

# Continuous Probability Density Functions

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# Outline

**Properties and Summaries**

**Uniform and Beta PDF**

**Normal PDF**

**Inverse Gamma**

**Discussion**

**Homework**

# Properties and Summaries

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

$$x \in (a, b)$$

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

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

where  $x$  can be defined within an infinite interval

and  $\theta$  are any parameters that the PDF depends on.

# Properties and Summaries

The accumulation of probability from left to right

is called the cumulative distribution function (PDF)

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

and

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

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

# Properties and Summaries

Given an arbitrary continuous probability density

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

summaries of it such as:

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

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

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

# Properties and Summaries

Given an arbitrary continuous probability density

$f(x|\theta)$ , 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 2<sup>nd</sup> derivative neg at point.

Check boundary points for max.

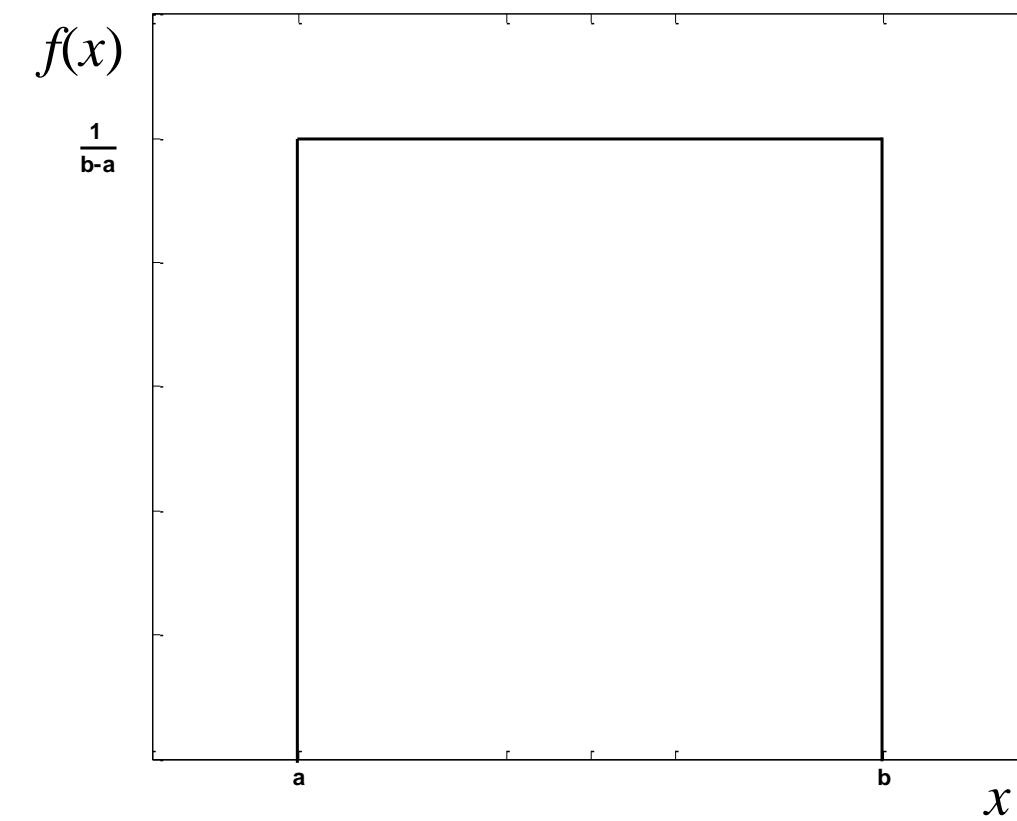
May need to be numerically calculated.

# Uniform and Beta PDF

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,  $a < x < b$ ,  $a < b$ .

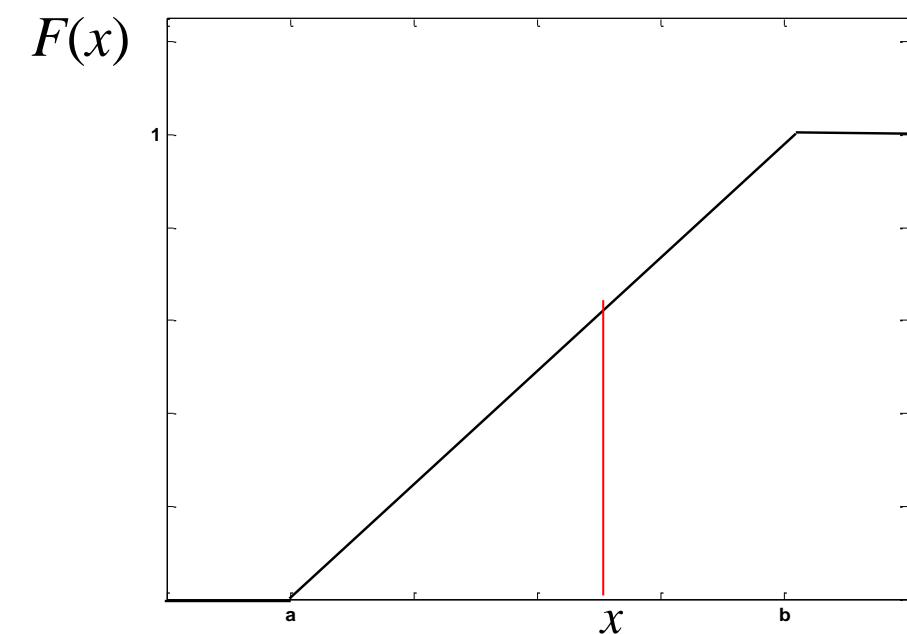
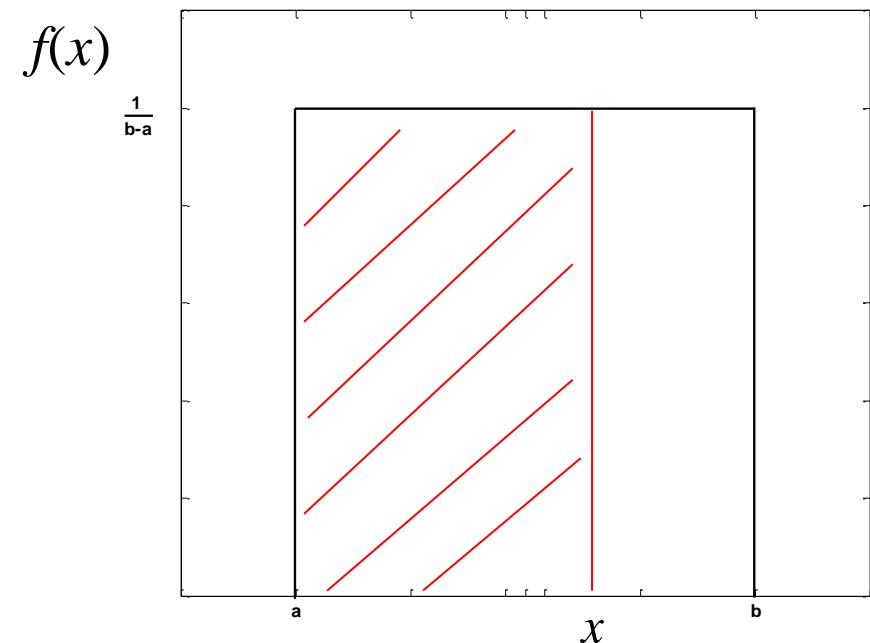


# Uniform and Beta PDF

The CDF of the continuous uniform distribution is

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

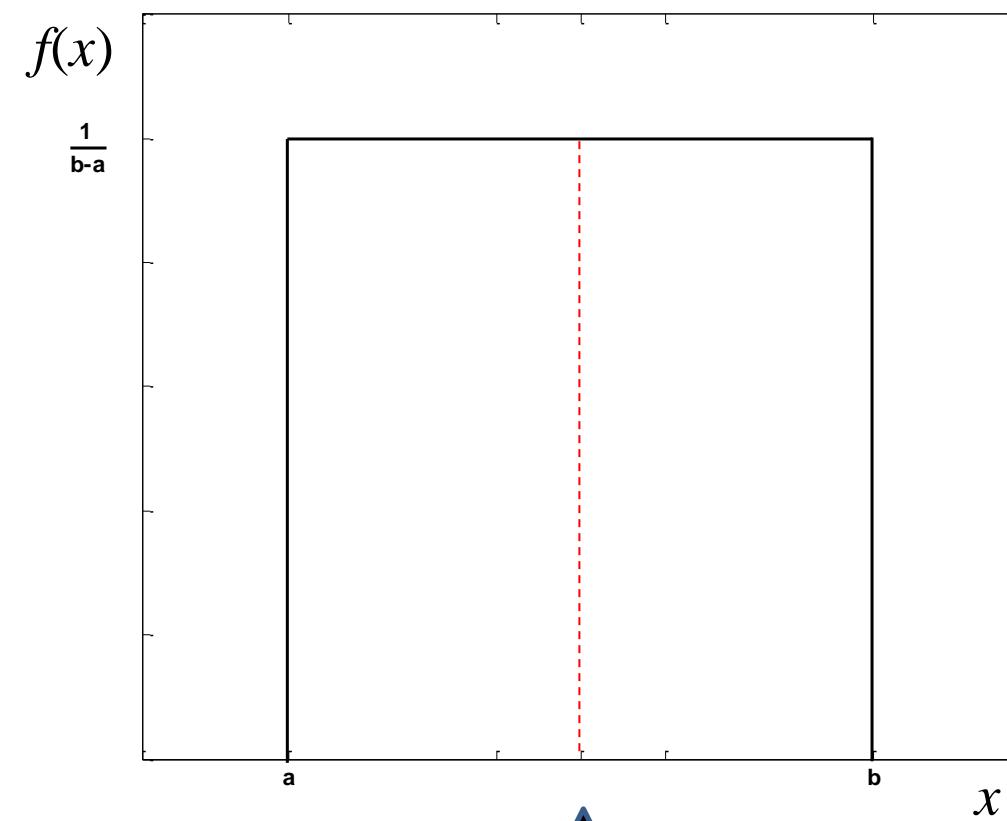
$$\begin{aligned} F(x | a, b) &= \int_{t=a}^x \frac{1}{b-a} dt \\ &= \begin{cases} 0 & x < a \\ \frac{x-a}{b-a} & x \in [a, b] \\ 1 & x > b \end{cases} \end{aligned}$$



