

L6

Basic Statistics For Economist

Spring 2020

Department of Statistics

Last time

Bayes' theorem

- Reverse conditional prob.: $P(E_i|A) = \frac{P(A|E_i)P(E_i)}{P(A)} = \frac{P(A|E_i)P(E_i)}{\sum_k P(A|E_k)P(E_k)}$

Discrete random variables

- Probability function: $P(X = x) = P_X(x)$
- Cumulative probability functions: $P(X \leq x) = F_X(x)$
- Expected value μ_X and variance σ_X^2 : how to calculate
- Linear combination $Y = a + bX$ and standardization $Z = \frac{X - \mu_X}{\sigma_X}$

Bernoulli experiment, Bernoulli distribution, 0-1 outcome

- parameter: P = probability of success = $P(X = 1)$



Last time

Many Bernoulli experiments = **Binomial distribution** = *Bin*(n, P)

- Sum of n independent Bernoulli with identical P
- parameters n and P
- *Bernoulli* = *Bin*(1, P)
- $\mu_X = E(X) = nP, \quad \sigma_X^2 = \text{Var}(X) = nP(1 - P)$
- Probability function: $P(x) = \binom{n}{x} P^x (1 - P)^{n-x}$
- Calculate probabilities
 - Direct calculation using the probability function
 - Using table $F(x) = P(X \leq x); P(x) = F(x) - F(x-1)$

Today

- **Continuous random variables**

- Probability functions for **continuous** r.v.
- **The density function** and the cumulative probability function
- Expected value and variance of continuous r.v.
- A particularly common distribution:

- ***The normal distribution***

- Standardized normal distribution
- Calculate probabilities using a table
- Approximate Binomial with a Normal (Ch. 5.4 is left for F8)



Continuous variables

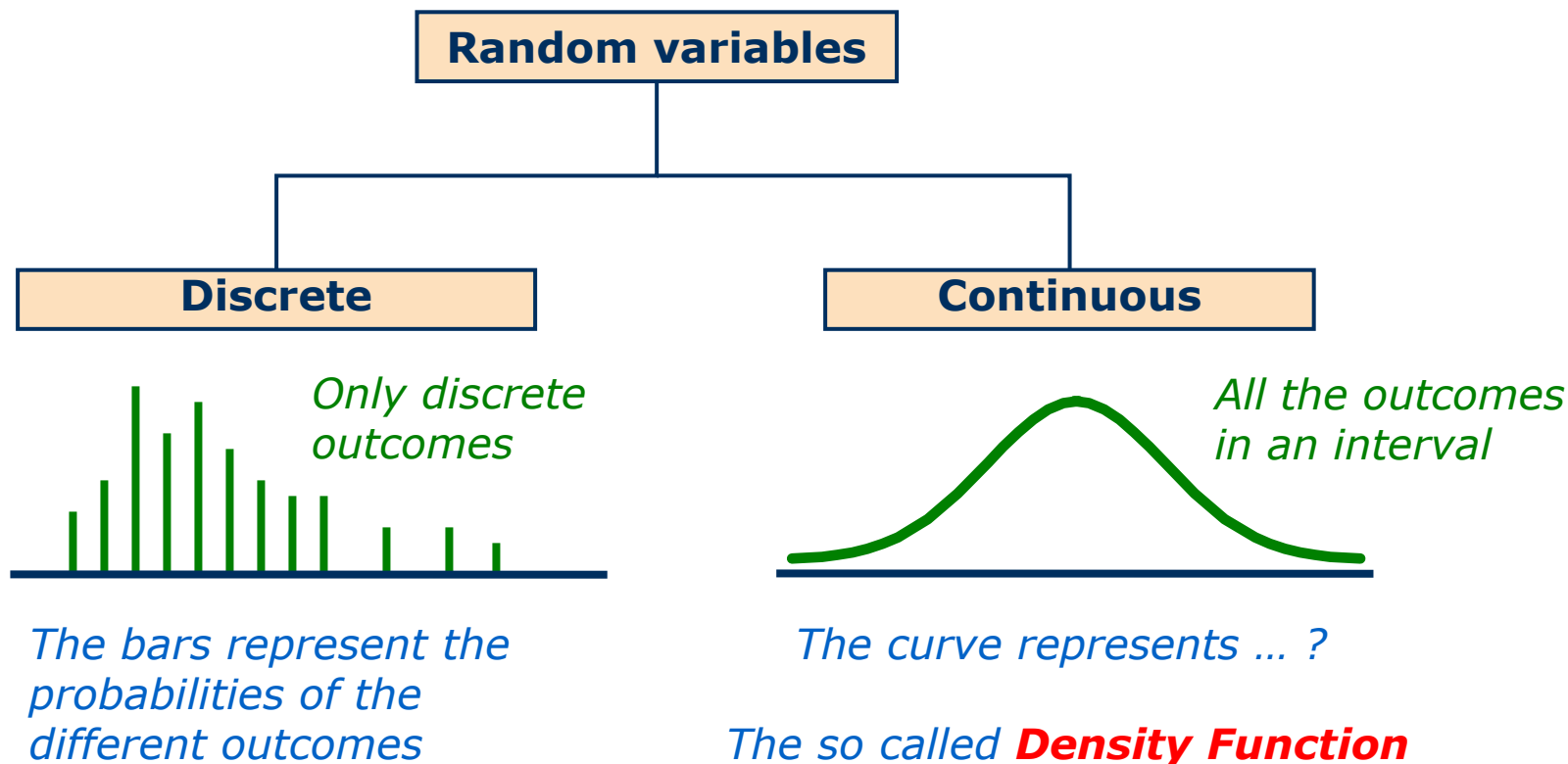
- A **continuous variable** can *map* to **any value in an interval** (or many intervals). A discrete variable can only map to a *discrete* collection of values.

