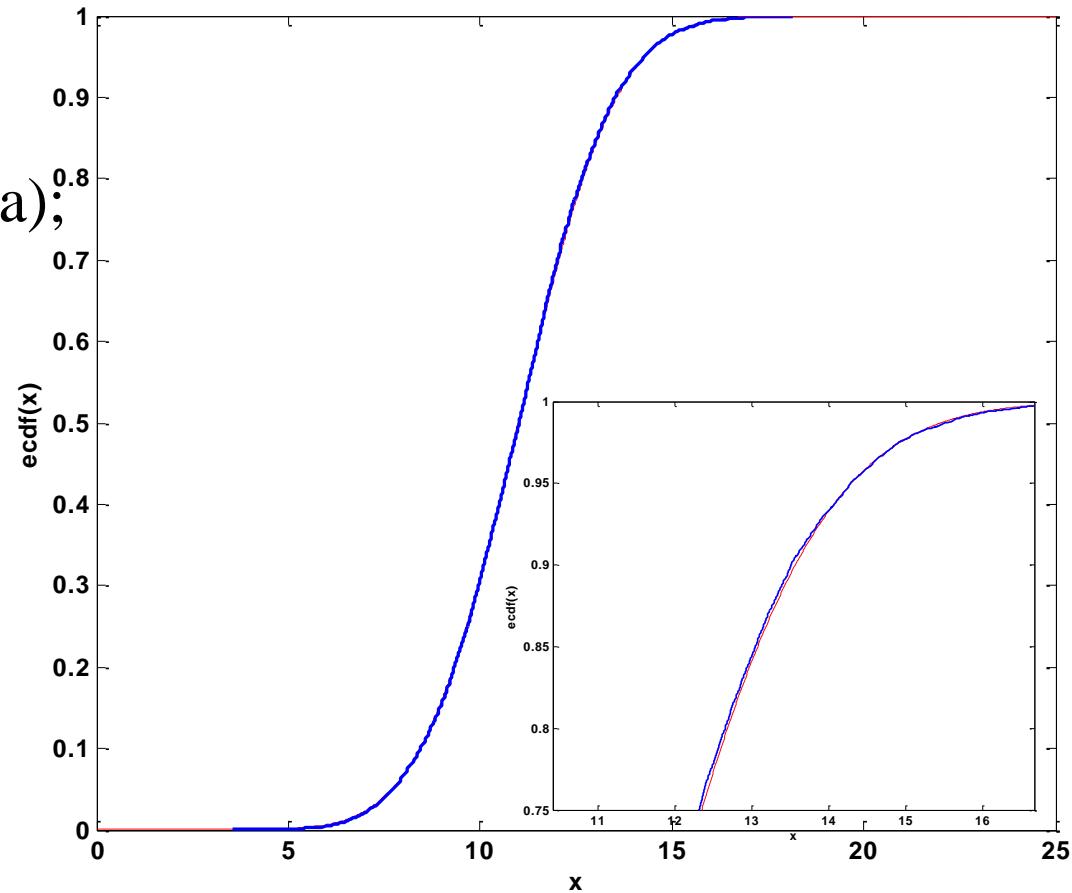
Can also find and plot ECDF.

Continuous Distributions

Normal:

$$f(x | \mu, \sigma^2) = \frac{e^{-\frac{(x-\mu)^2}{2\sigma^2}}}{\sqrt{2\pi\sigma^2}}$$

```
mu=11; sigma=2;  
y=normcdf((0:.01:25),mu,sigma);  
plot((0:.01:25),y,'r')  
hold on  
[F,xx]=ecdf(x);  
stairs(xx,F,'LineWidth',2)
```



Homework 3:

- 1) Show analytically that $P(\mu-\sigma \leq x \leq \mu+\sigma) = 1/\sqrt{3}$ for a uniform distribution. First find μ and σ .
- 2) Numerically integrate (rectangles) the uniform $(0,1)$ PDF from $\mu-\sigma$ to $\mu+\sigma$ with $\mu=1/2$ and $\sigma=1/\sqrt{12}$.
- 3) Generate $n=10^6$ random $U(0,1)$ observations.
What fraction are within $1/2 - 1/\sqrt{12}$ to $1/2 + 1/\sqrt{12}$.
- 4) Make an empirical CDF from the values in 3).
(Note that this is a Monte Carlo technique for determining percentiles and thus p -values.)

Homework 3:

- 5) Analytically show that the normal pdf integrates to 1.
- 6) Analytically show that the mean of normal pdf is μ .
- 7) Numerically integrate the normal pdf to find the 50th and 99th percentiles for $\mu=11$, $\sigma=2$.
i.e. find x_0 such that $P(x \leq x_0) = .50$ and $P(x \leq x_0) = .99$
- 8) Generate 10^6 normal $\mu=11$, $\sigma=2$ random variables.
Empirically determine the 50th and 99th percentiles.
i.e. Find the $.50 * 10^6$ and $.99 * 10^6$ largest values x_0 .
such that $P(x \leq x_0) = .50$ and $P(x \leq x_0) = .99$. Compare to 5).

Homework 3:

- 9) Make a histogram of the random variables in 6)
with at least 50 bins.

- 10) Make an empirical CDF from the values in 6).
(Note that this is a Monte Carlo technique for determining percentiles
and thus p -values.)