# Uniform and Beta PDF

It can be shown that

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



# Uniform and Beta PDF

It can be shown that

median

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

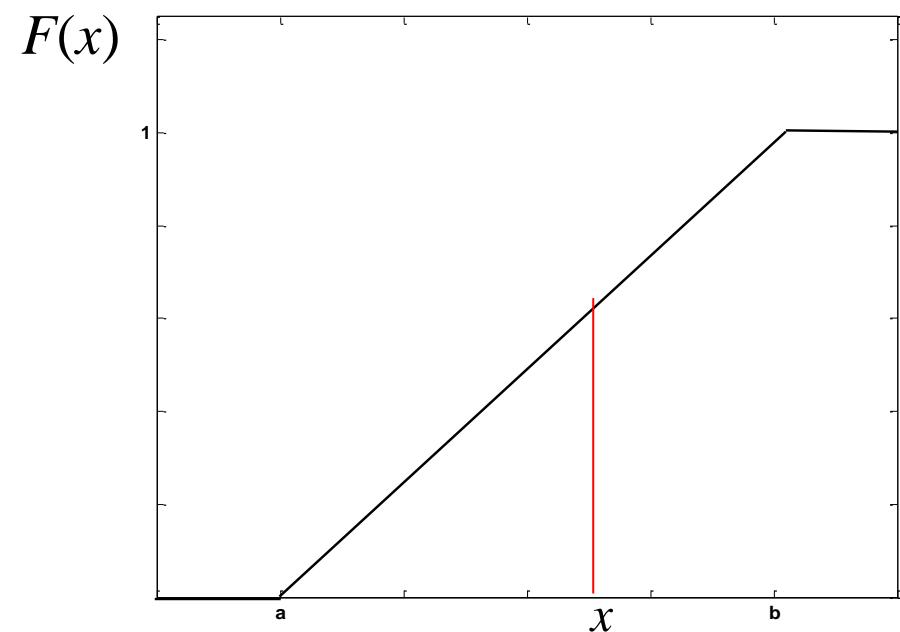
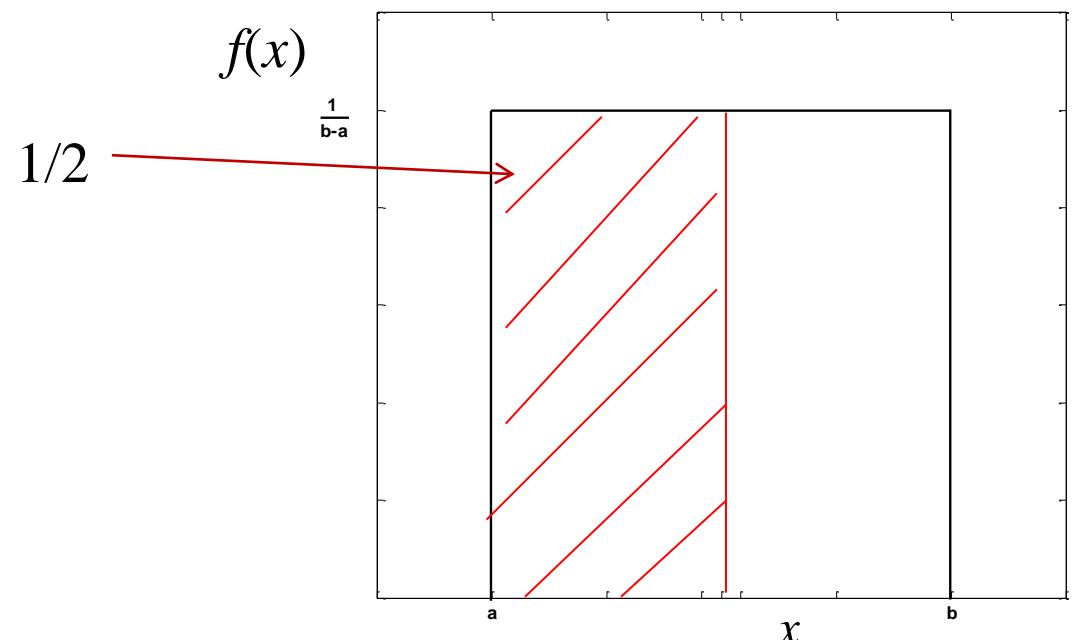
$$\tilde{x} = \frac{b+a}{2}$$

mode

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

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

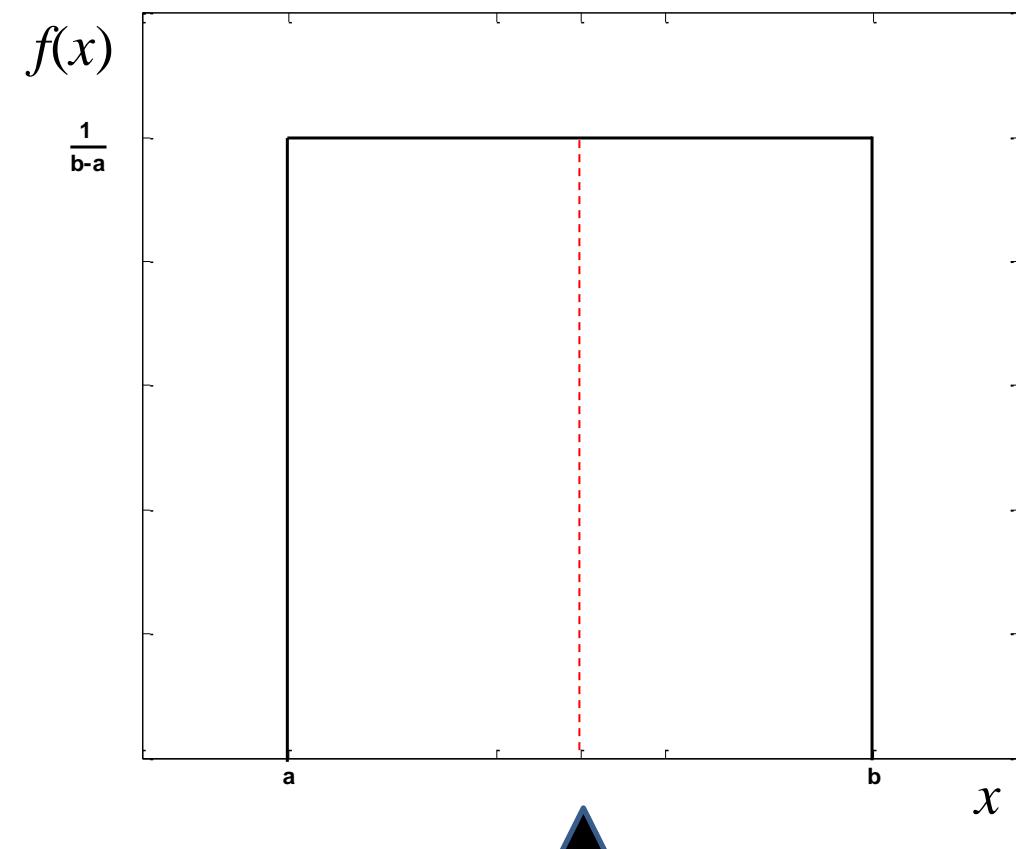
flat distribution



# Uniform and Beta PDF

that

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



# Uniform and Beta PDF

```

rng('default'), a=1;,b=2;,num=10^4;
x=a+(b-a)*rand(num,1);
figure;
histogram(x,'BinLimits',[1,2],'normalization','pdf','FaceColor','blue')
hold on
line([1,2],[1,1],'Color','red','LineWidth',1.5)
xlim([1,2])

```

True	Simulated
$\mu$	1.4996
$\sigma^2$	0.0829

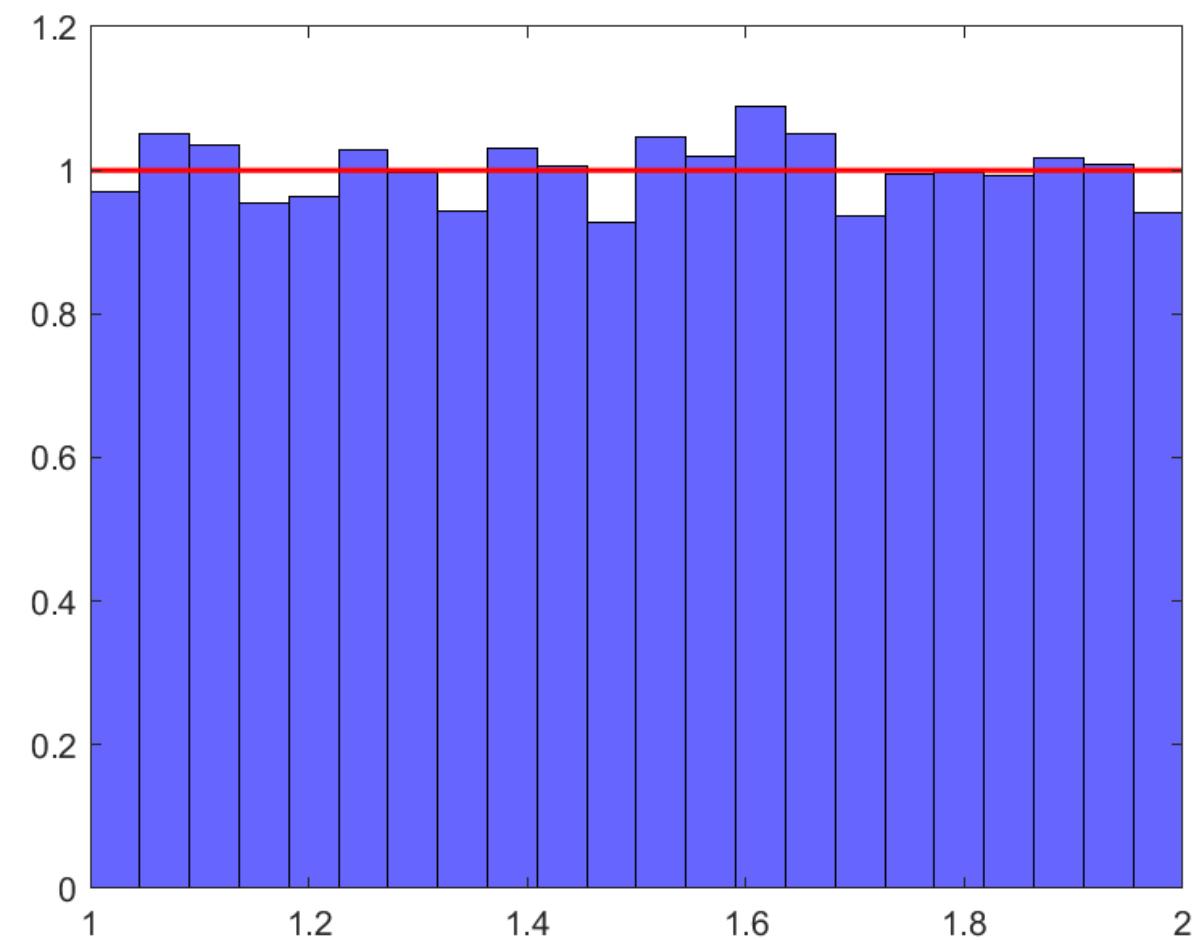
Can also find and plot ECDF.

```

themean=(b-a)/2
thevar=(b-a)^2/12
[mean(x),var(x)]