Example:

- The number of units that need repairing among 10 chosen ones is a discrete variable.
 - Can be listed: ex. $S_X = \{0, 1, 2, \dots, 10\}$
- The time it takes to repair a unit is a continuous variable.
 - Cannot be listed
 - Continuous interval: e.g. $S_X = (0 \leq x \leq 100)$



Random variables, continued



Example of continuous random variables

Let X = "the repair time of a unit"

- Sample space: $S_X = \{0 \leq X \leq 100\}$
- S_X = the elementary outcomes (disjoint) of the experiment are numbers, but they cannot be listed!
- We may define events as **sub intervals** of S_X :

$A = [0,2]$	$0 \leq X \leq 2$	$X \leq 2$
$\bar{A} = (2,100]$	$2 < X \leq 100$	$X > 2$
$B = [2,100]$	$2 \leq X \leq 100$	$X \geq 2$
$C = \{11\}$	$X = 11$, i.e. exactly = 11	

Notice! For continuous variables it does not matter whether the inequality is strict or not!

Cumulative distribution functions

A cumulative distribution function

$$F_X(x) = P(X \leq x)$$

Probability that X does not exceed the value x expressed as a function of x .

The probability of an event = the probability of a subinterval of S_X

Calculated as

$$P(a \leq X \leq b) = P(X \leq b) - P(X < a) = F_X(b) - F_X(a)$$

where $a < b$.



The density function for a continuous r.v.

- The probability density function (*abbreviated pdf*)
- Denoted $f_X(x) \neq P(X = x)$
- The density function is typically a continuous “curve” but its values are not probabilities but “densities” as a function of x .

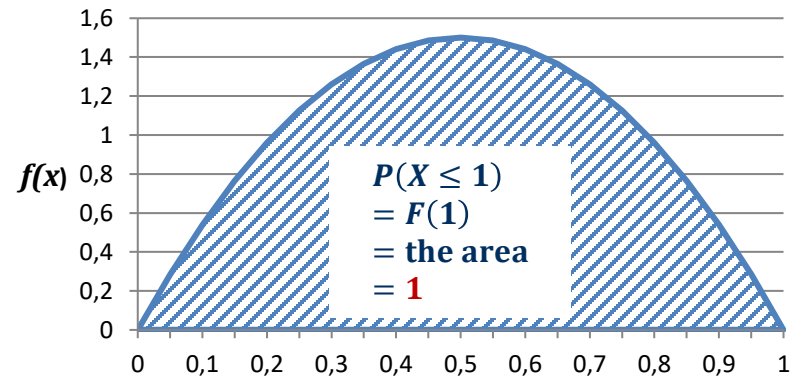
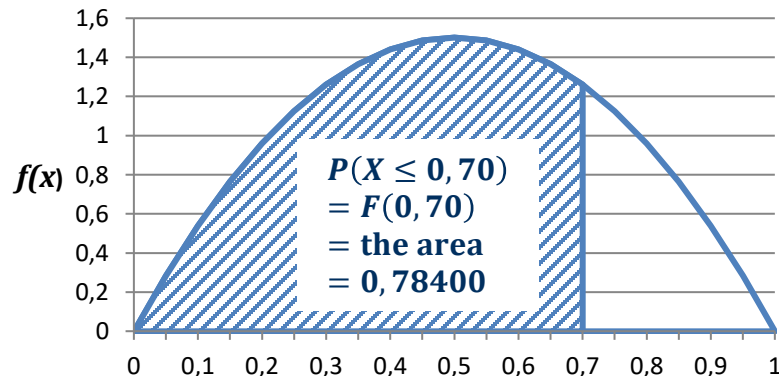
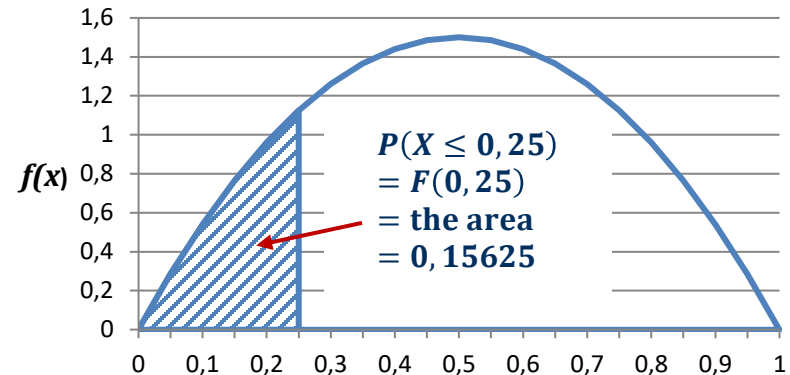
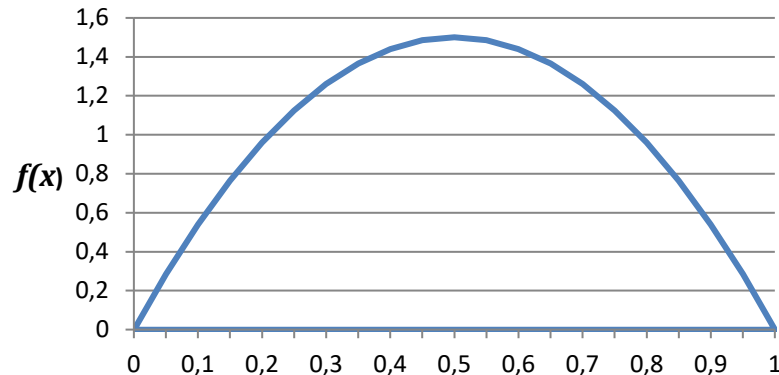
1. A density function has the property $f_X(x) \geq 0$ for all $x \in S_X$
2. The area under the curve of $f_X(x)$ over all of S_X must be $= 1$
3. Let a and b be two points in S_X such that $a < b$. Then the probability that X takes a value between a and b = the area under the curve between these points: $P(a < X < b) = \int_a^b f_X(x) dx$



