



## **Quantitative Risk Analysis Using Probability Distributions**

**COORDINATOR:**

**Dr Lihai Zhang**

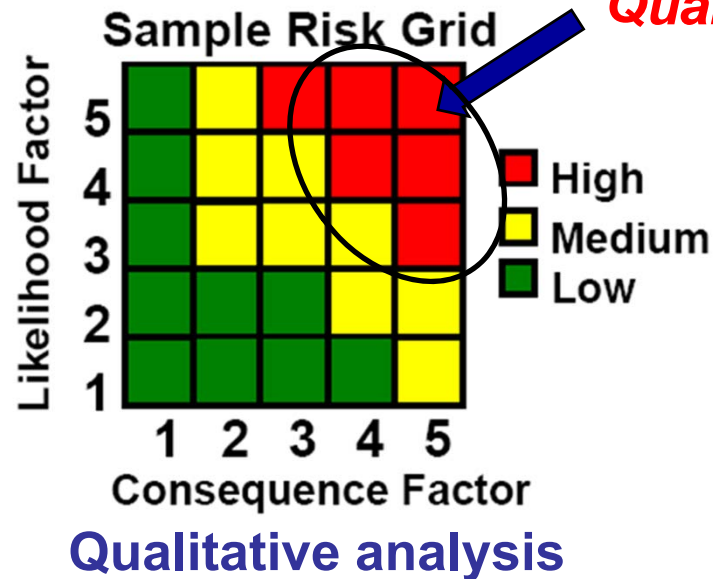
**Department of Infrastructure Engineering**

**[lihzhang@unimelb.edu.au](mailto:lihzhang@unimelb.edu.au)**





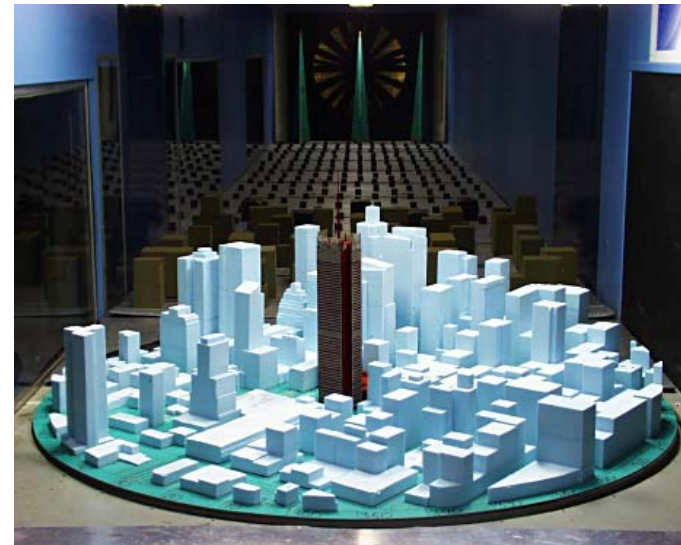
# Why Quantitative Risk Analysis ?



*Quantitative risk analysis is required !*



**Hurricane Risks**



**Wind Tunnel Test**



# Risks in society

- Selected risks in society (Robert E. Melchers, 2002)

Activity	Approx. death rate ( $10^{-9}$ deaths/h exposure)	Typical exposure (h/year)	Typical risk of death ( $10^{-6}$ /year)
Construction works	70 ~ 200	2200	150 ~ 440
Coal mining (UK)	210	1500	300
Building fires	1 ~ 3	8000	8 ~ 24
Structural failures	0.02	6000	0.1
Smoking	2500	400	1000
Air travel	1200	20	24
Car travel	700	300	200
Alpine climbing	30,000 ~ 40,000	50	1500 ~ 2000

***How an insurance company determines your premiums?***



# Risks in society

- Typical collapse failure rates for structures  
(Robert E. Melchers, 2002)

Structural type	Data cover	Average life (years)	Probability of failure
Apartments	Demark	30	0.000003%
Mixed housing	Canada	50	0.1%
Large suspension bridge	World	40	0.3%



# OHS Risks

For a large construction project, the contractor estimates that the average rate of on-the-job accidents is **three times per year**. From past experience, the contractor also estimates that the cost incurred for each accident may be modeled as a lognormal random variable with **a median of \$6,000** and **COV of 20%**. The cost of each accident can be assumed to be statistically independent.



- (1) What is the probability that there will be no accident in the first month of construction?
- (2) What is the probability that an accident will incur a loss exceeding \$4,000?





# Risk Analysis using Distributions

- **Distributions** where **random variable** takes on a number of specific values with certain probabilities

Discrete

Continuous

distribution

continuous  
curve  
symmetrical  
probability  
Normal  
variable  
bell-shaped  
around  
random  
forms  
mean



- **Discrete**

- **Example: Fair Coin**

A fair coin is flipped,  $X$  to be the random variable, “head” to be 1, and “tail” to be 0. What is the probability that the coin is a head

$P(X=1) = 50\%$ ,  $P(X=0) = 50\%$ ,

- **Continuous**

- **Example:**

The number of floods in a given year at a particular location.

The strength of a concrete cylinder

The distance between cracks in a roadway.

The inches of rainfall during a storm.



- **Continuous Probability Distributions**

- Normal or Gaussian Distribution
- Lognormal Distribution

- **Discrete Probability Distributions**

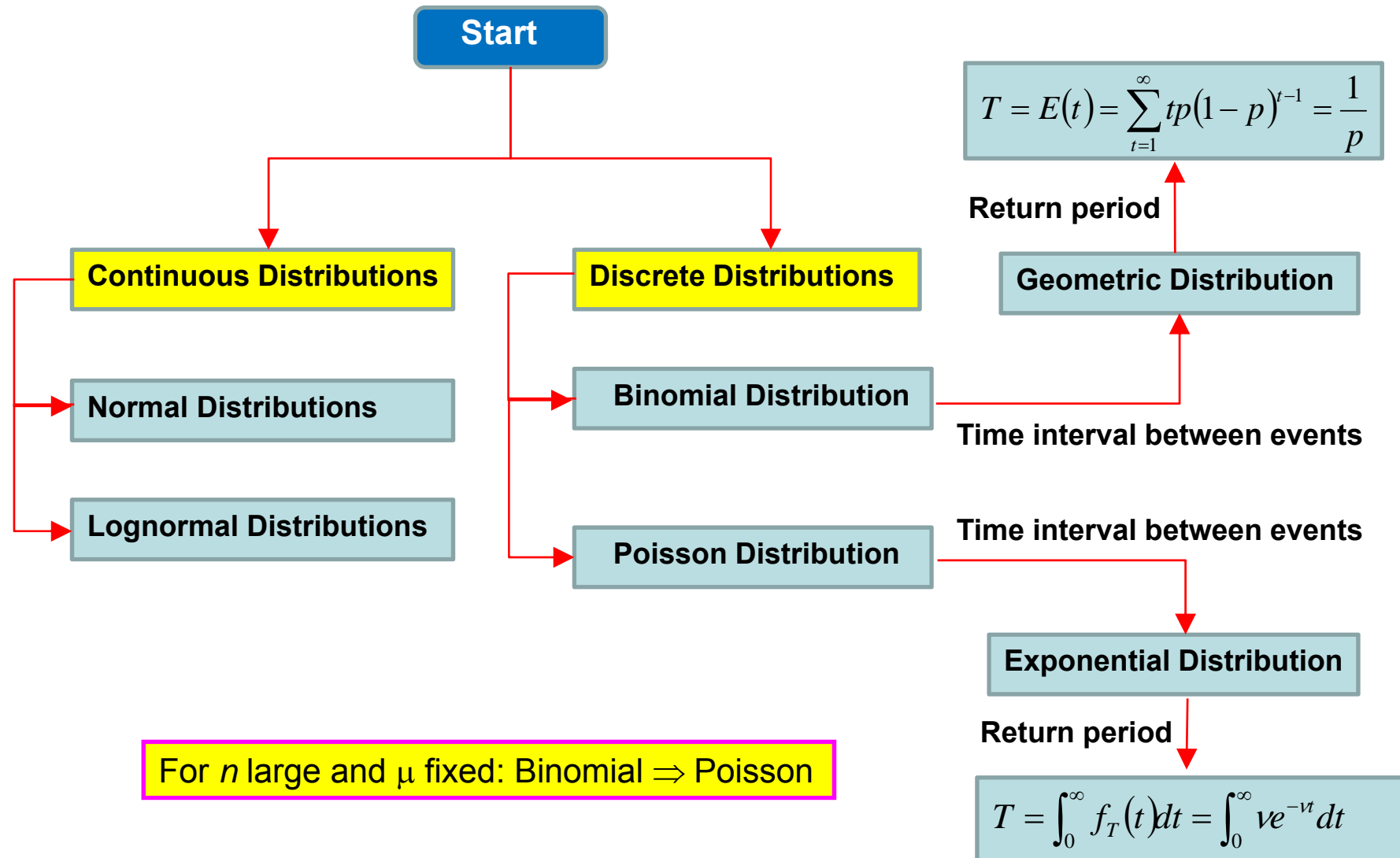
- Binomial Distribution
- Poisson Distribution







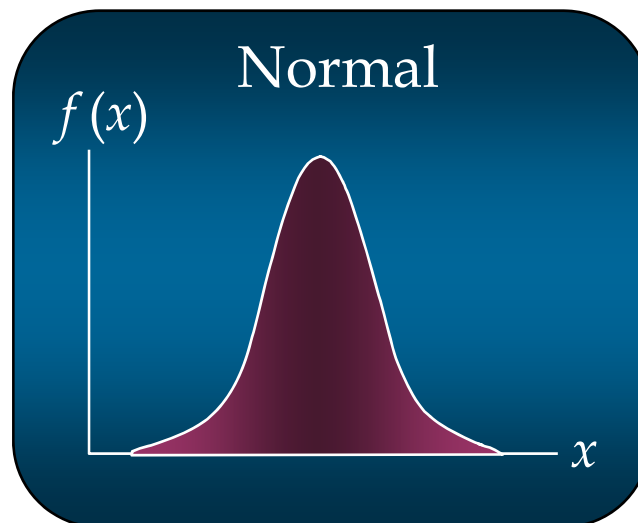
# Risk Analysis using Distributions





- **Normal or Gaussian Distribution**

- The most important distribution for describing a continuous random variable.
- The normal (or Gaussian) distribution is a continuous probability distribution that has a bell-shaped probability density function.