```

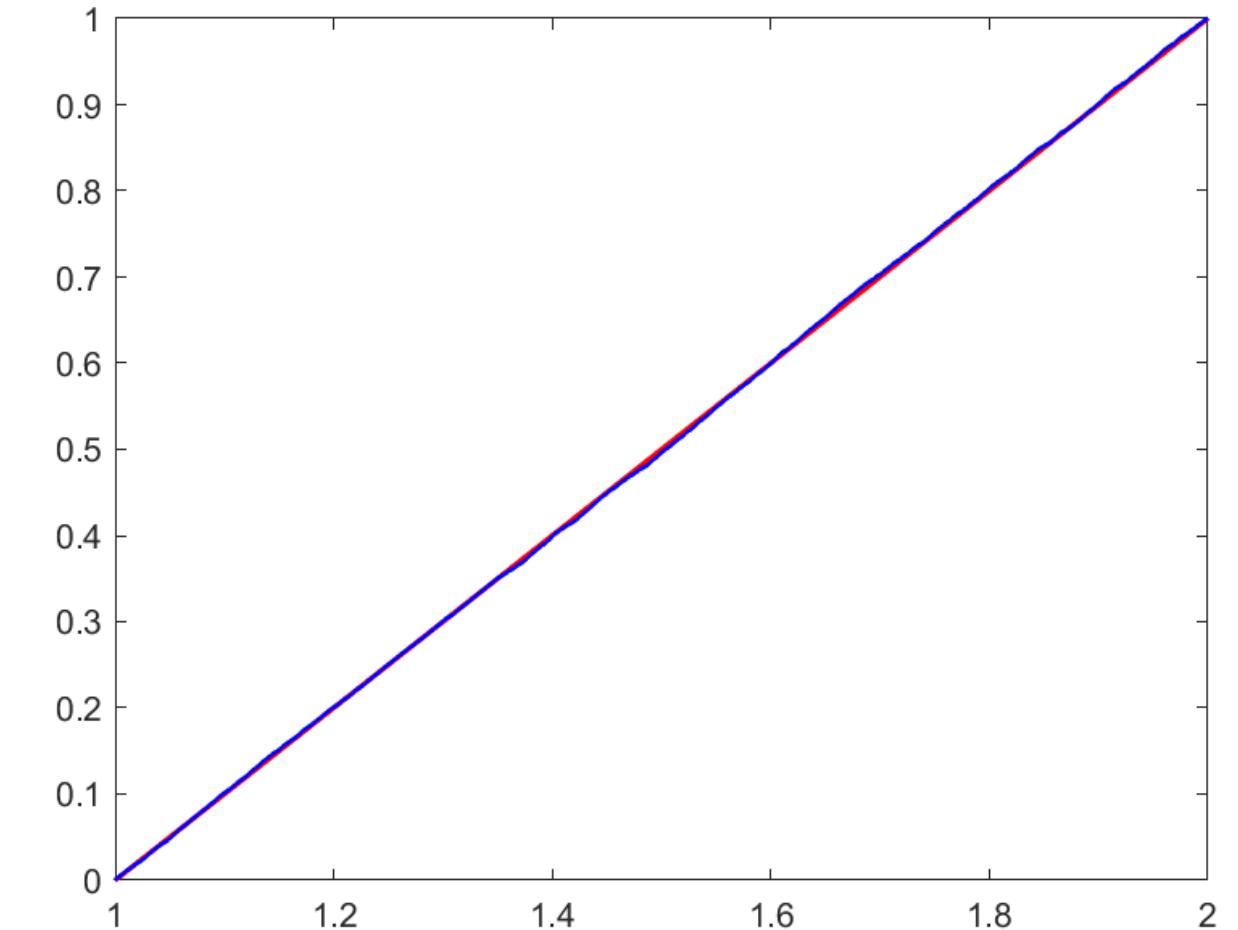
$$f(x | a, b) = \frac{1}{b - a}$$



# Uniform and Beta PDF

```
y =cdf('unif',(a:.01:b),a,b);
figure;
plot(1:.01:2),y,'Color','red','LineWidth',1.5)
[F,xx]=ecdf(x);
hold on
stairs(xx,F,'Color','blue','LineWidth',1.5)
```

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



# Uniform and Beta PDF

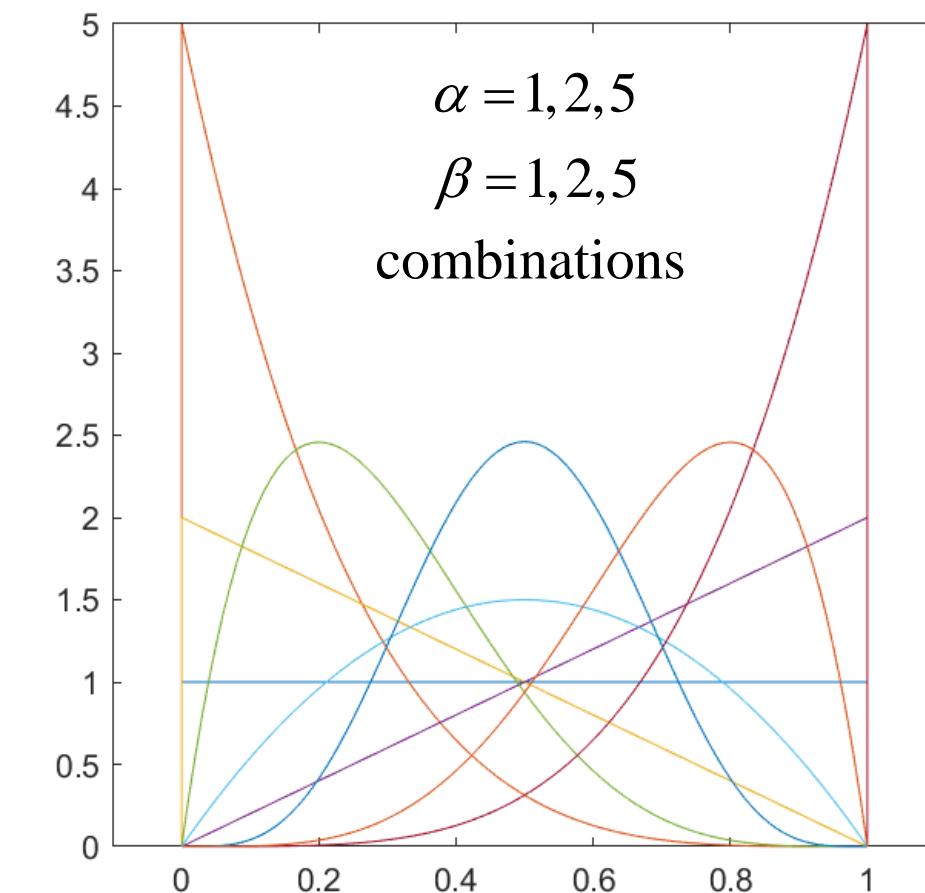
A random variable  $x$  has a continuous Beta distribution,  $x \sim \text{Beta}(\alpha, \beta)$  if

$$f(x | \alpha, \beta) = \frac{\Gamma(\alpha + \beta)}{\Gamma(\alpha)\Gamma(\beta)} x^{\alpha-1} (1-x)^{\beta-1},$$

where,  $\alpha, \beta > 0$ ,  $x \in [a, b]$ .

And can take on many shapes.

$$\Gamma(a) = (a - 1)! \text{ for integer } a$$



# Uniform and Beta PDF

There is no closed form analytic formula for the CDF  $F(x|\alpha,\beta)$ .

It can be shown that the mean is

$$\begin{aligned}\mu &= \int_x xf(x|\theta)dx \\ &= \int_{x=0}^1 x \frac{\Gamma(\alpha + \beta)}{\Gamma(\alpha)\Gamma(\beta)} x^{\alpha-1} (1-x)^{\beta-1} dx \\ &= \frac{\alpha}{\alpha + \beta}\end{aligned}$$

# Uniform and Beta PDF

It can be shown that the variance is

$$\begin{aligned}\sigma^2 &= \int_x (x - \mu)^2 f(x | \theta) dx \\ &= \int_{x=0}^1 (x - \mu)^2 \frac{\Gamma(\alpha + \beta)}{\Gamma(\alpha)\Gamma(\beta)} x^{\alpha-1} (1-x)^{\beta-1} dx \\ &= \frac{\alpha\beta}{(\alpha + \beta)^2(\alpha + \beta + 1)}\end{aligned}$$

# Uniform and Beta PDF

```

rng('default'), a=2; b=5; num=10^4;
x=betarnd(a,b,[num,1]);
figure;
histogram(x,'BinLimits',[0,1],'normalization','pdf','FaceColor','blue')
hold on
fx = @(x) factorial(a+b-1)/factorial(a-1)/factorial(b-1)...
.*x.^(a-1).*(1-x).^(b-1);
fplot(fx,[0,1],'r','LineWidth',1.5)