Notice! $f(x)$ may be > 1
But the **areas** under $f(x)$ must be ≤ 1

Graphical illustration

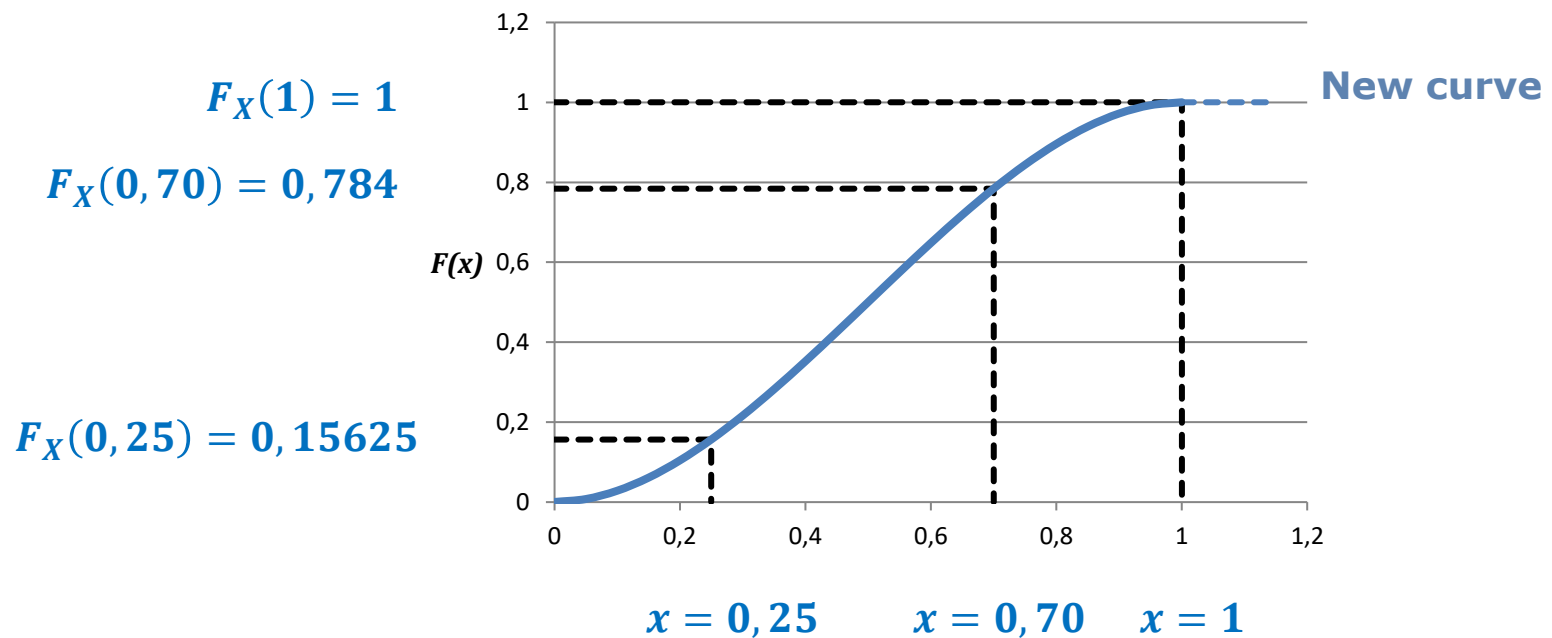
- A simple density function: $f_X(x) = 6x - 6x^2$, $S_X = (0,1)$