# Quality Risks

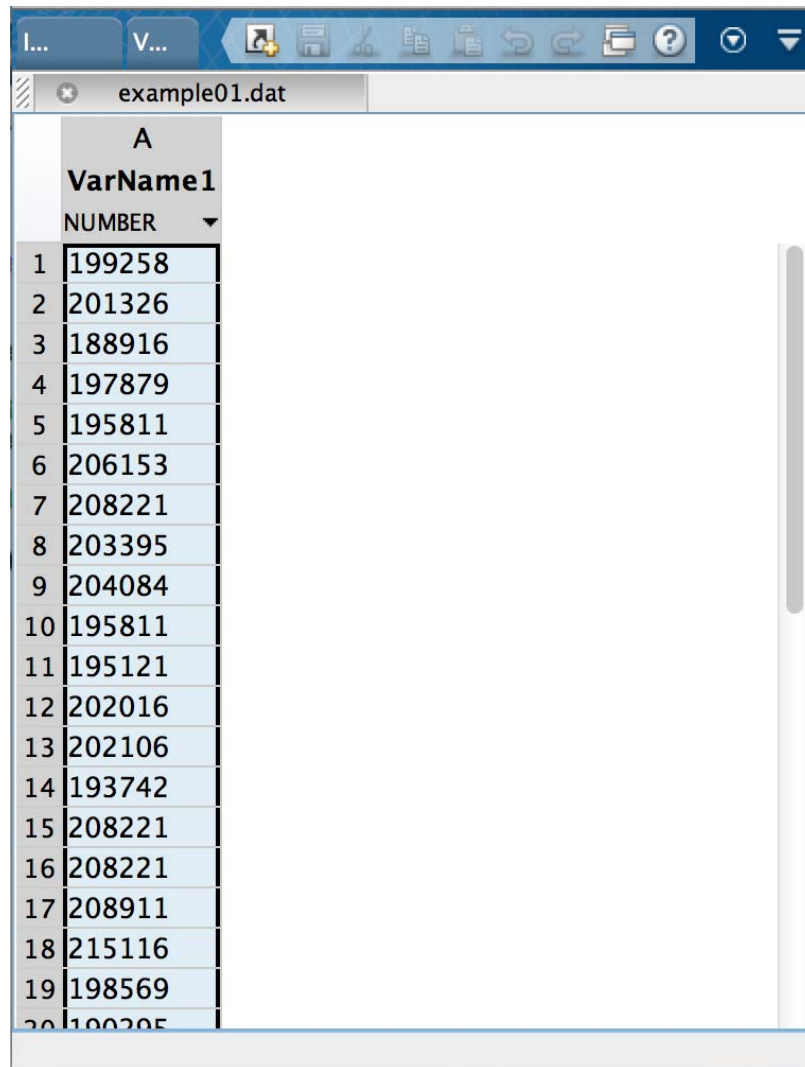
Consider the values of the Young's modulus given in below.

Test no.	E (MPa)	Test no.	E (MPa)	Test no.	E (MPa)
1	199,258	12	202,016	23	220,632
2	201,326	13	202,016	24	230,284
3	188,916	14	193,742	25	210,979
4	197,879	15	208,221	26	225,458
5	195,811	16	208,221	27	215,805
6	206,153	17	208,911	28	210,290
7	208,221	18	215,116	29	215,805
8	203,395	19	198,569	30	199,947
9	204,084	20	190,295	31	202,705
10	195,811	21	204,084	32	195,121
11	195,121	22	178,574	33	210,290



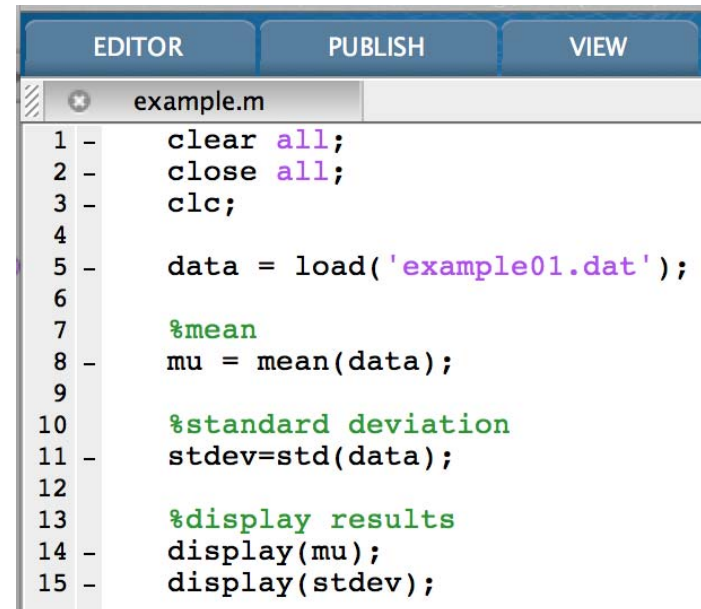


# Quality Risks



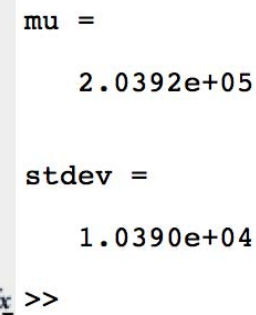
The Data Editor window displays a table with 20 rows and 2 columns. The first column is labeled 'A' and the second column is labeled 'VarName1'. The data in the 'VarName1' column is as follows:

	A	VarName1
1	199258	NUMBER
2	201326	
3	188916	
4	197879	
5	195811	
6	206153	
7	208221	
8	203395	
9	204084	
10	195811	
11	195121	
12	202016	
13	202106	
14	193742	
15	208221	
16	208221	
17	208911	
18	215116	
19	198569	
20	199258	



```
1 - clear all;  
2 - close all;  
3 - clc;  
4  
5 - data = load('example01.dat');  
6  
7 %mean  
8 - mu = mean(data);  
9  
10 %standard deviation  
11 - stdev=std(data);  
12  
13 %display results  
14 - display(mu);  
15 - display(stdev);
```

## Command Window



```
mu =  
  
    2.0392e+05  
  
stdev =  
  
    1.0390e+04  
fx >>
```



# Normal Probability Distributions

- The normal probability density function (PDF) is

$$f(x) = \frac{1}{\sigma\sqrt{2\pi}} \cdot e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2}$$

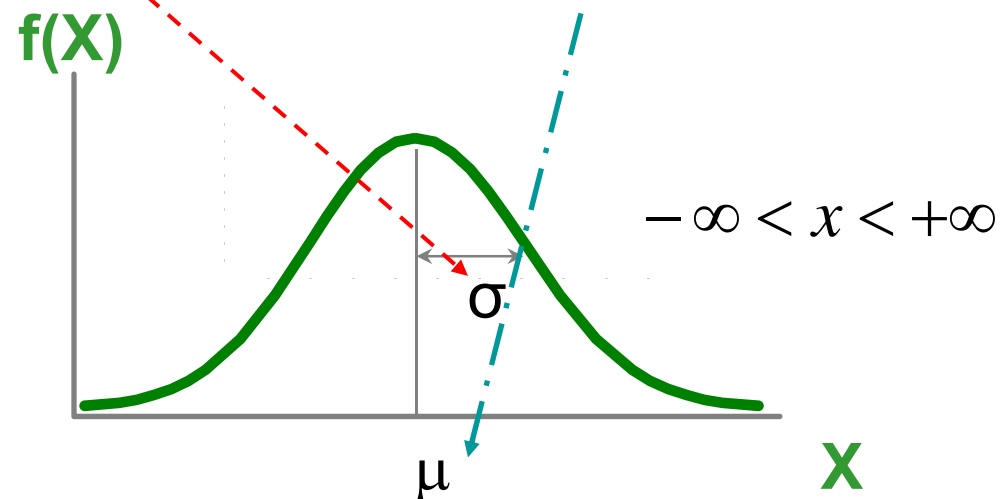
where

$\mu$ : mean

$\sigma$ : standard deviation

$\pi = 3.14159$

$e = 2.71828$



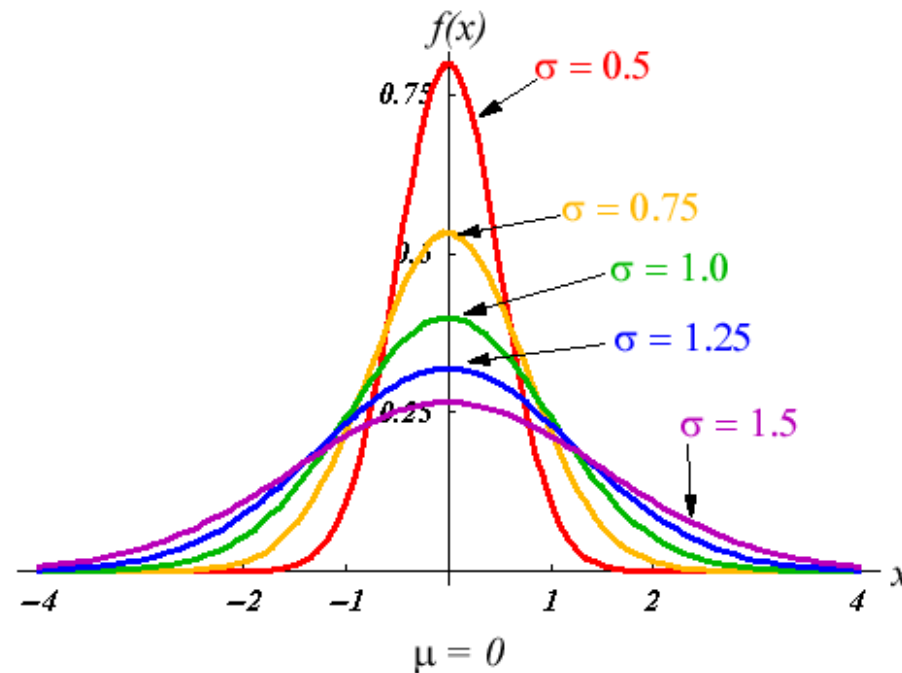
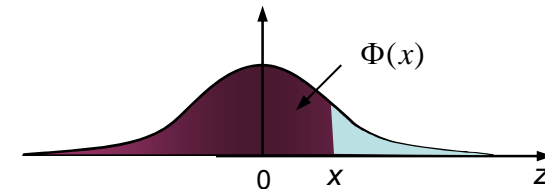