```

True	Simulated
$\mu$	0.2857
$\sigma^2$	0.0255

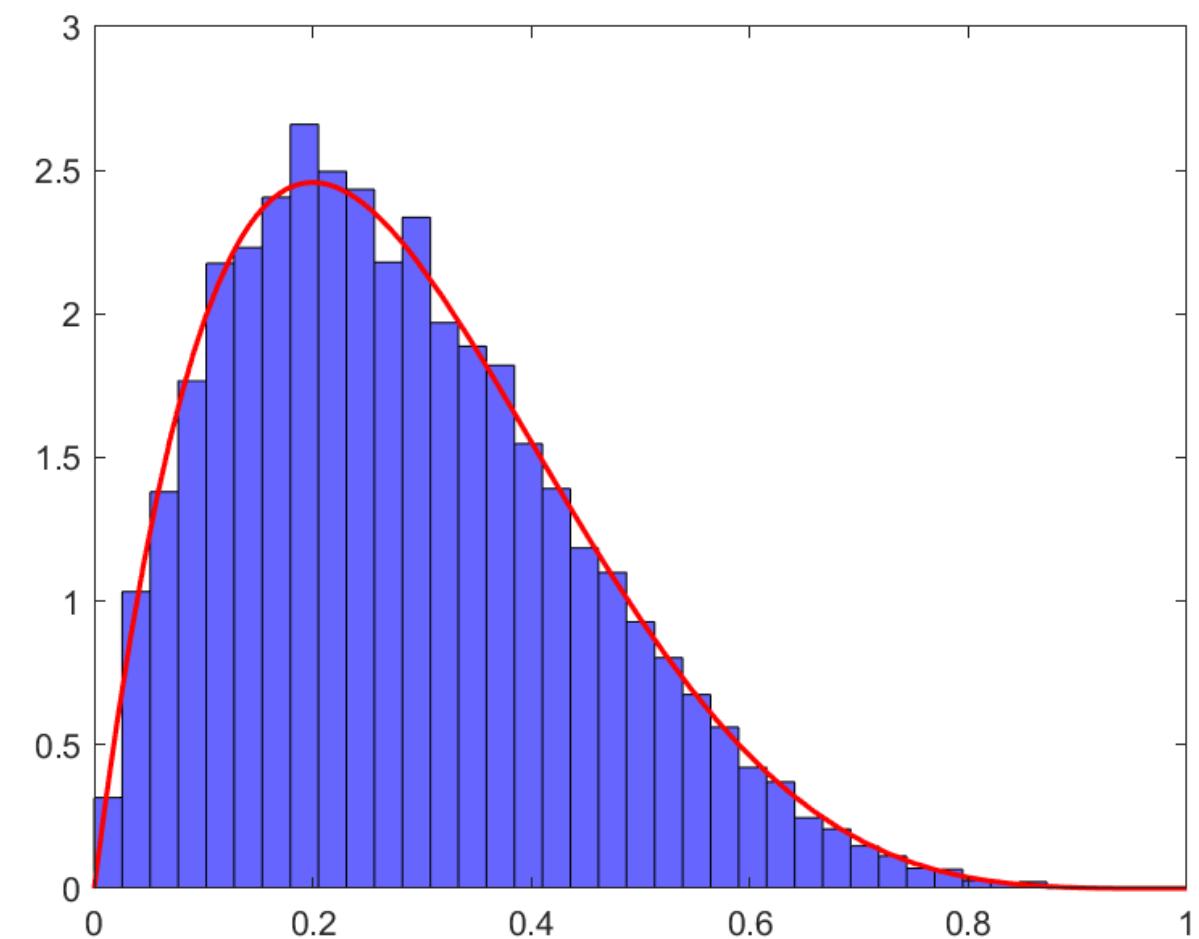
Can also find and plot ECDF.

```

themean=a/(a+b)
thevar=(a*b)/((a+b)^2*(a+b+1))
[mean(x),var(x)]

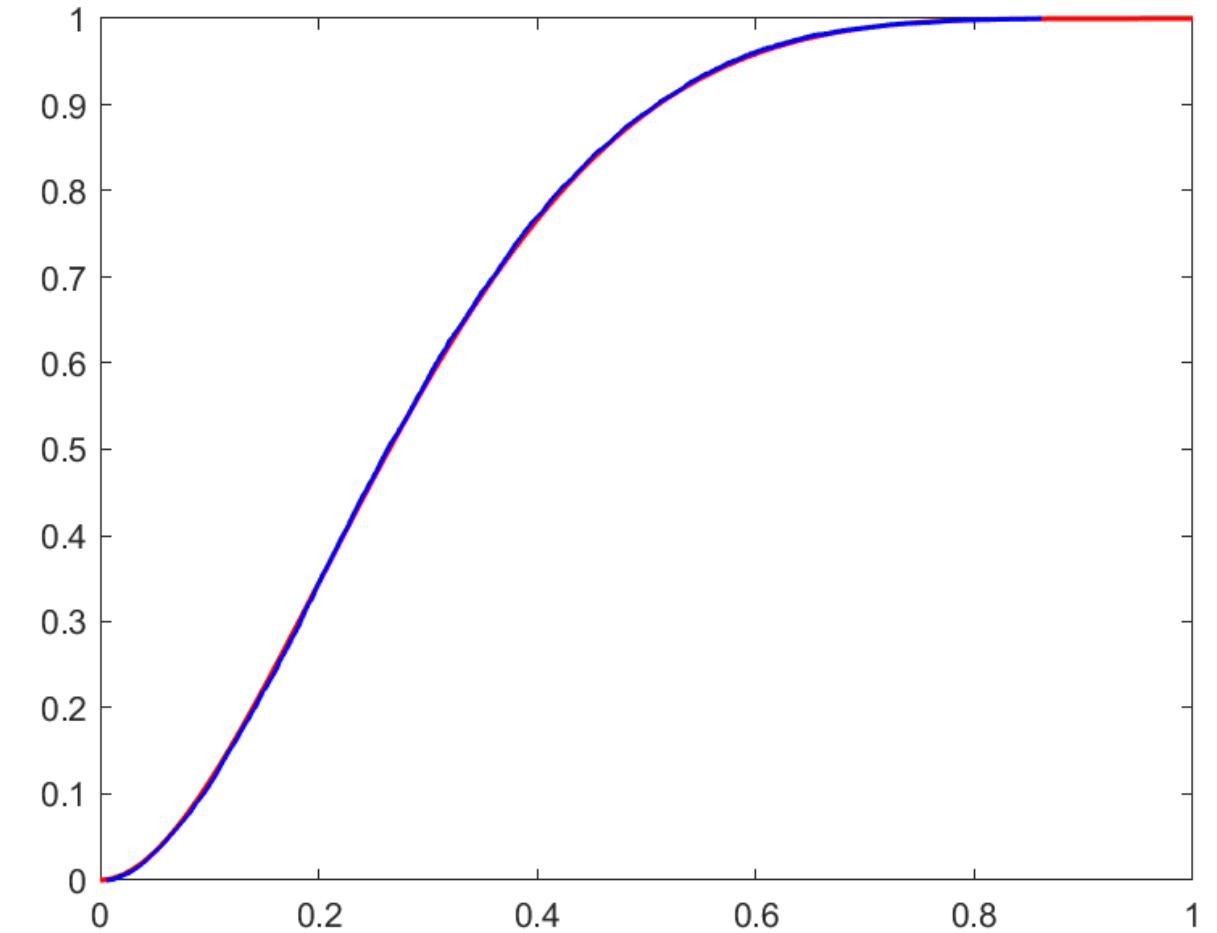
```

$$f(x | \alpha, \beta) = \frac{\Gamma(\alpha + \beta)}{\Gamma(\alpha)\Gamma(\beta)} x^{\alpha-1} (1-x)^{\beta-1}$$



# Uniform and Beta PDF

```
y =cdf('beta',(0:.01:1),a,b);
figure;
plot((0:.01:1),y,'Color','red','LineWidth',1.5)
[F,xx]=ecdf(x);
hold on
stairs(xx,F,'Color','blue','LineWidth',1.5)
```

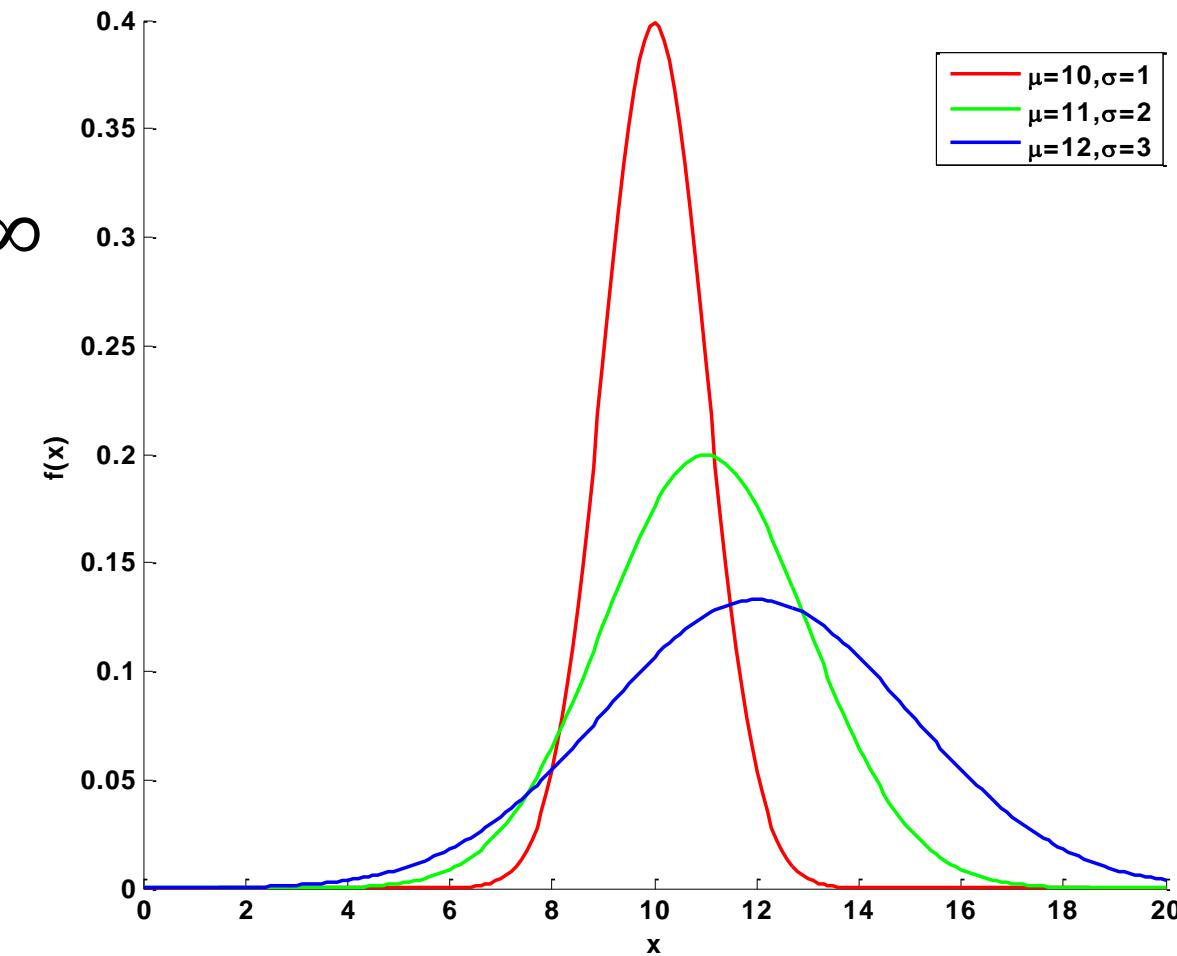


# Normal PDF

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

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

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



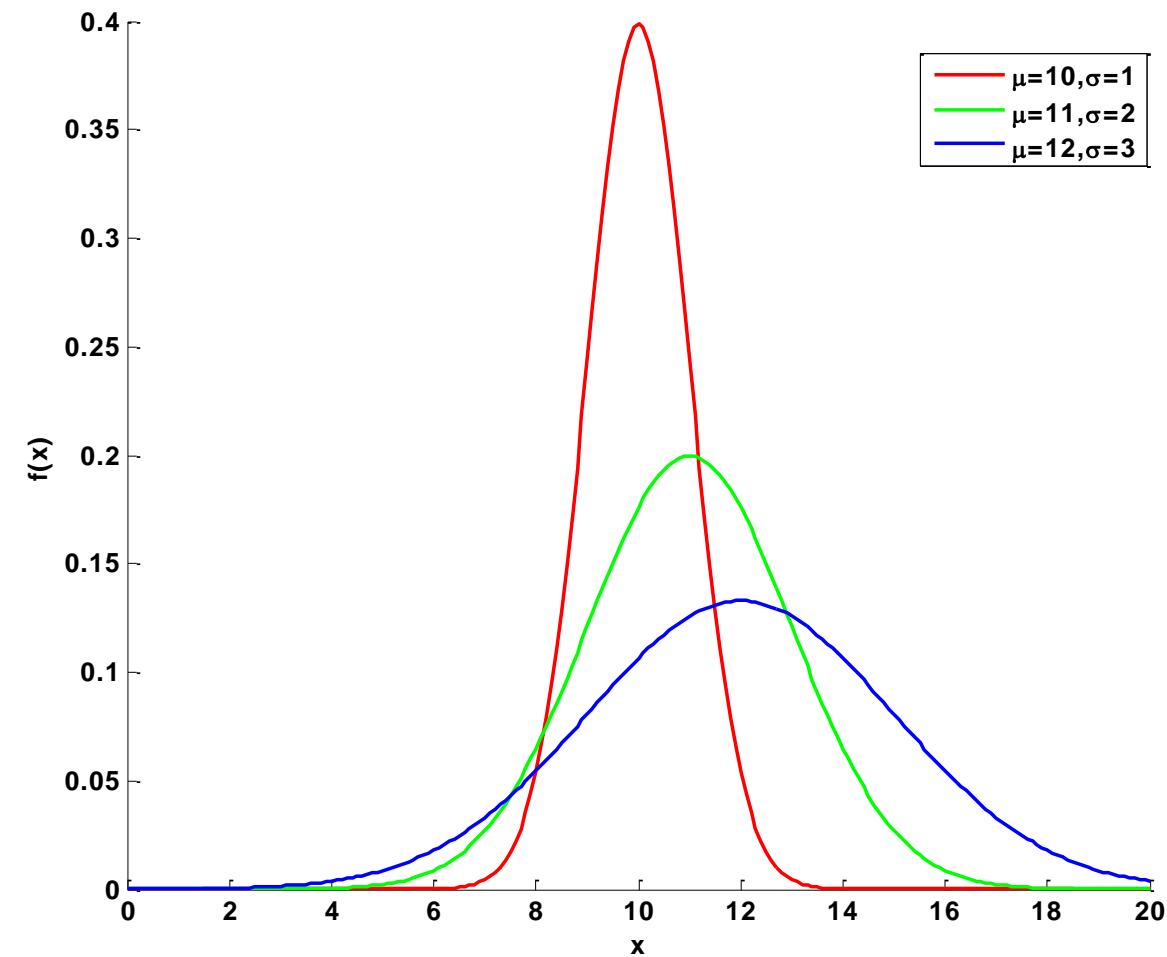
# Normal PDF

```

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])