*You do not need to know
how I got $F_X(x)$ from $f_X(x)$.*

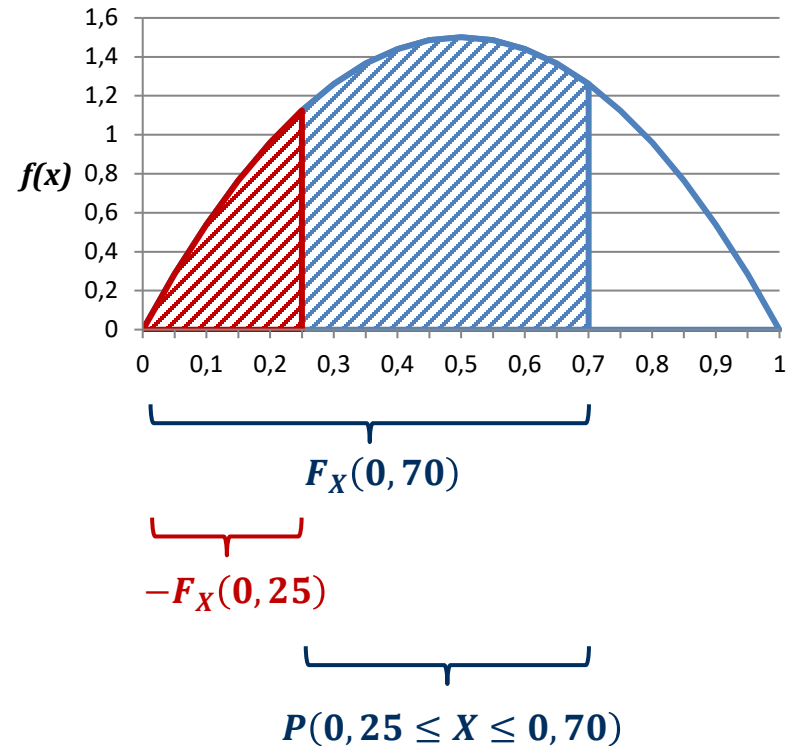
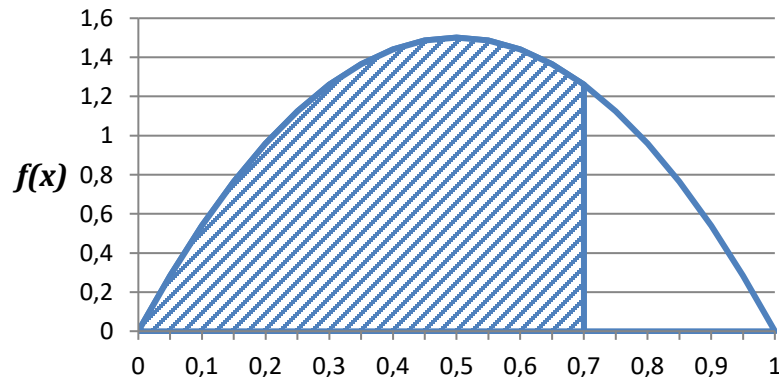
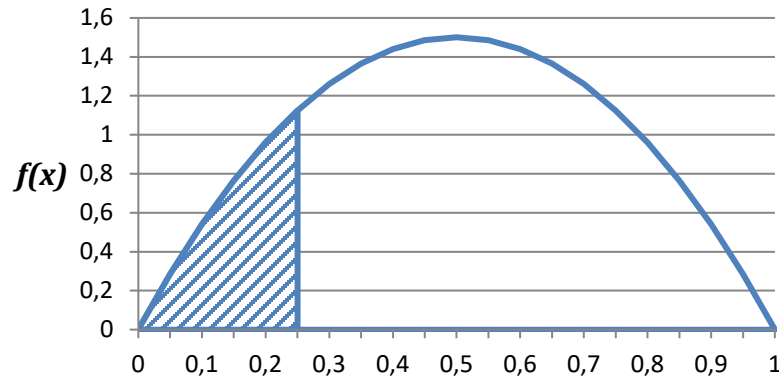
Graphical illustration, cont.

- The cumulative probability function: $F_X(x) = 3x^2 - 2x^3$



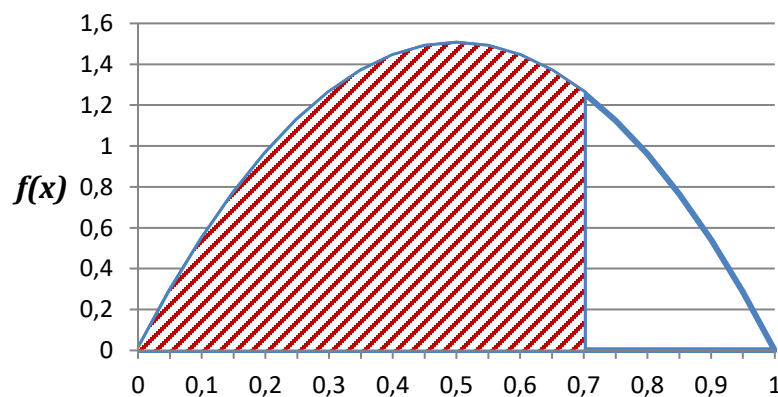
Graphic illustration, cont.

- Calculate probabilities: $P(0,25 \leq X \leq 0,70) = F_X(0,70) - F_X(0,25)$



The probability of a single point x

- Suppose you want to calculate $P(X = x)$.
- We have that $P(X = x) = P(X < x) - P(X \leq x)$. *Draw!*
- The probability is defined as the area under the curve $f_X(x)$.



- Remove all area except at $x = 0,70$
- What is the area of a line?
- **Zero. (0)**

- It does NOT matter if it says $<$ or \leq , it is the same!
- $P(X < x) = P(X \leq x) = F_X(x)$



Summary measures

- The density function $f_X(x)$ alternatively the cumulative density function $F_X(x)$ includes all necessary information about X .
 - *Assumed but not stated: we know what S_X is*
- **Summary measures** describe some important attributes of a probability function.
- **"Location" – expected value**
 - theoretical average, what we can expect the average to be
- **"Spread" – variance**
 - How spread out the values are

The expected value of a continuous r.v.

Expected value is defined as: $\mu_X = E(X) = \int_{x \in S_X} x f_X(x) dx$

- The expected value is a **weighted average** of the elements in S_X weighted with respect to the density function
- **You do not need to know calculus!**

Expected value of a function $g(x)$: $E[g(X)] = \int_{x \in S_X} g(x) f_X(x) dx$

Variance for a continuous r.v.

Variance is defined as: $\sigma_X^2 = \text{Var}(X) = \int_{x \in S_X} (x - \mu_X)^2 f_X(x) dx$

- Weighted average of the squared distance to the mean expected value
- Again, **you do not need to know calculus!**

Linear combinations

- En function the type $Y = g(X) = a + bX$ is called a **linear combination**
- Suppose $E(X) = \mu_X$ and $Var(X) = \sigma_X^2$

Then:

$$\mu_Y = E(Y) = E(a + bX) = a + b\mu_X$$

$$\sigma_Y^2 = Var(Y) = Var(a + bX) = b^2\sigma_X^2$$

Notice! b squared, b^2

- If you have a linear function $Y = g(X)$ you may calculate expected value and variance of Y directly, **this is also true for continuous r.v.**