# Standard Normal Probability Distribution

- The probability density function (PDF) is

$$f_s(z) = \frac{1}{\sqrt{2\pi}} \cdot e^{-\frac{1}{2}z^2}$$

$$Z = \frac{x - \mu}{\sigma}$$



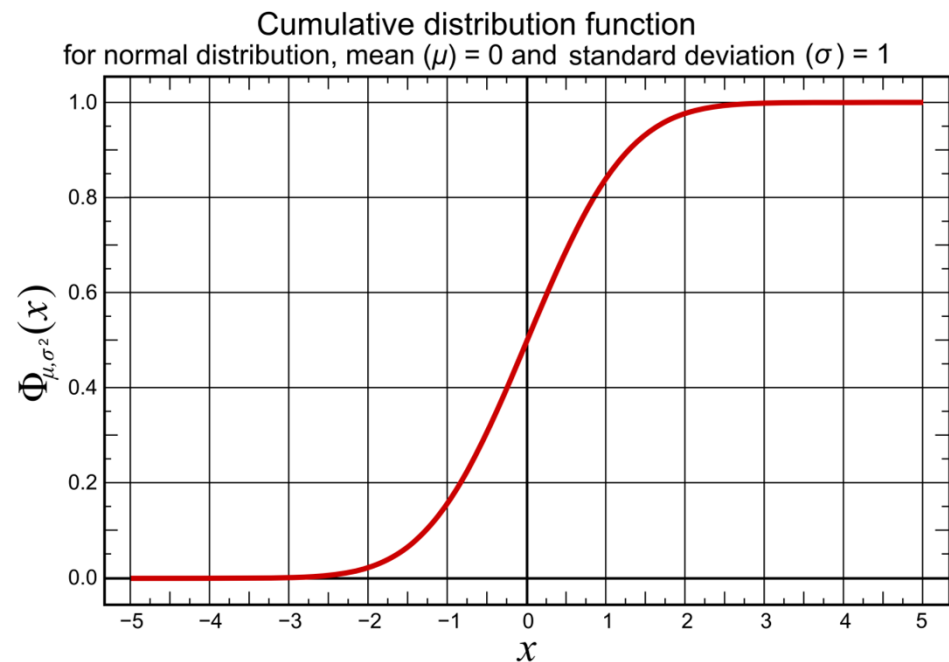
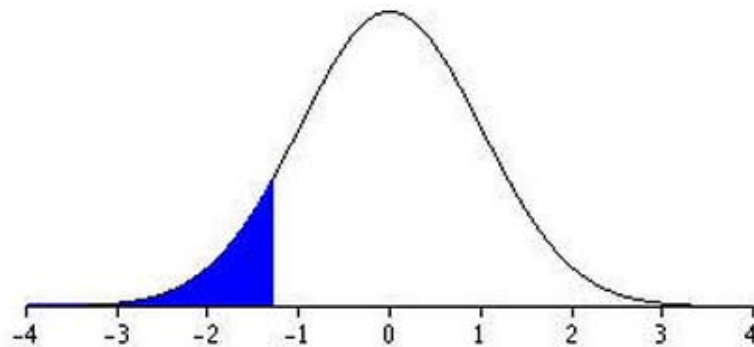


# Standard Normal Probability Distribution

- The cumulative distribution function (CDF) is

$$\Phi(z) = F_z(z) = \int_{-\infty}^z \frac{1}{\sqrt{2\pi}} \cdot e^{-\frac{1}{2}z^2} dz$$

$$P(a < T < b) = \Phi\left(\frac{b - \mu}{\sigma}\right) - \Phi\left(\frac{a - \mu}{\sigma}\right)$$

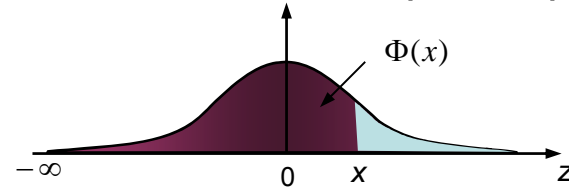




# Continuous Distributions

## Table of the cumulative distribution function (CFD)

$$\Phi(z) = \int_{-\infty}^z \frac{1}{\sqrt{2\pi}} \cdot e^{-\frac{1}{2}z^2} dz$$

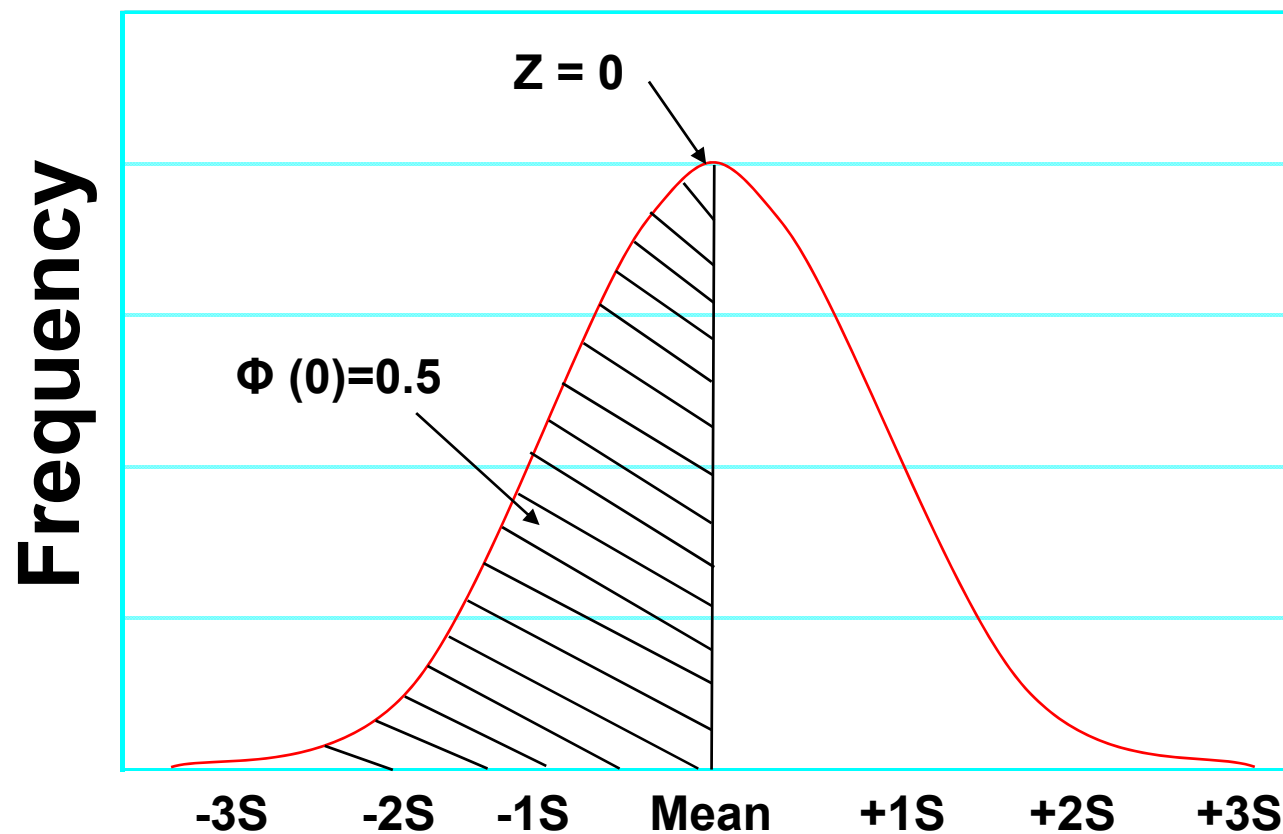


x	$\Phi(x)$
0.00	0.50000
0.01	0.50399
0.02	0.50798
0.03	0.51197
0.04	0.51595
0.05	0.51994
0.06	0.52392
0.07	0.52790
0.08	0.53188
0.09	0.53586

x	$\Phi(x)$
1.00	0.84134
1.01	0.84375
1.02	0.84614
1.03	0.84849
1.04	0.85083
1.05	0.85314
1.06	0.85543
1.07	0.85769
1.08	0.85993
1.09	0.86214



# Standard Normal Probability Distribution



$x$	$\Phi(x)$
0.00	0.50000
0.01	0.50399
0.02	0.50798
0.03	0.51197
0.04	0.51595
0.05	0.51994
0.06	0.52392
0.07	0.52790
0.08	0.53188
0.09	0.53586



## Example 1

Assume that the randomness in Young's modulus of steel  $E$  can be described by **normal random variable**. Calculate the probability of  $E$  having a value between 193,053 MPa and 203,395 MPa.







Consider the values of the Young's modulus given in below.