```

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



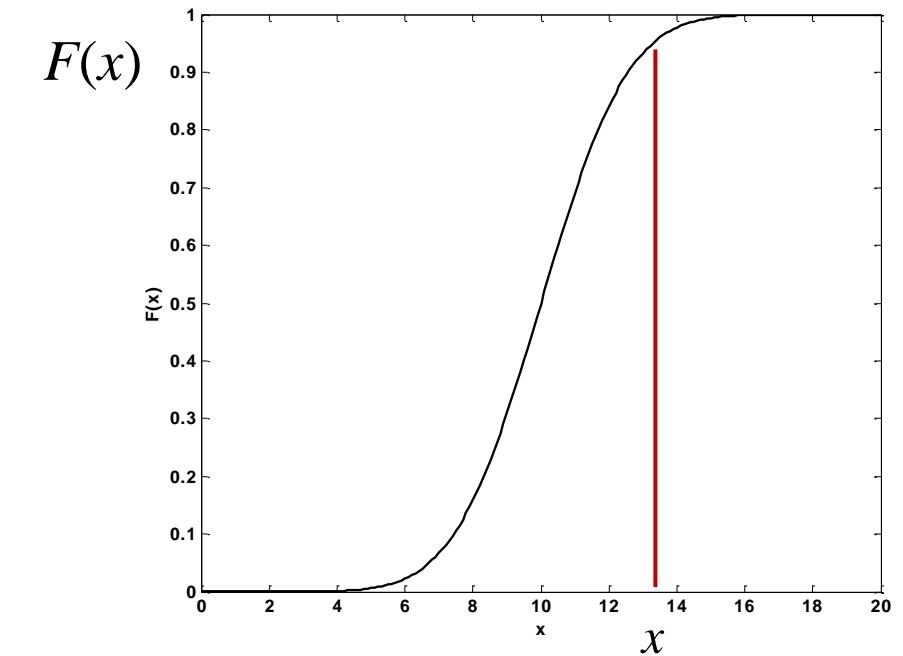
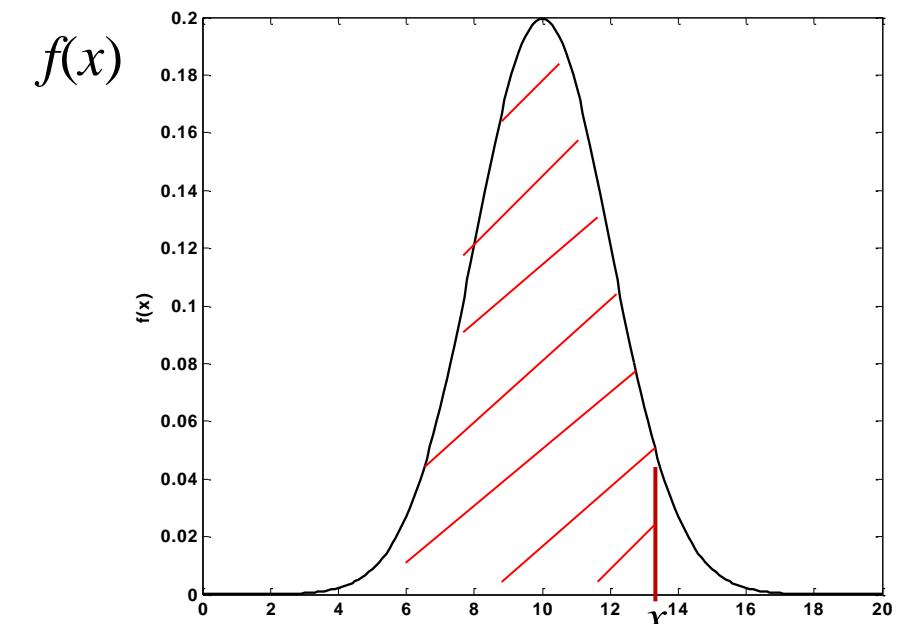
# Normal PDF

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.



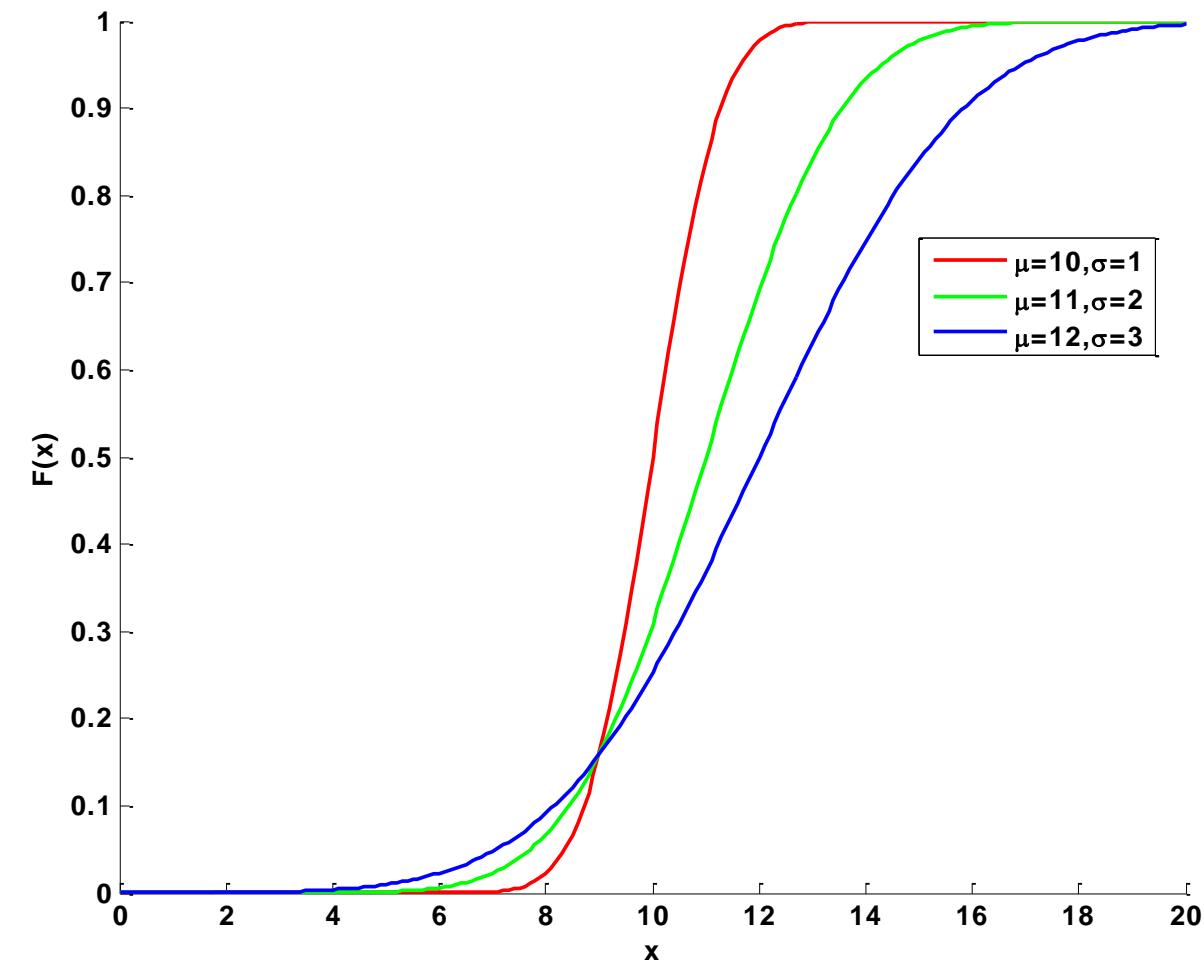
# Normal PDF

```

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])