Standardization

Standardization **IMPORTANT!**

Suppose
$$Z = \frac{X - \mu_X}{\sigma_X} = -\frac{\mu_X}{\sqrt{\sigma_X^2}} + \frac{1}{\sqrt{\sigma_X^2}} \cdot X$$

$$\mu_Z = -\frac{\mu_X}{\sqrt{\sigma_X^2}} + \frac{1}{\sqrt{\sigma_X^2}} \cdot \mu_X = 0 \qquad \sigma_Z^2 = \mathbf{b^2} \sigma_X^2 = \left(\frac{1}{\sqrt{\sigma_X^2}} \right)^2 \cdot \sigma_X^2 = 1$$

The normal distribution

- One of the most important, most famous, and most used probability functions of all time, for better or for worse
 - “magnificent” mathematical properties
 - Occurs in nature (or as good approximation)
 - Particularly useful when one has many observations of X (We will return to this in L8)

- Alternative names:

Bell curve

Gaussian distribution

after J.C.F. Gauss (1777-1855)



The normal distribution

A **normally distributed** r.v. X has the density function

$$f_X(x) = \frac{1}{\sqrt{2\pi\sigma_X^2}} \exp\left(-\frac{(x - \mu_X)^2}{2\sigma_X^2}\right)$$

where μ_X and σ_X^2 are the function's **parameters** which fully determine what the distribution looks like and how you write $X \sim N(\mu_X, \sigma_X^2)$.

The sample space: $S_X = (-\infty, \infty)$ = all real numbers

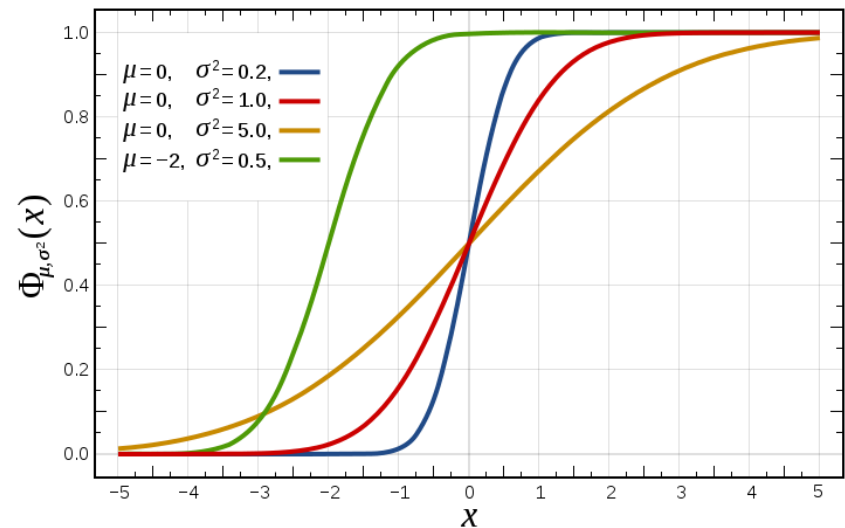
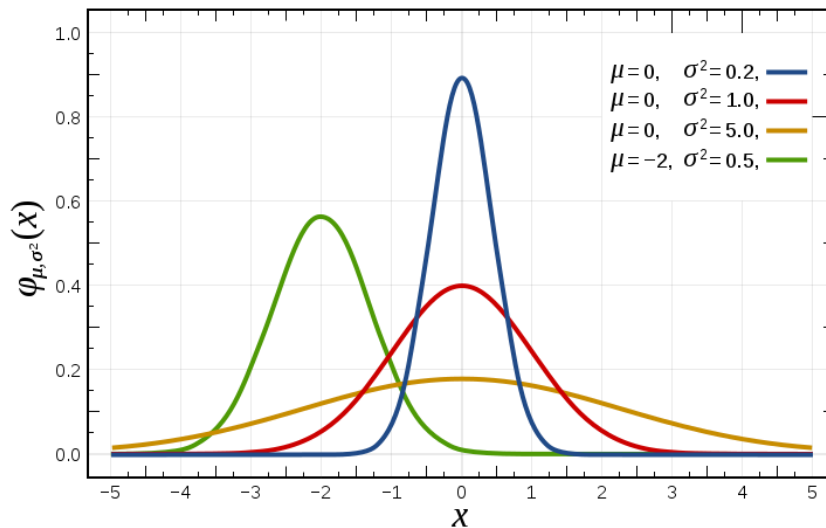
Expected value: $E[X] = \mu_X$ Variance: $Var(X) = \sigma_X^2$

The expected value and variance are equal to the parameters μ_X and σ_X^2 respectively.



Different normal distributions

- Density functions and cumulative probability functions for different values of μ_X and σ_X^2



(Source: Wikipedia, https://en.wikipedia.org/wiki/Normal_distribution)

Normal distribution, cont.

- The maximum of the curve
= expected value $E(X) = \mu_X$

- Symmetrical around μ_X

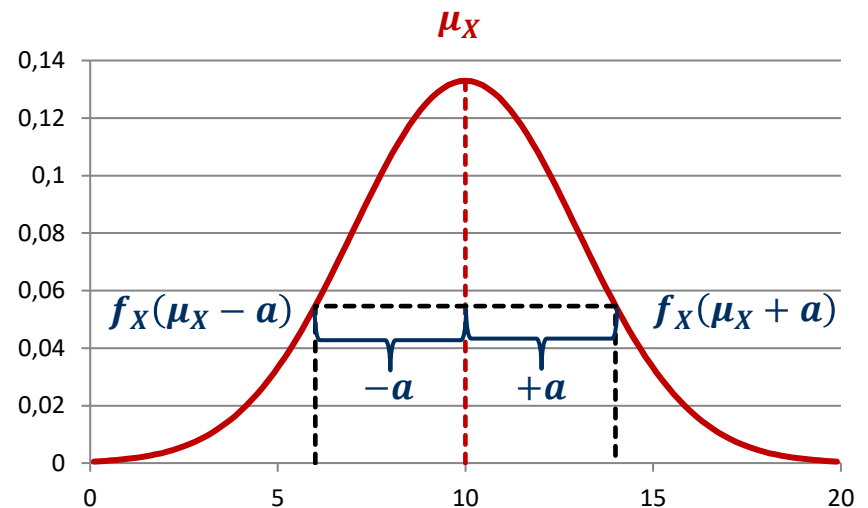
$$f_X(\mu_X - a) = f_X(\mu_X + a)$$

$$P(X > \mu_X + a) = P(X < \mu_X - a)$$

$$P(X < \mu_X + a) = P(X > \mu_X - a)$$

$$P(X < \mu_X) = P(X > \mu_X) = 0,5$$

Median: $Med(X) = \mu_X$ = the point where the left side has probability 0,5
and the right side has probability 0,5