Test no.	E (MPa)	Test no.	E (MPa)	Test no.	E (MPa)	Test no.	E (MPa)
1	199,258	12	202,016	23	220,632	34	214,426
2	201,326	13	202,016	24	230,284	35	202,016
3	188,916	14	193,742	25	210,979	36	188,916
4	197,879	15	208,221	26	225,458	37	202,016
5	195,811	16	208,221	27	215,805	38	202,016
6	206,153	17	208,911	28	210,290	39	215,805
7	208,221	18	215,116	29	215,805	40	189,605
8	203,395	19	198,569	30	199,947	41	202,705
9	204,084	20	190,295	31	202,705		
10	195,811	21	204,084	32	195,121		
11	195,121	22	178,574	33	210,290		



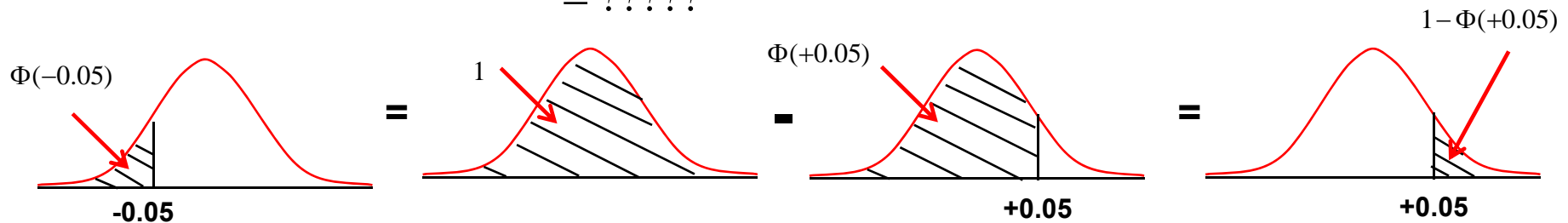
## ***Solution***

## Solution

$$\text{Mean} = E(X) = \mu = \frac{1}{n} \sum_{i=1}^n x_i = 203,919 \text{ MPa}$$

$$\text{Standard deviation} = \sigma = \sqrt{\text{Var}(X)} = \sqrt{\frac{1}{n-1} \sum_{i=1}^n (x_i - \mu)^2} = 10,390 \text{ MPa}$$

$$\begin{aligned} P(193,053 < E \leq 203,395) &= \Phi\left(\frac{203,395 - 203,919}{10,390}\right) - \Phi\left(\frac{193,053 - 203,919}{10,390}\right) \\ &= \Phi(-0.05) - \Phi(-1.05) \\ &= [1 - \Phi(0.05)] - [1 - \Phi(1.05)] \\ &= \text{?????} \end{aligned}$$

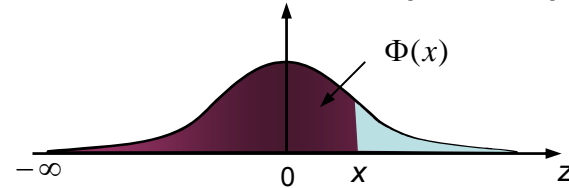




# Quality Risks

## Table of the cumulative distribution function (CFD)

$$\Phi(x) = \frac{1}{\sqrt{2}} \int_{-\infty}^x e^{-(z^2/2)} dz$$



x	$\Phi(x)$
0.00	0.50000
0.01	0.50399
0.02	0.50798
0.03	0.51197
0.04	0.51595
0.05	0.51994
0.06	0.52392
0.07	0.52790
0.08	0.53188
0.09	0.53586

x	$\Phi(x)$
1.00	0.84134
1.01	0.84375
1.02	0.84614
1.03	0.84849
1.04	0.85083
1.05	0.85314
1.06	0.85543
1.07	0.85769
1.08	0.85993
1.09	0.86214

## *Solution*

$$P(193,053 < E \leq 203,395) = \dots$$

$\vdots$

$$= [1 - \Phi(0.05)] - [1 - \Phi(1.05)]$$

$$= (1 - 0.51994) - (1 - 0.85314)$$

$$= 0.33320$$



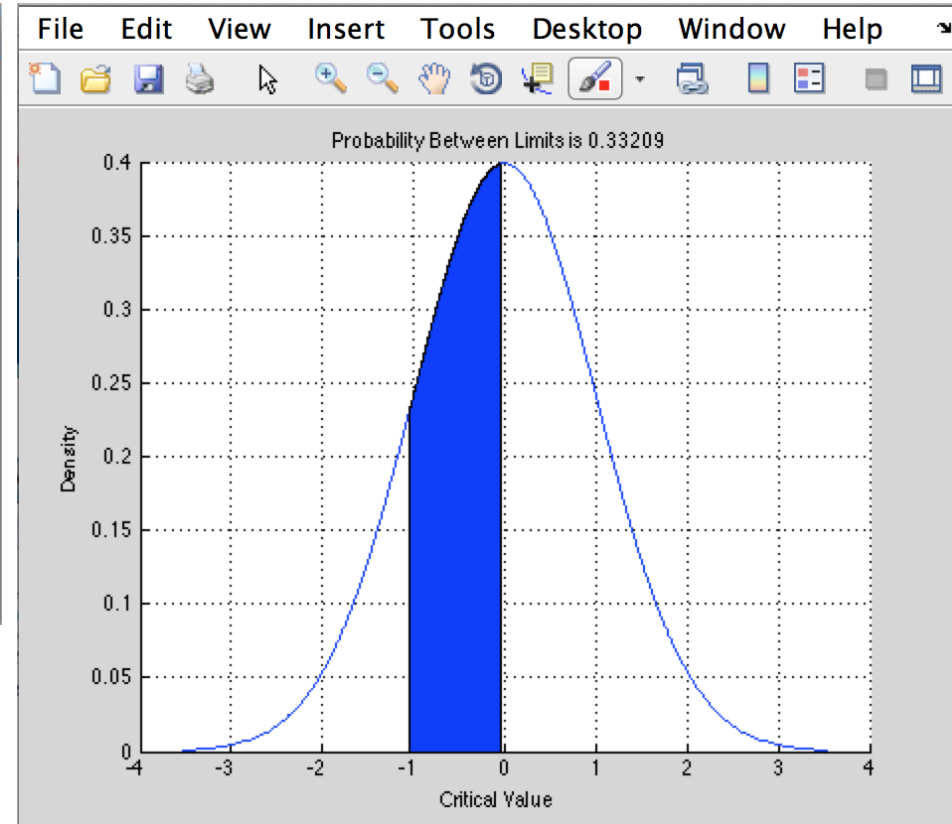




# Continuous Distributions

```
example_1a.m
1 - clear all;
2 - close all;
3 - clc;
4
5 - data = load('example01.dat');
6
7 - mu = mean(data);
8 - stdev=std(data);
9
10 %standardize lower and upper limits
11 - lower = (193053-mu)/stdev;
12 - upper = (203395-mu)/stdev;
13
14 %plot the z distribution
15 - normspec([lower upper], 0,1);
16 - grid on;
```

script Ln 14 Col 25



## Example (continued)

Assume the **design value** of Young's modulus for steel is **199,947** MPa, calculate

- (1) The probability of the Young's modulus being less than the **design value**.
- (2) The probability that Young's modulus will be at least the **design value**.



## *Solution (1)*

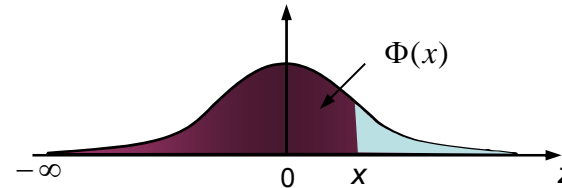
$$\mu = 203,919 \text{ MPa} \quad \sigma = 10,390 \text{ MPa}$$

$$\begin{aligned} P(E \leq 199,947) &= P(-\infty < E \leq 199,947) = \Phi\left(\frac{199,947 - 203,919}{10,390}\right) - \Phi\left(\frac{-\infty - 203,919}{10,390}\right) \\ &= \Phi(-0.38) - \Phi(-\infty) \\ &= [1 - \Phi(0.38)] - \Phi(-\infty) \\ &= \text{?????} \end{aligned}$$



## Table of the cumulative distribution function (CFD)

$$\Phi(x) = \frac{1}{\sqrt{2}} \int_{-\infty}^x e^{-(z^2/2)} dz$$



x	$\Phi(x)$
0.30	0.61791
0.31	0.62172
0.32	0.62552
0.33	0.62930
0.34	0.63307
0.35	0.63683
0.36	0.64058
0.37	0.64431
0.38	0.64803
0.39	0.65173

----->



## ***Solution (1)***

$$\mu = 203,919 \text{ MPa} \quad \sigma = 10,390 \text{ MPa}$$

$$P(E \leq 199,947) = \dots$$

$$= \vdots$$

$$= [1 - \Phi(0.38)] - \Phi(-\infty)$$

$$= (1 - 0.64803) - 0.0$$

$$= 0.35197$$

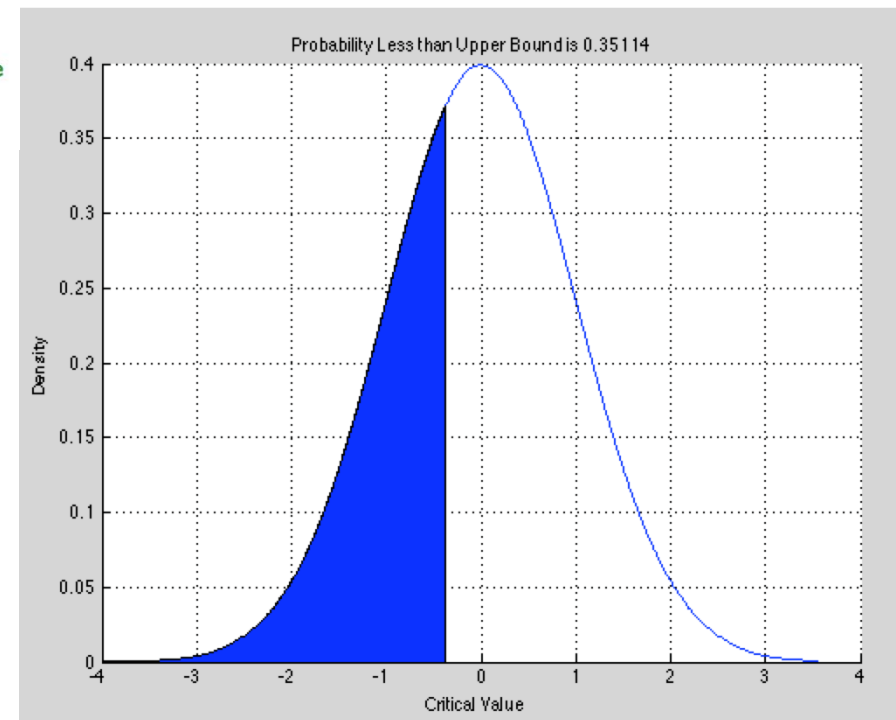
This means that the design value of  $E$  is approximately the 35<sup>th</sup> percentile value for the data given in the Table.