```

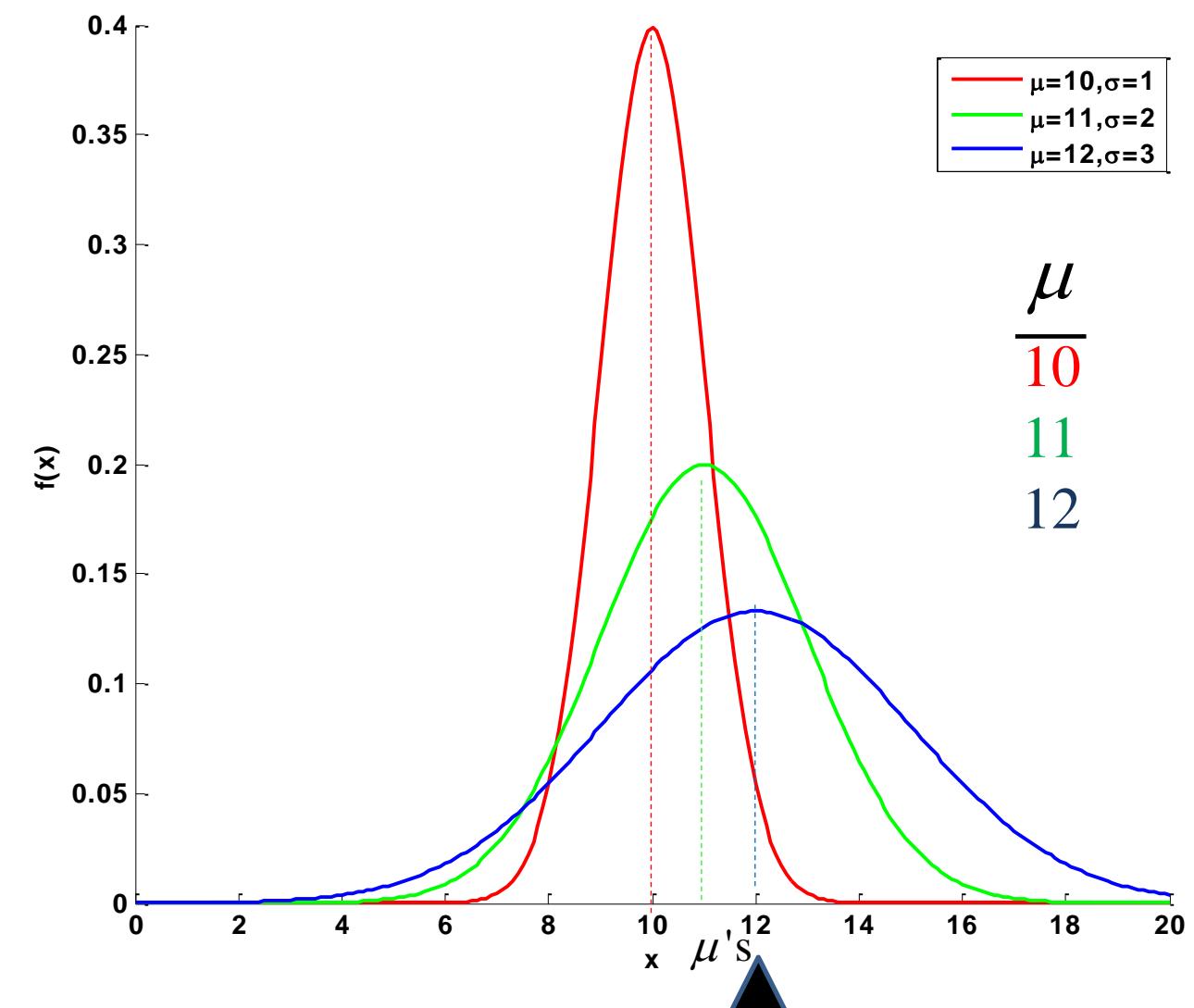
$$f(x | \mu, \sigma^2) = \frac{e^{-\frac{(x-\mu)^2}{2\sigma^2}}}{\sqrt{2\pi\sigma^2}}$$



# Normal PDF

It can be shown that

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



# Normal PDF

It can be shown that

median

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

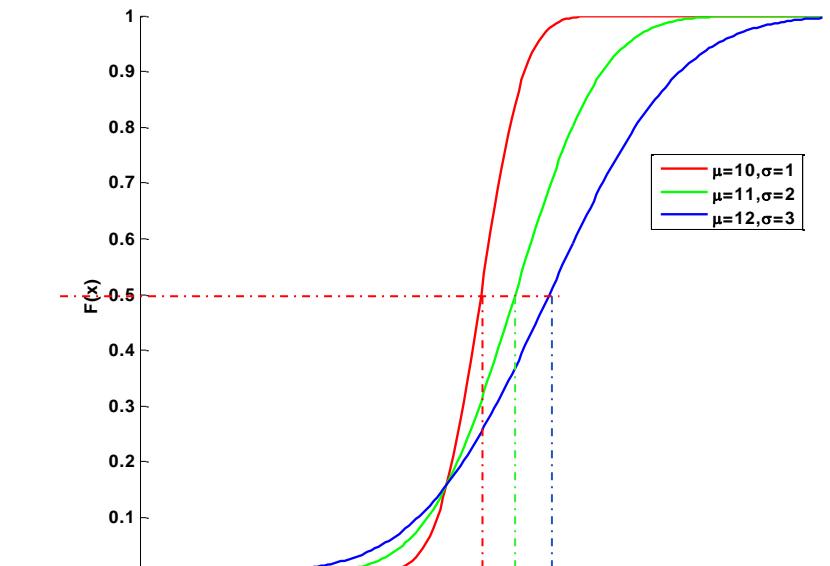
$$\tilde{x} = \mu$$

mode

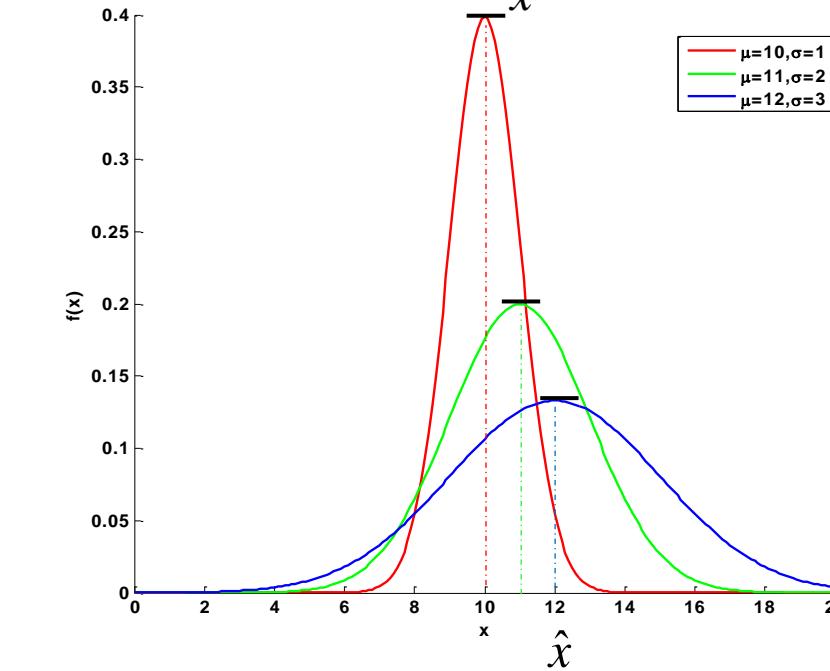
$$\left. \frac{\partial}{\partial x} f(x | \theta) \right|_{\hat{x}} = 0$$

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

$1/2$



$$\begin{matrix} \tilde{x} \\ \hline 10 \\ 11 \\ 12 \end{matrix}$$

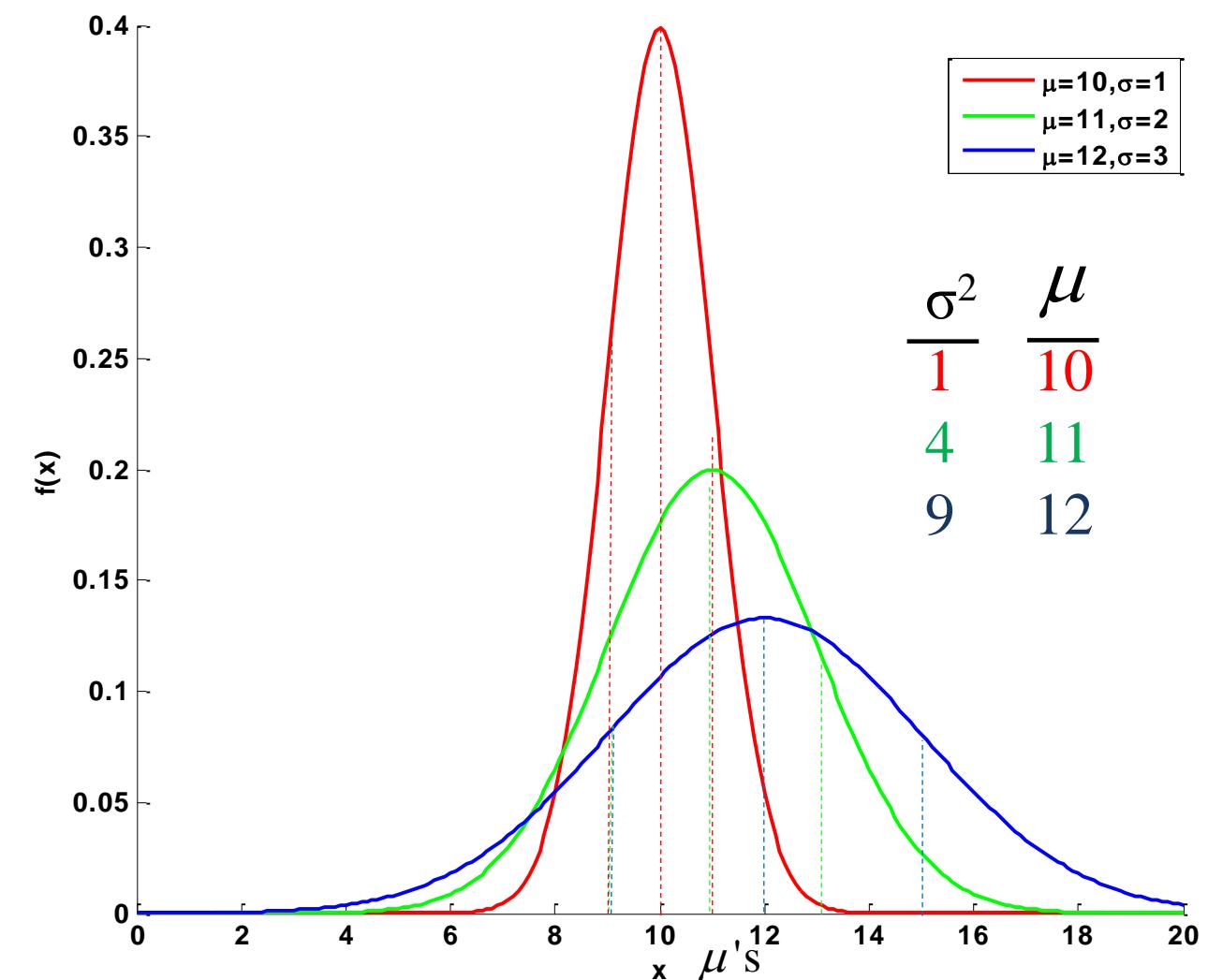


$$\begin{matrix} \hat{x} \\ \hline 10 \\ 11 \\ 12 \end{matrix}$$

# Normal PDF

that

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



# Normal PDF

```

rng('default'), mu=11;,sigma=2;,num=10^4;
x=sigma*randn(num,1)+mu;
figure;
histogram(x,'BinLimits',[2,20],'normalization','pdf','FaceColor','blue')
hold on
fx = @(z) 1/sqrt(2*pi*sigma^2).*exp(-(z-mu).^2/(2*sigma^2));
fplot(fx,[0,22], 'r', 'LineWidth',1.5)
xlim([2,20])