Linear combination of normal dist.

- If X is normally dist., then $Y = a + bX$ is **normally distributed**

$$\mu_Y = E(Y) = E(a + bX) = a + b\mu_X$$

$$\sigma_Y^2 = \text{Var}(Y) = \text{Var}(a + bX) = \mathbf{b^2} \sigma_X^2$$

$$\begin{aligned} X &\sim N(\mu_X, \sigma_X^2) \\ \Rightarrow Y &\sim N(a + b\mu_X, b^2 \sigma_X^2) \end{aligned}$$

Standardized normal distribution

- Special case of linear combination

$$\mu_Z = -\frac{\mu_X}{\sqrt{\sigma_X^2}} + \frac{1}{\sqrt{\sigma_X^2}} \cdot \mu_X = 0$$

$$\sigma_Z^2 = \mathbf{b^2} \sigma_X^2 = \frac{1}{\sigma_X^2} \cdot \sigma_X^2 = 1$$

$$Z = \frac{X - \mu_X}{\sigma_X} = -\frac{\mu_X}{\sigma_X} + \frac{1}{\sigma_X} \cdot X$$

$$X \sim N(\mu_X, \sigma_X^2) \Rightarrow Z \sim N(0, 1)$$



Calculations of probabilities

- **Cannot be done by hand!**
- There does not exist a simple formula for $F_X(x)$
- Advanced numerical methods are required.
- **Alternatives:** tables or computers
- Swedish Excel:

$$F_X(x) = \int_{-\infty}^x f_X(t) dt$$

$f_X(t)$ write "=NORM.FÖRD(x;mu;sigma;0)"

$F_X(x)$ write "=NORM.FÖRD(x;mu;sigma;1)"

Substitute numbers in place of x , μ and σ (NOTE! not σ squared!)

English language Excel: NORMDIST(x,mu,sigma,0)

Step 1: Standardize

- Standardize X to Z
- If $X \sim N(\mu_X, \sigma_X^2)$ then $Z = \frac{X - \mu_X}{\sigma_X} \sim N(0, 1)$
- If you want the probability $P(X \leq x)$, then use:

$$P(X \leq x) = P\left(Z \leq \frac{x - \mu_X}{\sigma_X}\right)$$

Write Z instead of X and exchange x for $z = \frac{x - \mu_X}{\sigma_X}$

Example

$X \sim N(10, 9)$ **NOTE!** $\sigma_X^2 = 9$ so $\sigma_X = 3$

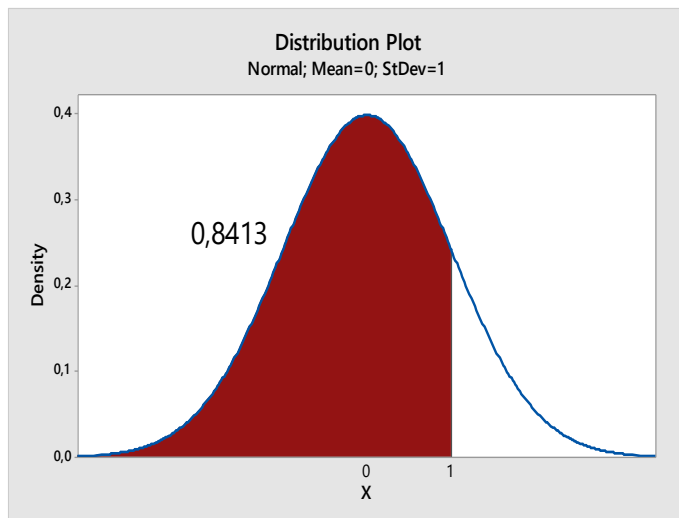
- $P(X \leq 13) = P\left(Z \leq \frac{13-10}{3}\right) = P\left(Z \leq \frac{3}{3}\right) = P(Z \leq 1)$
- $P(X \leq 7) = P\left(Z \leq \frac{7-10}{3}\right) = P\left(Z \leq -\frac{3}{3}\right) = P(Z \leq -1)$
- $P(4 \leq X \leq 13) = P\left(\frac{4-10}{3} \leq Z \leq \frac{13-10}{3}\right) = P\left(-\frac{6}{3} \leq Z \leq \frac{3}{3}\right) = P(-2 \leq Z \leq 1)$
$$= P(Z \leq 1) - P(Z \leq -2)$$

$$= F_X(1) - F_X(-2)$$

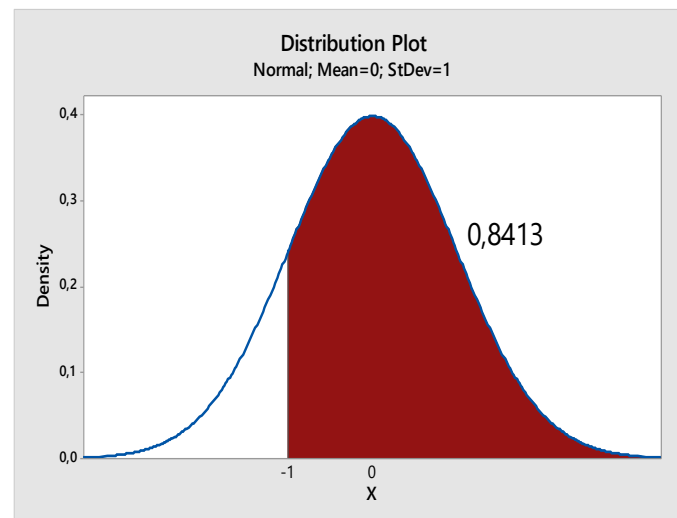
Step 2: Use the symmetry around $\mu_Z = 0$