# Continuous Distributions

```
example_1b.m
1 - clear all;
2 - close all;
3 - clc;
4
5 - data = load('example01.dat');
6
7 - mu = mean(data);
8 - stdev=std(data);
9
10 - design_value = 199947;
11
12 - design_value_stand = (design_value-mu)/stdev;
13
14 %The probability of the E being less tha the design value
15 - normspec([-inf design_value_stand], 0,1);
16 - grid on;
17
```





## ***Solution (2)***

$$\mu = 203,919 \text{ MPa} \quad \sigma = 10,390 \text{ MPa}$$

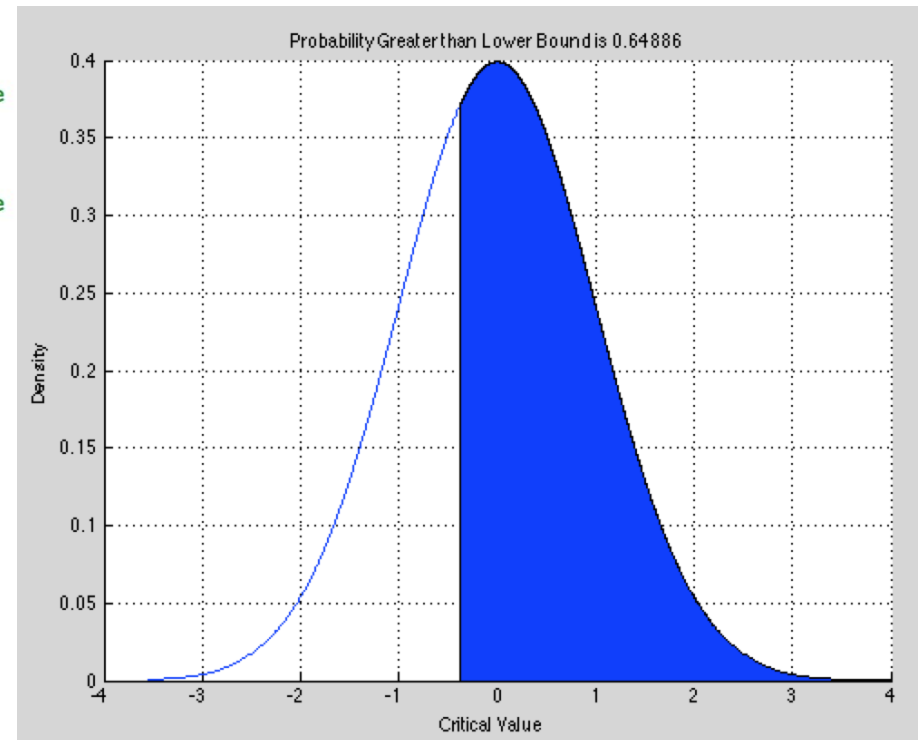
## **Similarly**

$$\begin{aligned} P(E \geq 199,947) &= P(199,947 < E \leq +\infty) = \Phi\left(\frac{+\infty - 203,919}{10,390}\right) - \Phi\left(\frac{199,947 - 203,919}{10,390}\right) \\ &= \Phi(+\infty) - \Phi(-0.38) \\ &= 1 - [1 - \Phi(0.38)] \\ &= 1 - (1 - 0.64803) \\ &= 0.64803 \end{aligned}$$



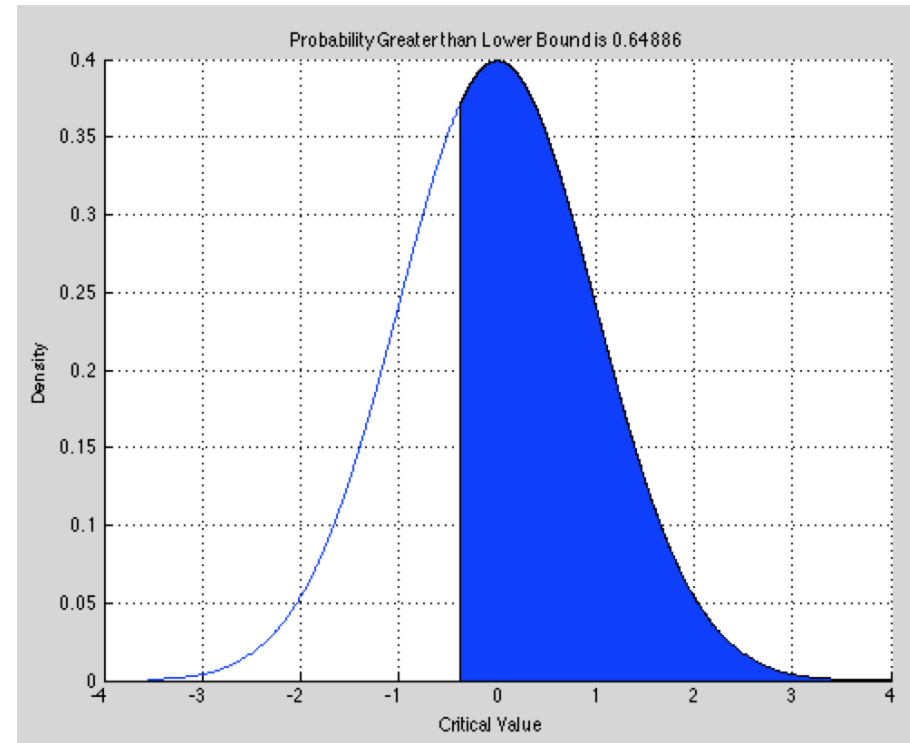
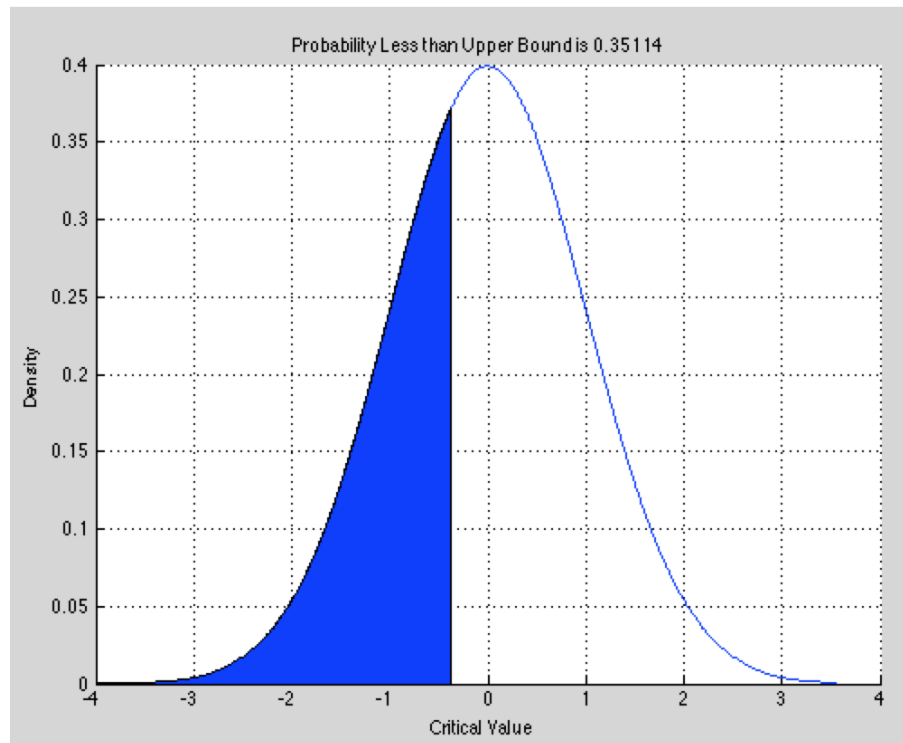
# Quality Risks

```
example_1b.m
1 - clear all;
2 - close all;
3 - clc;
4
5 - data = load('example01.dat');
6
7 - mu = mean(data);
8 - stdev=std(data);
9
10 - design_value = 199947;
11
12 - design_value_stand = (design_value-mu)/stdev;
13
14 %The probability of the E being less tha the design value
15 %normspec([-inf design_value_stand], 0,1);
16 %grid on;
17
18 %The probability of the E being at least the design value
19 normspec([design_value_stand inf], 0,1);
20 grid on;
```





# Quality Risks



## Example 2

Suppose a steel cable has to carry a weight of **44.5 kN**. Information on the strength of similar cables indicates that the strength of the cable,  $R$ , can be modeled by a normal random variable with a mean of 111.2kN and a standard deviation of 22.2 kN. Calculate the probability that the cable will break.





## ***Solution***



## ***Solution***

$$P(a < T < b) = \Phi\left(\frac{b - \mu}{\sigma}\right) - \Phi\left(\frac{a - \mu}{\sigma}\right)$$

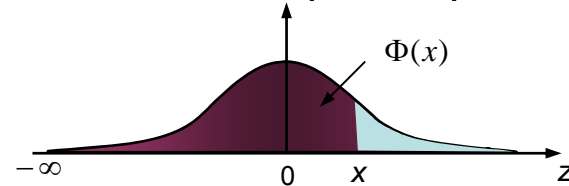
$$\begin{aligned} P(\text{the cable will break}) &= P(\text{failure}) = P(R \leq 44.5) = P(-\infty < R \leq 44.5) \\ &= \Phi\left(\frac{44.5 - 111.2}{22.2}\right) - \Phi\left(\frac{-\infty - 111.2}{22.2}\right) \\ &= \Phi(-3) - \Phi(-\infty) \\ &= 1 - \Phi(3) \\ &= \text{?????} \end{aligned}$$





## Table of the cumulative distribution function (CFD)

$$\Phi(x) = \frac{1}{\sqrt{2}} \int_{-\infty}^x e^{-(z^2/2)} dz$$



x	Φ(x)		x	Φ(x)
2.95	0.99841		3.05	0.99886
2.96	0.99846		3.06	0.99889
2.97	0.99851		3.07	0.99893
2.98	0.99856		3.08	0.99896
2.99	0.99861		3.09	0.99900
→ 3.00	0.99865		3.10	0.99903
3.01	0.99869		3.11	0.99906
3.02	0.99874		3.12	0.99910
3.03	0.99878		3.13	0.99913
3.04	0.99882		3.14	0.99916



## ***Solution***

$P(\text{the cable will break}) = \dots$

$\vdots$

$$= 1 - \Phi(3)$$

$$= 1 - 0.99865$$

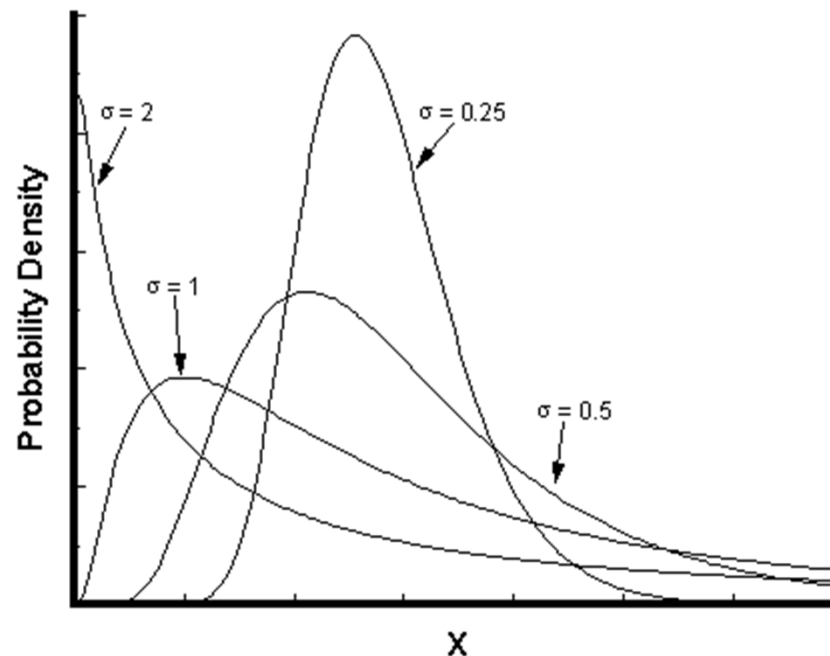
$$= 0.00135$$





- **Lognormal Distribution**

- A log-normal distribution is a probability distribution of a random variable whose logarithm is normally distributed
- In many engineering problems, a random variable cannot have negative value due to the physical aspects of the problem.