```

True	Simulated
$\mu$	11.0033
$\sigma^2$	3.9321

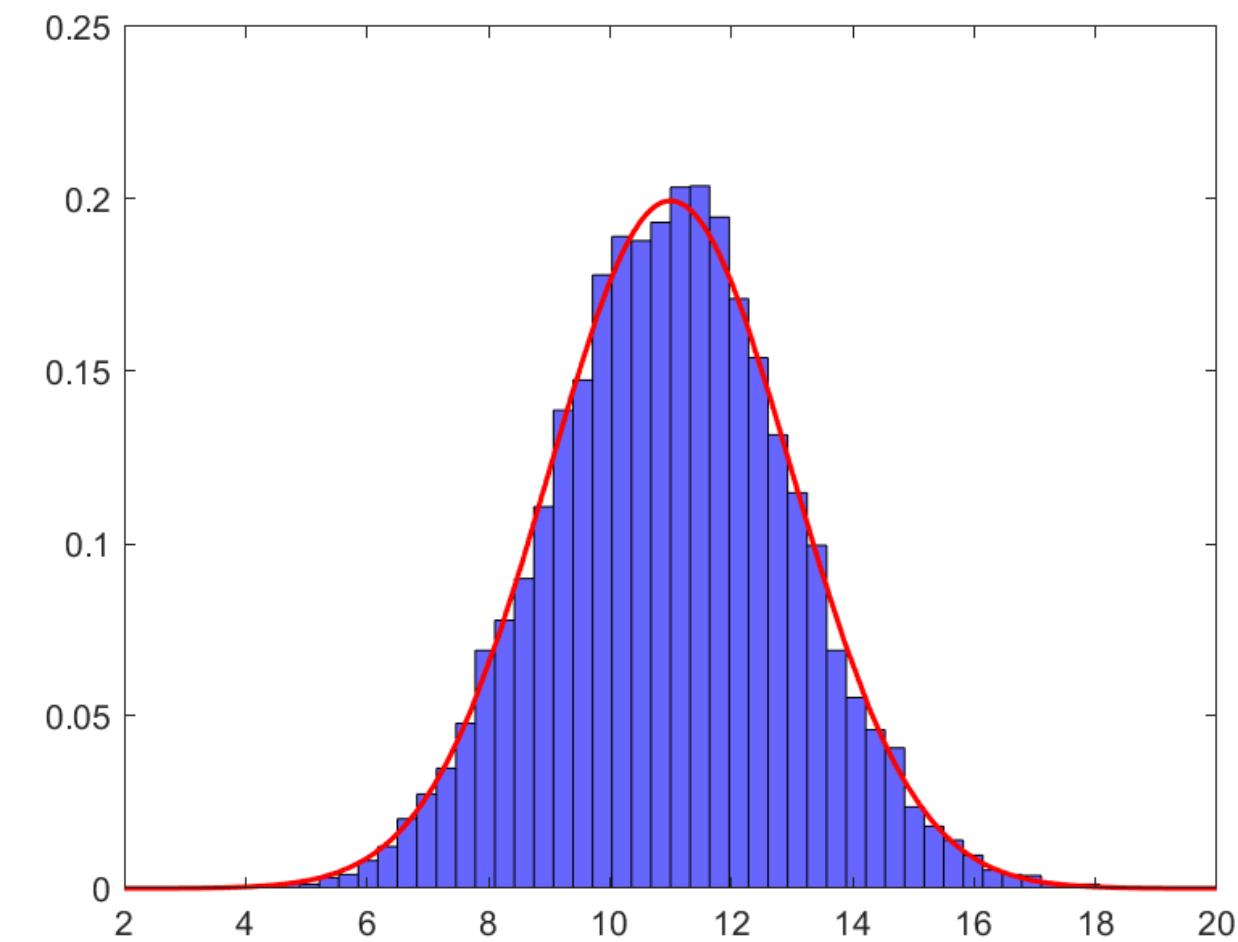
Can also find and plot ECDF.

```

themean=mu
thevar=sigma^2
[mean(x),var(x)]

```

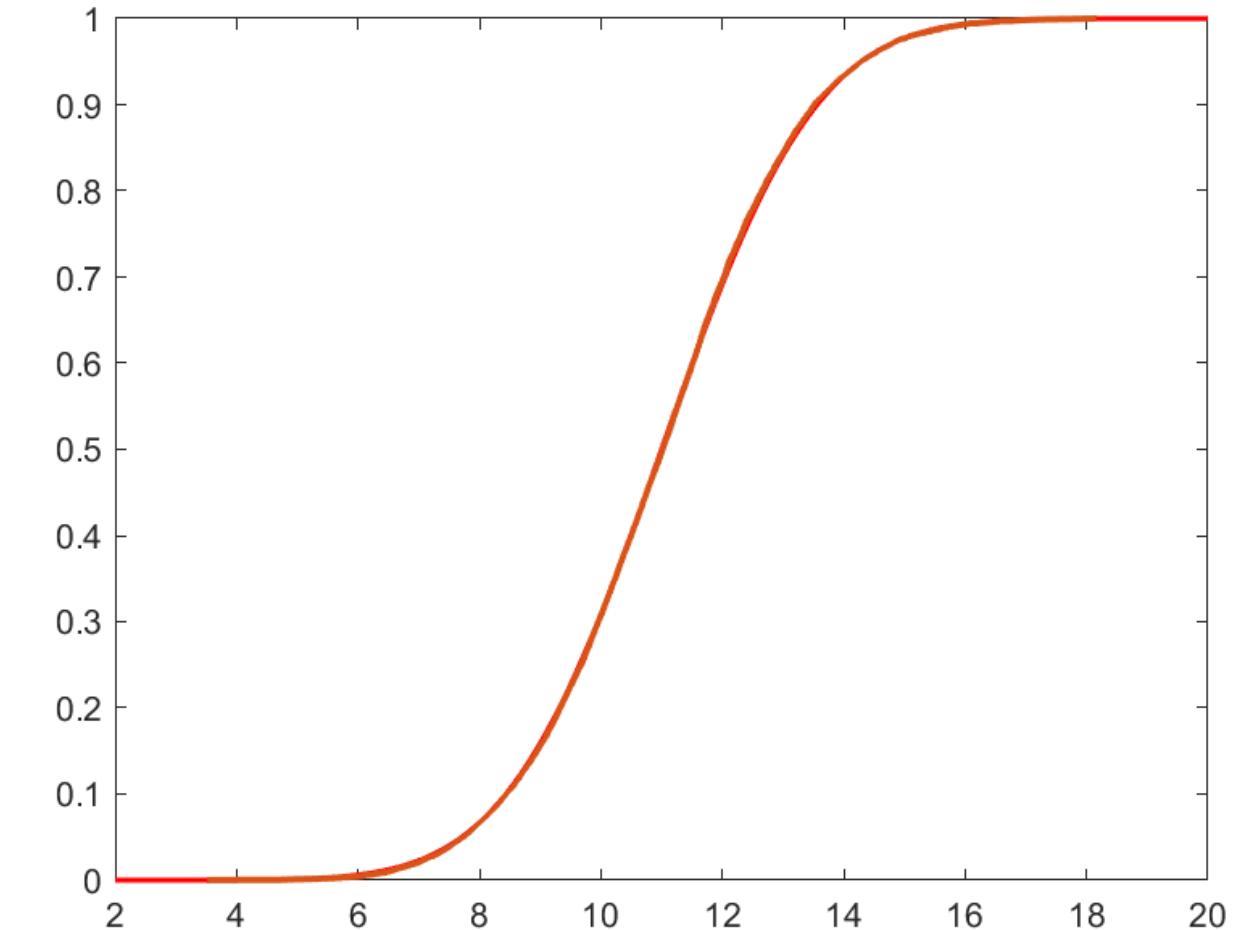
$$f(x | \mu, \sigma^2) = \frac{e^{-\frac{(x-\mu)^2}{2\sigma^2}}}{\sqrt{2\pi\sigma^2}}$$



# Normal PDF

```
y =cdf('Normal',(2:.01:20),mu,sigma);
figure;
plot(2:.01:20),y,'Color','red','LineWidth',1.5)
[F,xx]=ecdf(x);
hold on
stairs(xx,F,'Color','blue','LineWidth',1.5)
```

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



# Inverse Gamma PDF

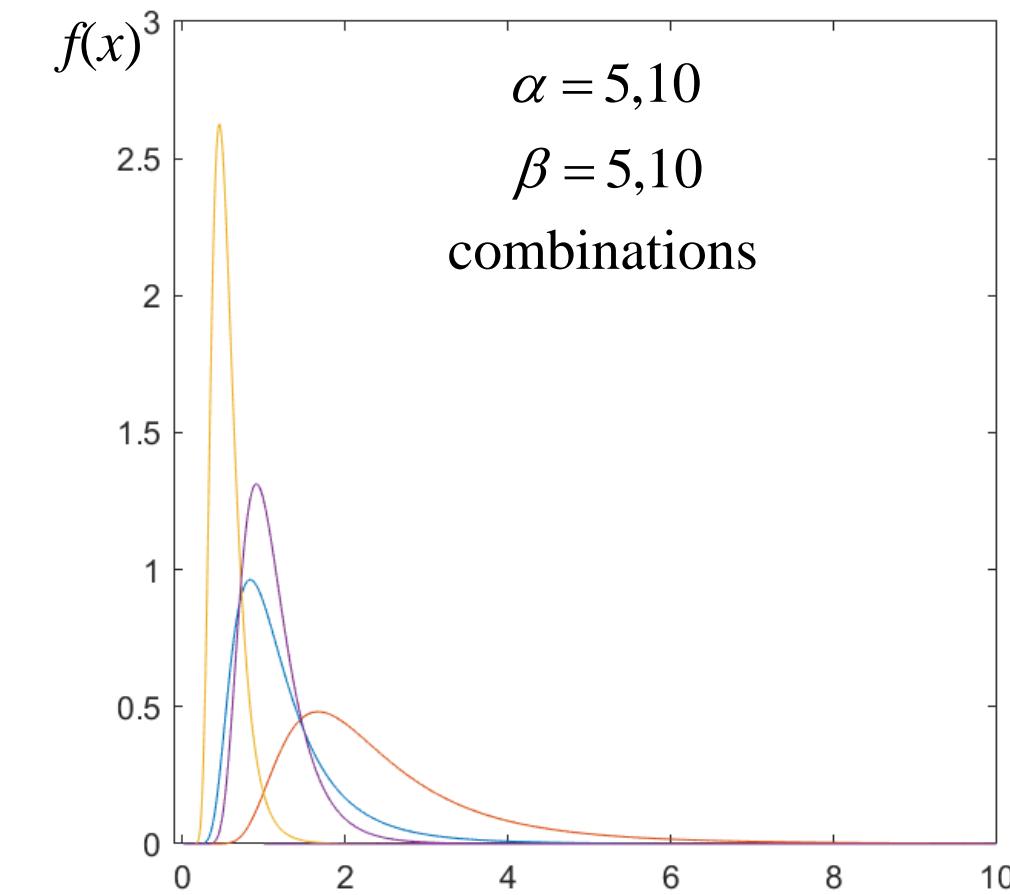
A random variable  $x$  has a continuous Inverse Gamma distribution,  $x \sim \text{Inv}\Gamma(\alpha, \beta)$  if

$$f(x | \alpha, \beta) = \frac{\beta^\alpha}{\Gamma(\alpha)} x^{-\alpha-1} e^{-\beta/x},$$

where,  $\alpha, \beta > 0$ ,  $0 < x < \infty$ .

And can take on many shapes.

$$\Gamma(a) = (a - 1)! \text{ for integer } a$$



# Inverse Gamma PDF

There is no closed form analytic formula for the CDF  $F(x | \alpha, \beta)$ .