- Of what form is the probability we seek?

$$P(Z \leq 1,00) = [\text{table}] = 0,8413$$



$$P(Z \geq -1,00) = P(Z \leq 1,00)$$



- You want it on the form:

$$P(Z \leq z) = F_Z(z)$$

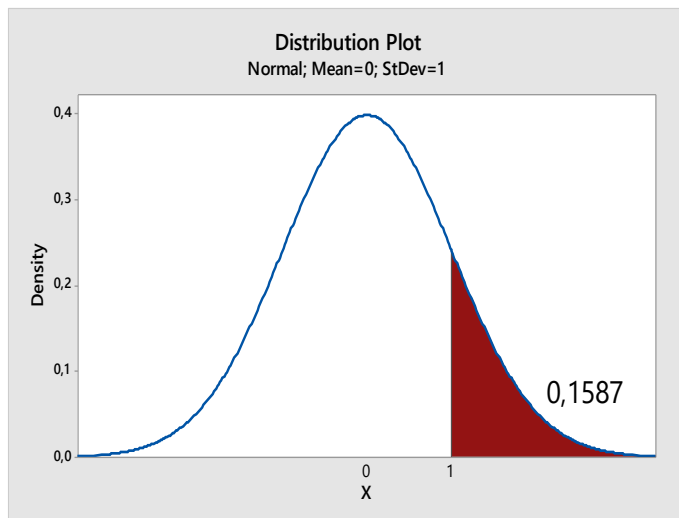
where z is positive



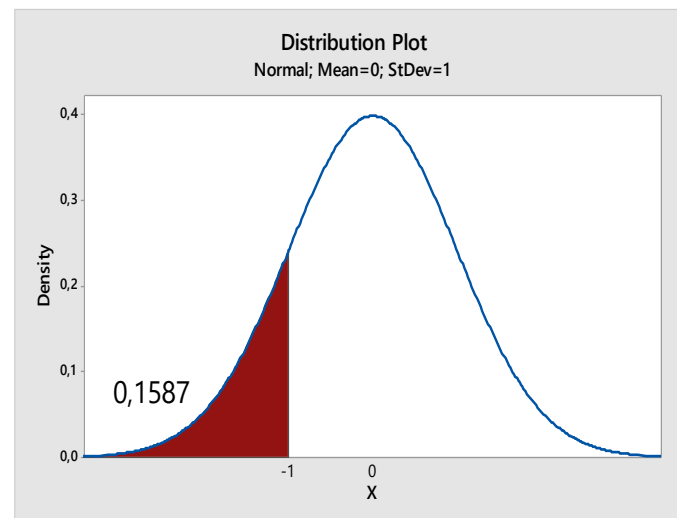
Steg 2: Use the symmetry around μ_Z , cont.

- Of what form is the probability we seek?

$$P(Z > 1,00) = 1 - P(Z \leq 1,00)$$



$$P(Z < -1,00) = P(Z > 1,00) \\ = 1 - P(Z \leq 1,00)$$



- You want the form to be:

$$P(Z \leq z) = F_Z(z)$$

where z is positive



Example

$X \sim N(10, 9)$ **NOTICE!** $\sigma_X^2 = 9$ so $\sigma_X = 3$

- $P(X \leq 13) = P(Z \leq 1) = F_x(\textcircled{1})$ **positive number, 1**

positive number, 1

- $P(X \leq 7) = P(Z \leq -1) = P(Z \geq 1) = 1 - P(Z \leq 1) = 1 - F_x(\textcircled{1})$

- $P(4 \leq X \leq 13) = P(-2 \leq Z \leq 1) = P(Z \leq 1) - P(Z \leq -2)$

$$= P(Z \leq 1) - [1 - P(Z \leq 2)] = P(Z \leq 1) + P(Z \leq 2) - 1$$

$$= F_x(\textcircled{1}) + F_x(\textcircled{2}) - 1$$

positive numbers, 1 and 2



Step 3: Use the table

- The table shows values for the **standardized** normal distribution with expected value = 0 och variance = 1;

$$Z \sim N(0, 1)$$

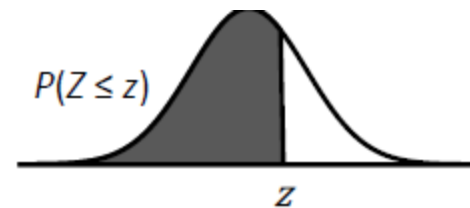
- **Standardize**

- The table gives cumulative probability $P(Z \leq z) = F_Z(z)$
 - **Use symmetry and reformulate the probabilities so that they are in the form $P(Z \leq z)$**
- And only positive values of z , i.e. for $z \geq 0$
 - **Positive values of z**



TABELL 1. Normalfördelningen, standardiserad

$\Phi(z) = P(Z \leq z)$ där $Z \in N(0, 1)$.



För negativa värden, utnyttja att $\Phi(-z) = 1 - \Phi(z)$.