- The Lognormal probability density function (PDF) is

$$f(x) = \begin{cases} \frac{1}{\sigma x \sqrt{2\pi}} \cdot e^{-\frac{1}{2} \left( \frac{\ln x - \mu}{\sigma} \right)^2} & \text{if } x > 0 \\ 0 & \text{if } x \leq 0 \end{cases}$$

where

$\mu$ : mean

$\sigma$ : standard deviation

$\pi = 3.14159$

$e = 2.71828$

- The cumulative distribution function (CDF) is

$$Z = \frac{\ln x - \lambda_X}{\xi_X}$$

$$P(a < T < b) = F(b) - F(a)$$

$$= \frac{1}{\sqrt{2\pi}} \int_{\frac{\ln a - \lambda_X}{\xi_X}}^{\frac{\ln b - \lambda_X}{\xi_X}} e^{-\frac{1}{2}z^2} dz = \Phi\left(\frac{\ln b - \lambda_X}{\xi_X}\right) - \Phi\left(\frac{\ln a - \lambda_X}{\xi_X}\right)$$

$$\lambda_X = \ln \mu_X - \frac{1}{2} \xi_X^2 \quad \xi_X^2 = \ln \left[ 1 + \left( \frac{\sigma_X}{\mu_X} \right)^2 \right] = \ln(1 + COV^2)$$

$$\text{If } COV = \sigma / \mu \leq 0.3 \quad \xi_X = COV$$

### Example 3

The Young's modulus example with a **mean of 29,576 Pa** and a **standard deviation of 1507 Pa** can be considered.

It is assumed that the Young's modulus is log-normally distributed. Calculate

(1) The probability of  $E$  being less than the design value of 29,500 Pa.

(2) The probability of  $E$  having a value between 28,000 Pa and 29,500 Pa.



## ***Solution (1)***

$$COV = \sigma / \mu = 1507 / 29576 = 0.051 \leq 0.3$$

$$\xi = 0.051 \quad \lambda = \ln 29576 - 0.5 \times 0.051^2 = 10.293$$

$$P(E \leq 29,500)$$

$$= \Phi\left(\frac{\ln 29,500 - 10.293}{0.051}\right) - \Phi(-\infty)$$

$$= \Phi(-0.017)$$

$$= 1.0 - \Phi(0.017)$$

$$= 1.0 - 0.50678$$

$$= 0.493$$



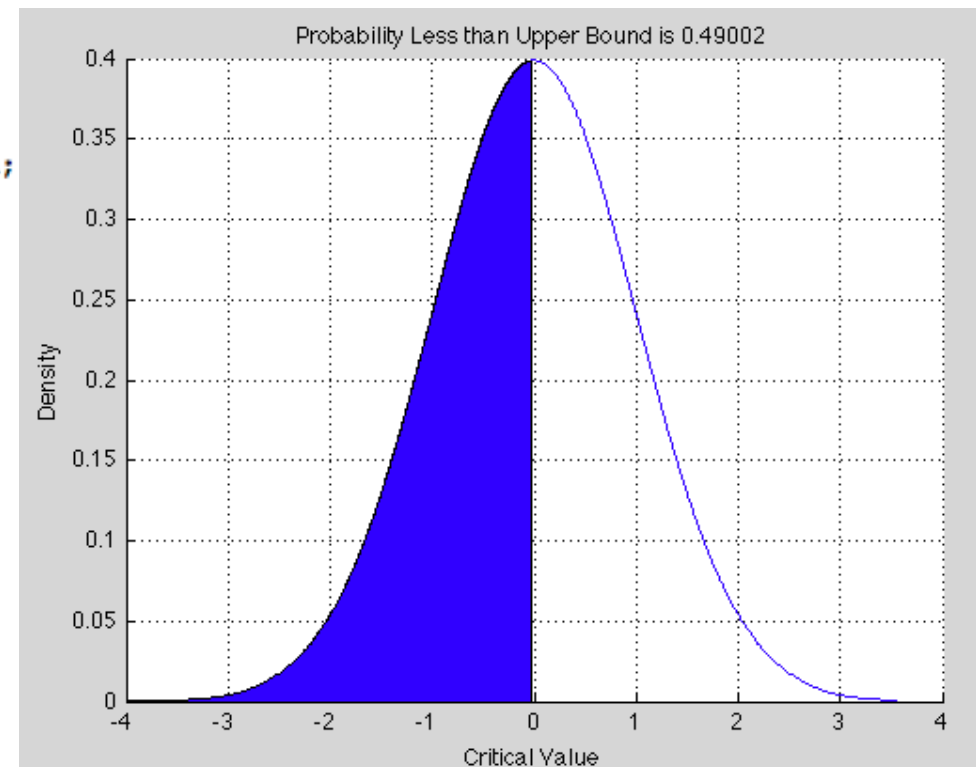
## ***Solution (2)***

$$\begin{aligned} &P(28,000 < E \leq 29,500) \\ &= \Phi\left(\frac{\ln 29,500 - 10.293}{0.051}\right) - \Phi\left(\frac{\ln 28,000 - 10.293}{0.051}\right) \\ &= \Phi(-0.017) - \Phi(-1.04) \\ &= (1.0 - \Phi(0.017)) - (1.0 - \Phi(1.04)) \\ &= (1.0 - 0.50678) - (1.0 - 0.85083) \\ &= 0.34405 \end{aligned}$$



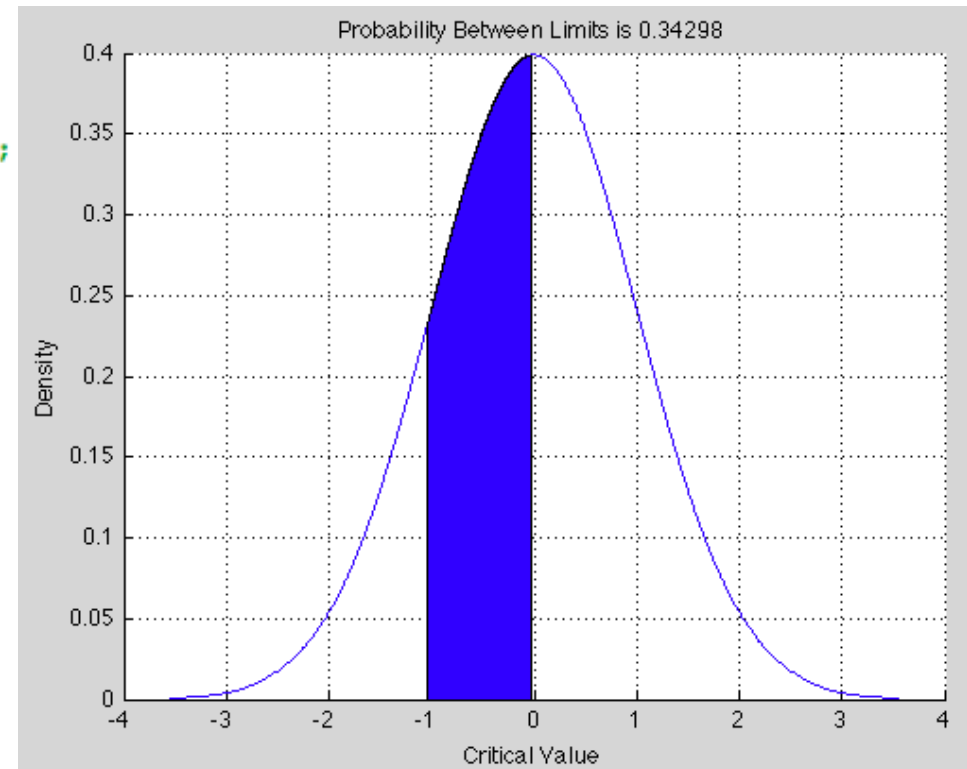


```
8 - mu = 29576;  
9 - stdev = 1507;  
10  
11 % Design value problem  
12 - design_value = 29500;  
13  
14 - cov = stdev/mu;  
15  
16 - if cov < 0.3  
17 -     zeta = cov;  
18 - else  
19 -     return;  
20 - end  
21  
22 - lambda = log(mu)-0.5*zeta^2;  
23  
24 - design_stand = (log(design_value)-lambda)/zeta;  
25  
26 - normspec([-inf design_stand],0,1);  
27 - grid on;  
28
```





```
8 - mu = 29576;  
9 - stdev = 1507;  
10  
11 % Design value problem  
12 design_value = 29500;  
13  
14 cov = stdev/mu;  
15  
16 if cov < 0.3  
17     zeta = cov;  
18 else  
19     return;  
20 end  
21  
22 lambda = log(mu)-0.5*zeta^2;  
23  
24 %design_stand = (log(design_value)-lambda)/zeta;  
25  
26 %normspec([-inf design_stand],0,1);  
27 %grid on;  
28  
29 % probability interval  
30  
31 lower = (log(28000)-lambda)/zeta;  
32 upper = (log(29500)-lambda)/zeta;  
33  
34 normspec([lower upper],0,1);  
35 grid on;
```