It can be shown that the mean is

$$\begin{aligned}\mu &= \int_x xf(x | \theta)dx \\ &= \int_{x=0}^{\infty} x \frac{\beta^\alpha}{\Gamma(\alpha)} x^{-\alpha-1} e^{-\beta/x} dx \\ &= \frac{\beta}{\alpha - 1}\end{aligned}$$

# Inverse Gamma PDF

It can be shown that the variance is

$$\begin{aligned}\sigma^2 &= \int_x (x - \mu)^2 f(x | \theta) dx \\ &= \int_{x=0}^{\infty} (x - \mu)^2 \frac{\beta^\alpha}{\Gamma(\alpha)} x^{-\alpha-1} e^{-\beta/x} dx \\ &= \frac{\beta^2}{(\alpha - 1)^2 (\alpha - 2)}\end{aligned}$$

# Inverse Gamma PDF

```

rng('default'), a=5; b=10; c=1/b; num=10^4;
x=1./gamrnd(a,c,[num,1]);
figure;
histogram(x,'BinLimits',[0,10], 'normalization','pdf','FaceColor','blue')
hold on
fx = @(z) b.^a/factorial(a-1).*z.^(-a-1).*exp(-b./z);
fplot(fx,[0,10], 'r', 'LineWidth',1.5)
xlim([0,10])

```

True	Simulated
$\mu$	2.5000
$\sigma^2$	2.0833

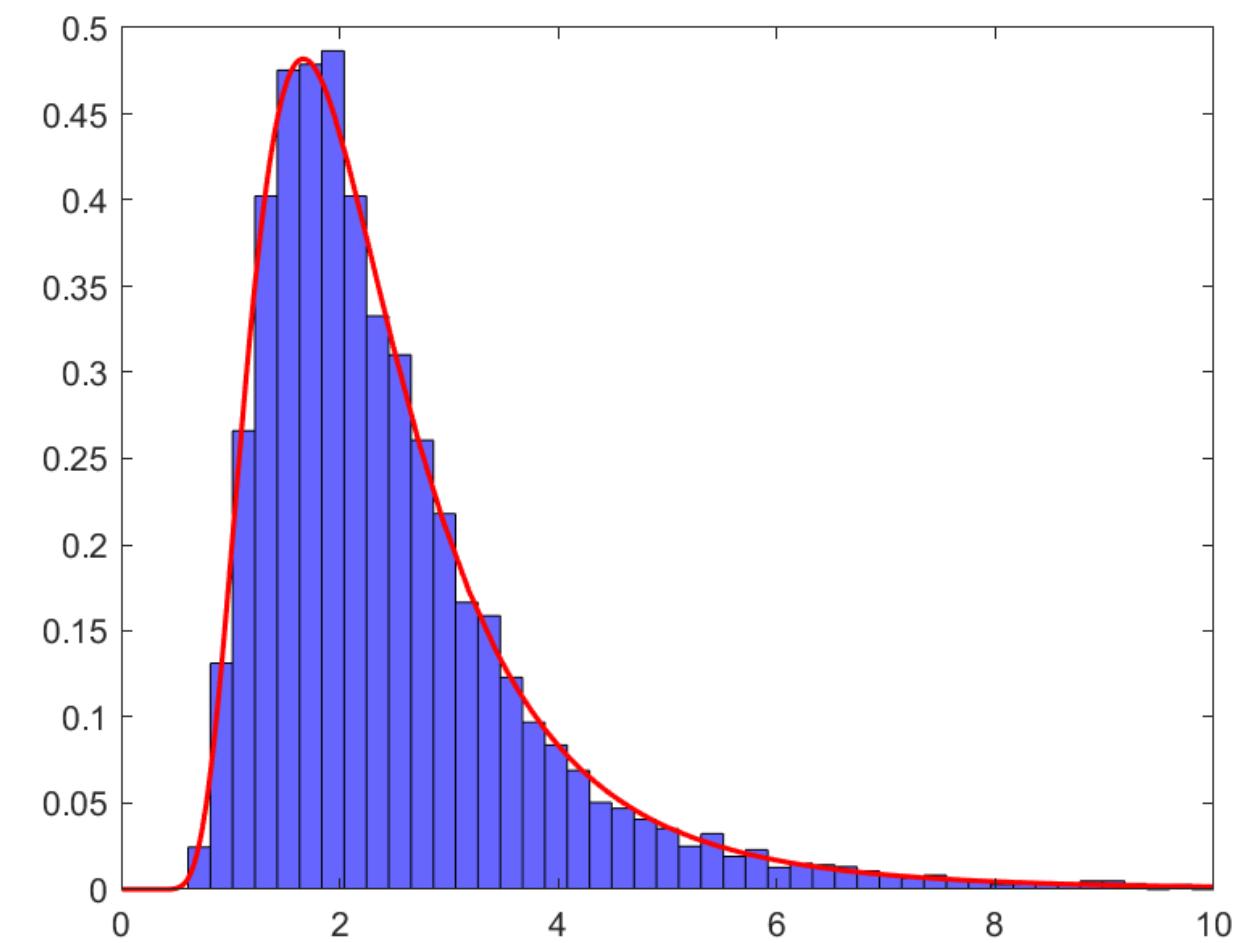
Can also find and plot ECDF.

```

themean=b/(a-1)
thevar=(b^2)/((a-1)^2*(a-2))
[mean(x),var(x)]

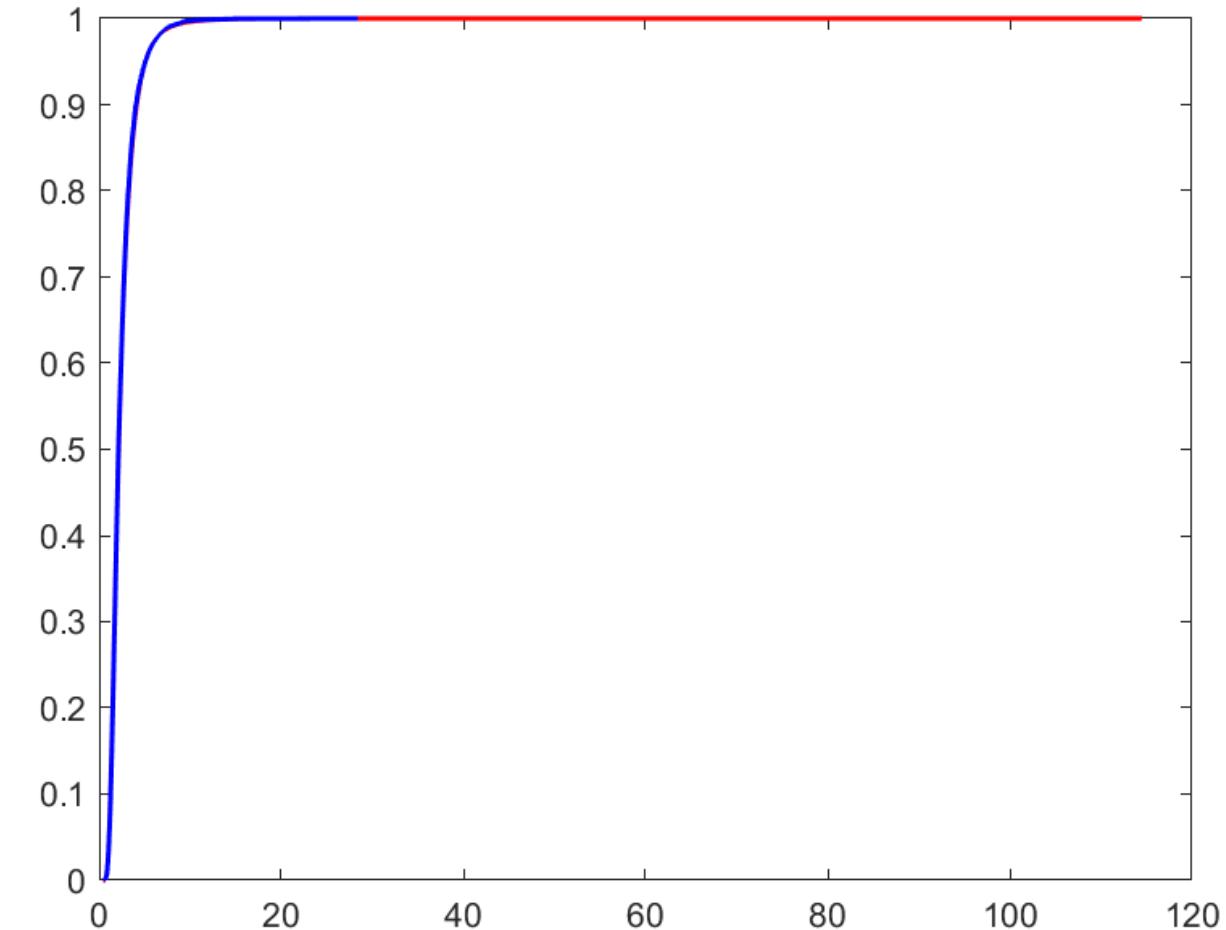
```

$$f(x | \alpha, \beta) = \frac{\beta^\alpha}{\Gamma(\alpha)} x^{-\alpha-1} e^{-\beta/x}$$



# Inverse Gamma PDF

```
y=1./gamrnd(a,c,[5*10^6,1]);
[Fy,yy]=ecdf(y);
figure;
stairs(yy,Fy,'Color','red','LineWidth',1.5)
[F,xx]=ecdf(x);
hold on
stairs(xx,F,'Color','blue','LineWidth',1.5)
```



# Questions?

# Homework 3

$$f(x | \alpha, \beta) = \frac{\Gamma(\alpha + \beta)}{\Gamma(\alpha)\Gamma(\beta)} x^{\alpha-1} (1-x)^{\beta-1}$$

1. Use Matlab to generate  $n=1000$  observations from the Beta PDF with  $\alpha=3$  and  $\beta=7$ . Change  $\alpha, \beta$  to observe shapes.
  - a. Calculate the sample mean.
  - b. Calculate the sample variance.
  - c. Calculate the median.
  - d. Make a histogram.
  - e. Make an empirical CDF and calculate the 95<sup>th</sup> percentile

# Homework 3

2. Use Matlab to generate  $n=1000$  observations from the normal PDF with  $\mu=100$  and  $\sigma^2=5$ .
- Calculate the sample mean.
  - Calculate the sample variance.
  - Calculate the median.
  - Make a histogram.
  - Make an empirical CDF and calculate the 95<sup>th</sup> percentile.

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

# Homework 3

$$f(x|\alpha, \beta) = \frac{\beta^\alpha}{\Gamma(\alpha)} x^{\alpha-1} e^{-\beta x}$$

3. Assume  $y$  has a gamma distribution with parameters with  $\alpha=3$ ,  $\beta=7$ , and  $c=1/\beta$ . Change  $\alpha, \beta$  to observe shapes.
- a. Generate  $n=1000$  random  $y$  observations.  
Calculate the sample mean and variance.  
Make a histogram. Compare to the Gamma PDF.
- b\*. Using pencil and paper transform to  $x$  as  $x=1/y$ .
- c. Take  $x=1/y$  for each  $y$  random observation in part a.
- d. Repeat a. but now for  $y$ 's and compare to inverse gamma PDF.

\*For students in MSSC 5790