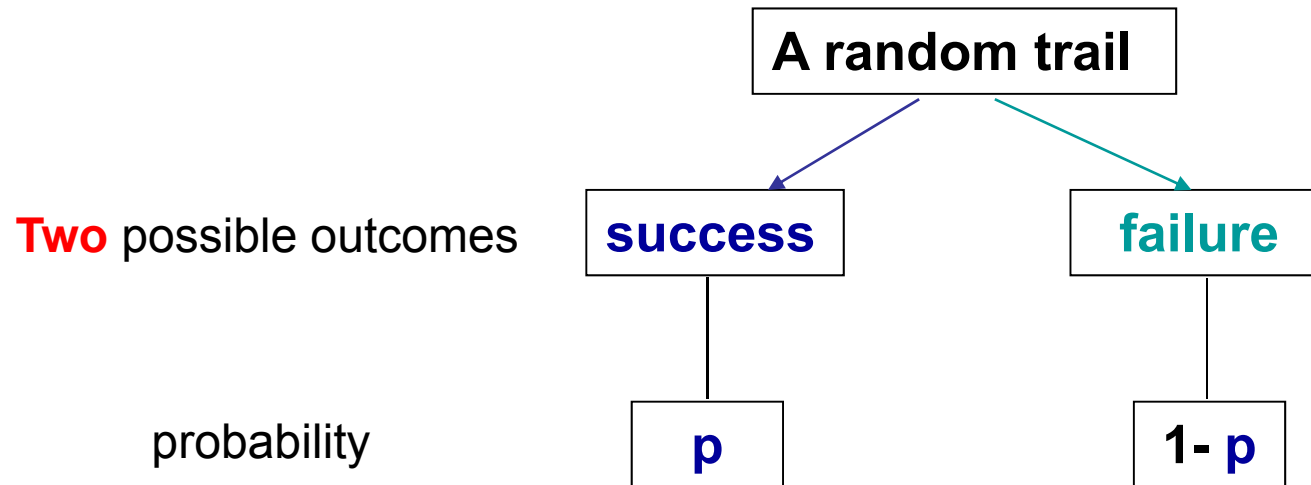


# Discrete Probability Distributions

- **Binomial** distribution
- **Poisson** distribution

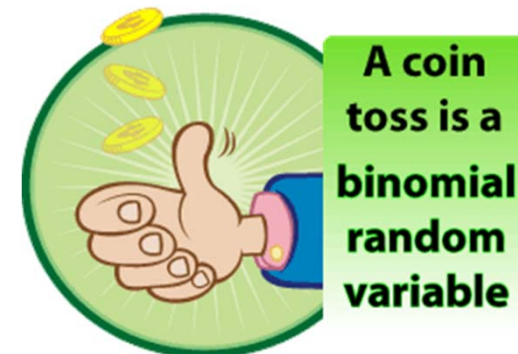


# Discrete Distribution – Binomial



$X$  – total number of successes,  $X \geq 0$

$S$  – sample space,  $S = \{0, 1, 2, \dots, n\}$



The probability distribution of random variable  $X$  is given by the

**binomial distribution.**

- Binomial Mean =  $n \times p$
- Binomial Standard Deviation =  $\sqrt{n \times p \times (1 - p)}$
- Binomial probability formula

$$P(X = x, n | p) = \binom{n}{x} p^x (1 - p)^{n-x}, \quad x = 0, 1, 2, \dots, n$$

By definition  $\binom{n}{x} = \frac{n!}{x!(n-x)!}$

where  $n! = n(n-1)(n-2) \dots 1$ , and  $0! = 1$

$n$ : Bernoulli trials      $p$ : probability

$x$ : exactly  $x$  successes out of  $n$  Bernoulli trials

## Example 4

A random sample of **15** valves is observed. From past experience, it is known that the probability of a given failure within **500** hours following maintenance is **0.18**.

Calculate the probability that these valves will experience **0**, **1**, and **2** independent failures within **500 hours** following their maintenance.







***Solution:***



## **Solution:**

Here the random variable **X** designate the failure of valve that can take on values of 0, 1, and 2.

$$p = 0.18 \quad P(X = x, n|p) = \binom{n}{x} p^x (1-p)^{n-x} = \frac{n!}{x!(n-x)!} p^x (1-p)^{n-x}$$

Using '**Binomial probability formula**'

$$p(X = 0, 15|0.18) = \frac{15!}{0!(15-0)!} 0.18^0 (1-0.18)^{15-0} = 5.10\%$$

$$p(X = 1, 15|0.18) = \frac{15!}{1!(15-1)!} 0.18^1 (1-0.18)^{15-1} = 16.8\%$$

$$p(X = 2, 15|0.18) = \frac{15!}{2!(15-2)!} 0.18^2 (1-0.18)^{15-2} = 25.8\%$$



```
1 % valve example - binomial distribution
2 - clear all;
3 - clc;
4
5 - failure_rate = 0.18;
6 - trail = 15;
7
8 %% 0 valve failure
9
10 - zero_failure = binopdf(0, trail, failure_rate);
11 - display(zero_failure);
12
13 %% 1 valve failure
14
15 - one_failure = binopdf(1, trail, failure_rate);
16 - display(one_failure);
17
18 %% 2 valve failure
19
20 - two_failure = binopdf(2, trail, failure_rate);
21 - display(two_failure);
```



Command Window

New to MATLAB? Watch this

zero\_failure =

0.0510

one\_failure =

0.1678

two\_failure =

0.2578



## Example 5

Suppose the probability of failure of a structure due to earthquakes is estimated as  $10^{-5}$  per year. Assuming that the design life of the structure is 50 years and the probability of failure in each year remains constant and independent during its lifetime.

Estimate the probability of no failure using the binomial distribution.





# Risk of Building Collapsing

***Solution:***

## Solution

$$P(X = x, n|p) = \binom{n}{x} p^x (1-p)^{n-x} = \frac{n!}{x!(n-x)!} p^x (1-p)^{n-x}$$

$$P(\text{no failure in 50 years}) = P(X = 0, 50|10^{-5})$$



$$= \binom{50}{0} (10^{-5})^0 (1-10^{-5})^{50-0}$$

$$= \frac{50!}{0!(50-0)!} (1-10^{-5})^{50}$$

$$\approx 1 - 50 \times 10^{-5} = 99.95\%$$

$$\begin{aligned} P(\text{failure in 50 years}) &= 1 - P(\text{no failure in 50 years}) \\ &= 1 - 0.99950 = 0.05\% \end{aligned}$$





## Example 6

The drainage system of a city has been designed for a rainfall intensity that will be exceeded on an average **once in 50 years**.

What is the probability that the city will be flooded in 2 out of 10 years?





***Solution:***



## Solution

Since the possible outcomes in each year consist of flooding, or nonflooding, the problem can be modeled as **binomial distribution**.

$$P(X = x, n|p) = \binom{n}{x} p^x (1-p)^{n-x} = \frac{n!}{x!(n-x)!} p^x (1-p)^{n-x}$$

The probability of flooding in one year

$$P = 1/50 = 0.02$$

$$P(\text{flooding in 2 out of 10 years}) = P(X = 2, 10|0.02)$$

$$= \binom{10}{2} (0.02)^2 (1-0.02)^{10-2}$$

$$= \frac{10!}{2!(10-2)!} (0.02)^2 (0.98)^8 = 1.5\%$$



- The first occurrence time of an event is of great interest in Engineering.
  - The first time the design wind speed will be exceeded in an area
  - The first time a structure will be damaged by earthquakes

## **Geometric Distribution**

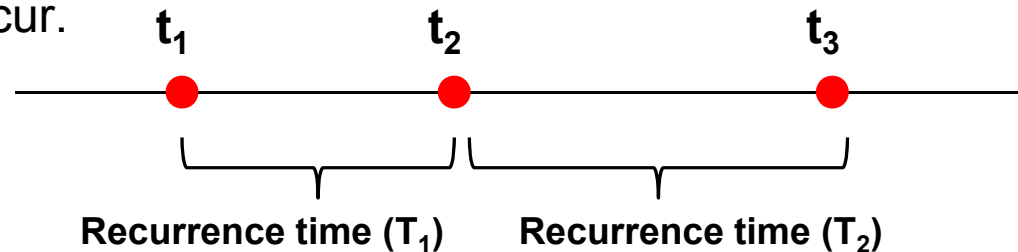
$$P(T = t) = p(1 - p)^{t-1}, t = 1, 2, \dots$$

- The events occur in a Bernoulli sequence
- $p$  is the probability of occurrence in the each trial, then the probability that the event will occur for the first time at the  $t$ -th trial.
- This is **no occurrence** in the previous  $(t-1)$  trials.



# Return Period - Binomial

- Return Period (also known as a recurrence interval)
  - An estimate of the likelihood of an event, such as an earthquake, flood or a river discharge flow to occur.



- Assuming a Bernoulli sequence, recurrence time must follow the probabilistic characteristics of the first occurrence (i.e. the geometric distribution)
- Mean recurrence time (**Return period**)

$$T = E(t) = \sum_{t=1}^{\infty} t p_T(t) = \sum_{t=1}^{\infty} t p (1-p)^{t-1} = p \underbrace{\left[ 1 + 2(1-p) + 3(1-p)^2 + 4(1-p)^3 + \dots \right]}_{1/p^2} = p \times \frac{1}{p^2} = \frac{1}{p}$$

**Example:** A return period of 50 years for the design flood level indicates that **on average** there will be a flood once every 50 years. However, there is a probability that no flood will occur in the next 50 years.

## Example 7

Suppose the probability of failure of a structure due to earthquakes is estimated as  $10^{-5}$  per year. Assuming that the probability of failure in each year remains constant and independent during its lifetime.

Estimate the probability of the failure of the structure due to earthquakes for the first time in the 10<sup>th</sup> year using the geometric distribution.





# Risk of Structure Collapsing

***Solution:***

## ***Solution***

**The probability of the failure of the structure in the 10<sup>th</sup> year.**

$$P(T = t) = p(1 - p)^{t-1}, t = 1, 2, \dots \quad p = 10^{-5} \quad t = 10$$

$$P(T = 10) = 10^{-5} (1 - 10^{-5})^{10-1} = 9.999 \times 10^{-6}$$





## Discrete Distribution – Poisson

- Poisson distribution fits cases of **rare events** that occur in a fixed amount of time **OR** in a specified region
- Poisson random variable  $X$  is successes in a time interval.
- The **Probability Mass Function** (PMF) is

$$P(x \text{ occurrences in time } t) = \frac{(vt)^x}{x!} e^{-vt} \quad x = 0, 1, 2, \dots$$

where

$t$ : time period

$v$ : mean occurrence rate of events at a location

$x$ : occurrences

$e = 2.71828$



Siméon Denis Poisson  
(1781-1840)

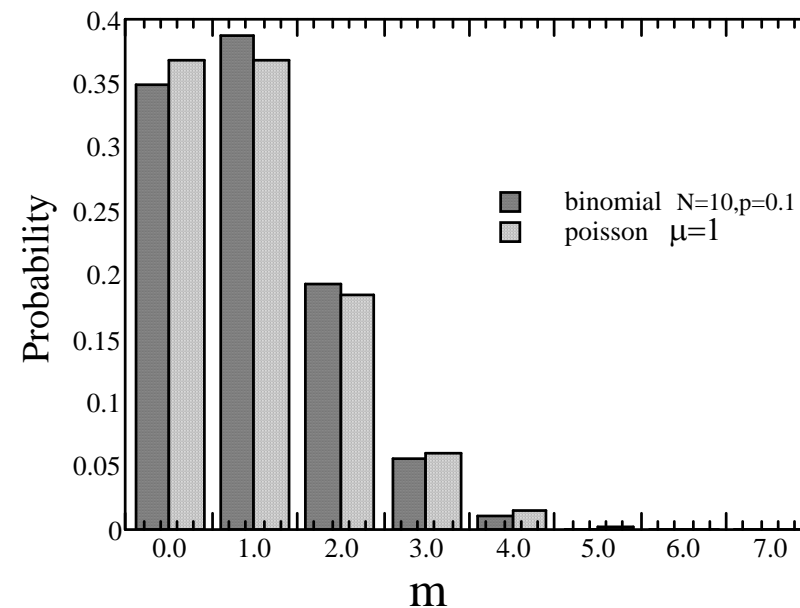
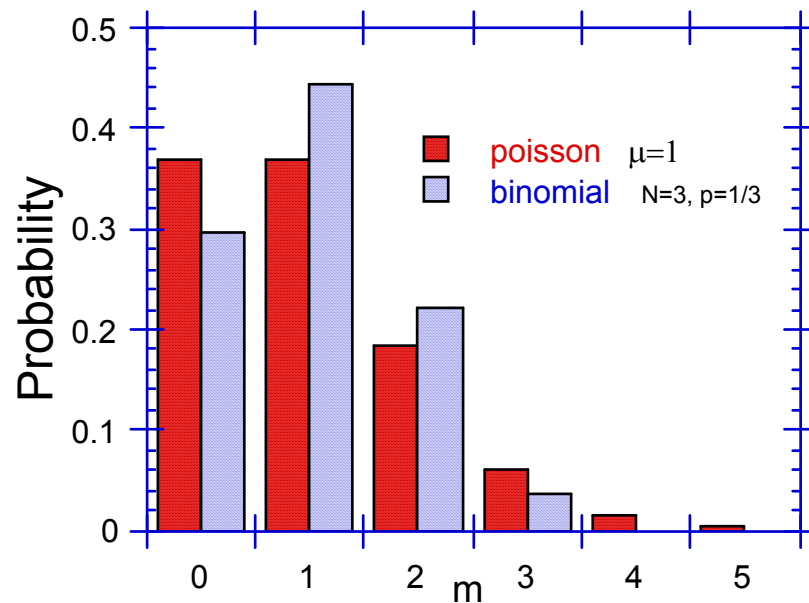


- Poisson distribution can be derived by taking the appropriate limits of the Binomial distribution

For large  $n$  and fixed  $\mu$ : Binomial  $\Rightarrow$  Poisson

Binomial Mean =  $n \times p$

Poisson Mean =  $\nu$



$m$  special events (success)



## Example 8

From records of past **50 years**, it is observed that tornadoes occur in a particular area an average of **two times a year**.

Calculate the probability of no tornadoes in the next year.





## ***Solution***

## **Solution**

$$P(x \text{ occurrences in time } t) = \frac{(vt)^x}{x!} e^{-vt}$$

In this case,  $v = 2/\text{year}$ , next year  $t = 1$  year

$x = 0$ , and  $t = 1$  year

$$P(\text{no tornado next year}) = \frac{(2 \times 1)^0 \cdot e^{-2 \times 1}}{0!} = 13.5\%$$

$x = 2$ , and  $t = 1$  year

$$P(\text{exactly 2 tornado next year}) = \frac{(2 \times 1)^2 \cdot e^{-2 \times 1}}{2!} = 27.1\%$$

$x = 0$ , and  $t = 50$  year

$$P(\text{no tornado in next 50 years}) = \frac{(2 \times 50)^0 \cdot e^{-2 \times 50}}{0!} = 3.72 \times 10^{-44}$$

**Impossible !**



# Risks of Weather-Related Disasters

```
%no tornado next year
t = 1; %time period
v = 2; %mean occurrence
x = 0; %occurrence

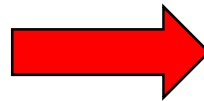
occurance = poisspdf(x,v*t);
display(occurance);

%2 tornado next year
t = 1; %time period
v = 2; %mean occurrence
x = 2; %occurrence

occurance = poisspdf(x,v*t);
display(occurance);

%0 tornado in 50 years
t = 50; %time period
v = 2; %mean occurrence
x = 0; %occurrence

occurance = poisspdf(x,v*t);
display(occurance);
```



## Command Window

```
occurance =
    0.1353

occurance =
    0.2707

occurance =
    3.7201e-44
```

## Example 9

The safety of a building in an earthquake-prone area is under consideration.

The past **100 years** of data indicate that there were **four strong earthquakes** in the area. Assume that damage event for different earthquakes are statistically independent.

(a) What is the probability that there will be no strong earthquakes in the area in 50 years, during the service life of the building?

(b) What is the probability that there will **only two strong** earthquakes in 50 years?





## ***Solution***



**Solution**  $P(x \text{ occurrences in time } t) = \frac{(vt)^x}{x!} e^{-vt}$

(a) The average rate of strong earthquake occurrence

$$\nu = 4/100 = 0.04 \text{ per year}$$

$$\text{Thus, } \nu t = 0.04 \times 50 = 2$$

$$P(\text{no strong earthquake in 50 years}) = P(X = 0)$$

$$= \frac{e^{-2} \times 2^0}{0!} = 0.13534$$

(b)  $P(\text{two strong earthquake in 50 years}) = P(X = 2)$

$$= \frac{e^{-2} \times 2^2}{2!} = 0.27067$$





## Example 10

A nuclear plant receives its electric power from a utility grid outside of the plant. From past experience, it is known that **loss of grid power occurs at a rate of once a year.**

What is the probability that over a period of **3** years no power outage will occur?







## ***Solution***



# Risk of Power Supply Failure

**Solution:**  $P(x \text{ occurrences in time } t) = \frac{(vt)^x}{x!} e^{-vt}$

Denote,  $v = 1/\text{year}$ ,  $t = 3 \text{ years}$ ,  $vt = 1 \times 3 = 3$

Using **Poisson Probability Distribution** find

$$P(X = 0) = \frac{3^0 \times e^{-3}}{0!} = 5\%$$



## Example 11

For a large construction project, the contractor estimates that the average rate of on-the-job accidents is **three times per year**. From past experience, the contractor also estimates that the cost incurred for each accident may be modeled as a lognormal random variable with **a median of \$6,000** and **COV of 20%**. The cost of each accident can be assumed to be statistically independent.





- (a) What is the probability that there will be no accident in the first month of construction?
- (b) What is the probability that an accident will incur a loss exceeding \$4,000?





## ***Solution***



## **Solution**

$$P(x \text{ occurrences in time } t) = \frac{(vt)^x}{x!} e^{-vt}$$

(a) For the Poisson distribution

$$v = 3 \text{ times year} = 3/12 = 1/4 \text{ time per month}$$

$$t = 1 \text{ month}$$

$$vt = 1/4 \cdot 1 = 1/4$$

$$x = 0 \text{ (no accident)}$$



$$P(\text{no accident in the month}) = \frac{e^{-\frac{1}{4}} \cdot \left(\frac{1}{4}\right)^0}{0!} = e^{-\frac{1}{4}} = 0.7788$$

## Solution

(b) The cost incurred for each accident is modeled as a **lognormal distribution**.

In this case,  $COV = 20\% = 0.2 \leq 0.3$       $\xi = COV = 0.2$

$$\lambda = \ln(\text{median}) = \ln 6,000 - 0.5 \times 0.2^2 = 8.7$$

$$\begin{aligned} P(\text{cost of accident} > \$4,000) &= \Phi\left(\frac{\ln \infty - 8.7}{0.20}\right) - \Phi\left(\frac{\ln 4,000 - 8.7}{0.20}\right) \\ &= 1 - \Phi(-2.027) \\ &= 1 - 0.0213 \\ &= 97.87\% \end{aligned}$$





- The probability distribution that describes the time between events in a Poisson process.
- It is the continuous analogue of the Geometric Distribution.
- Relationship between Exponential and Poisson distribution
  - Poisson is a discrete random variable that measures the number of occurrence of some given event over a specific interval (time, distance)
  - Exponential describes the length of the interval between occurrence.





- PDF of the exponential distribution is

$$f_T(t) = \nu e^{-\nu t}$$

where

$t$ : time period

$\nu$ : the average rate of occurrences

$e = 2.71828$

**Return Period (Poisson Distribution)**

$$T = \int_0^{\infty} t f_T(t) dt = \int_0^{\infty} t \nu e^{-\nu t} dt$$



## Example 12

Strong earthquakes in an area are assumed to occur according to the Poisson distribution with the average rate of occurrences of **0.04 per year**. Assuming the time between two consecutive occurrences of strong earthquakes can be modeled by an Exponential distribution, determine the return period of strong earthquakes.



## ***Solution***



## ***Solution***

$v = 0.04$  per year

## **Return Period**

$$T = \int_0^{\infty} t \times 0.04 \times e^{-0.04t} dt = 25 \text{ years}$$