The single digit place + the first decimal of z

The second decimal of z

z	0,00	0,01	0,02	0,03	0,04	0,05	0,06	0,07	0,08	0,09
0,0	0,50000	0,50399	0,50798	0,51197	0,51595	0,51994	0,52391	0,52786	0,53179	0,53576
0,1	0,53983	0,54380	0,54776	0,55172	0,55567	0,55962	0,56357	0,56750	0,57143	0,57536
0,2	0,57926	0,58317	0,58706	0,59095	0,59483	0,59871	0,60257	0,60642	0,61026	0,61409
0,3	0,61791	0,62172	0,62552	0,62930	0,63307	0,63683	0,64058	0,64431	0,64803	0,65173
0,4	0,65542	0,65910	0,66276	0,66640	0,67003	0,67364	0,67724	0,68082	0,68439	0,68793
0,5	0,69146	0,69497	0,69847	0,70194	0,70540	0,70884	0,71226	0,71566	0,71904	0,72240
0,6	0,72575	0,72907	0,73237	0,73565	0,73891	0,74215	0,74537	0,74857	0,75175	0,75490
0,7	0,75804	0,76115	0,76424	0,76730	0,77035	0,77337	0,77637	0,77935	0,78230	0,78524
0,8	0,78814	0,79103	0,79389	0,79673	0,79955	0,80234	0,80511	0,80785	0,81057	0,81327
0,9	0,81594	0,81859	0,82121	0,82381	0,82639	0,82894	0,83147	0,83398	0,83646	0,83891
1,0	0,84134	0,84375	0,84614	0,84849	0,85082	0,85312	0,85540	0,85769	0,85993	0,86214
1,1	0,86433	0,86650	0,86864	0,87076	0,87286	0,87493	0,87698	0,87900	0,88100	0,88298
1,2	0,88493	0,88686	0,88877	0,89065	0,89251	0,89435	0,89617	0,89796	0,89973	0,90147
1,3	0,90320	0,90490	0,90658	0,90824	0,90988	0,91149	0,91309	0,91466	0,91621	0,91774
1,4	0,91924	0,92073	0,92220	0,92364	0,92507	0,92647	0,92785	0,92922	0,93056	0,93189
1,5	0,93319	0,93448	0,93574	0,93699	0,93822	0,93943	0,94062	0,94179	0,94295	0,94408
1,6	0,94520	0,94630	0,94738	0,94845	0,94950	0,95053	0,95154	0,95254	0,95352	0,95449
1,7	0,95543	0,95637	0,95728	0,95818	0,95907	0,95994	0,96080	0,96164	0,96246	0,96327
1,8	0,96407	0,96485	0,96562	0,96638	0,96712	0,96784	0,96856	0,96926	0,96995	0,97062
1,9	0,97128	0,97193	0,97257	0,97320	0,97381	0,97441	0,97500	0,97558	0,97615	0,97670
2,0	0,97725	0,97778	0,97831	0,97882	0,97932	0,97982	0,98030	0,98077	0,98124	0,98169
2,1	0,98214	0,98257	0,98300	0,98341	0,98382	0,98422	0,98461	0,98500	0,98537	0,98574

i.e. $P(Z \leq 0,93)$

Exercise

The employees of an airline that traffics New York (JFK) note that the planes sometimes arrive a little early and sometimes a little late. Let X = “difference between actual arrival time and planned arrival time” and suppose that X is normally distributed with expectation 12 minutes and standard deviation 8,6 minutes.

1. What is the probability that a flight arrives more than 30 minutes late, i.e. what is $P(X > 30)$?
2. What is the probability that a flights arrives early, i.e. $P(X < 0)$?
3. What is the probability that a flight is exactly 12 minutes late?



Solution

1. Sought: $P(X > 30) =$

$$= [\text{standardize}] = P\left(Z > \frac{30 - 12}{8,6}\right) \approx P(Z > 2,09)$$

$$= [\text{substitute forms! draw!}] = 1 - P(Z < 2,09) = 1 - F_Z(2,09)$$

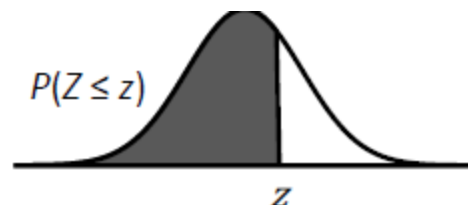
$$= [\text{use the table}] = 1 - 0,98169 = 0,01831$$

Answer: approximately 1,8 % probability

TABELL 1. Normalfördelningen, standardiserad

$\Phi(z) = P(Z \leq z)$ där $Z \in N(0, 1)$.

För negativa värden, utnyttja att $\Phi(-z) = 1 - \Phi(z)$.



z	0,00	0,01	0,02	0,03	0,04	0,05	0,06	0,07	0,08	0,09
0,0	0,50000	0,50399	0,50798	0,51197	0,51595	0,51994	0,52392	0,52790	0,53188	0,53586
0,1	0,53983	0,54380	0,54776	0,55172	0,55567	0,55962	0,56356	0,56749	0,57142	0,57535
0,2	0,57926	0,58317	0,58706	0,59095	0,59483	0,59871	0,60257	0,60642	0,61026	0,61409
0,3	0,61791	0,62172	0,62552	0,62930	0,63307	0,63683	0,64058	0,64431	0,64803	0,65173
0,4	0,65542	0,65910	0,66276	0,66640	0,67003	0,67364	0,67724	0,68082	0,68439	0,68793
0,5	0,69146	0,69497	0,69847	0,70194	0,70540	0,70884	0,71226	0,71566	0,71904	0,72240
0,6	0,72575	0,72907	0,73237	0,73565	0,73891	0,74215	0,74537	0,74857	0,75175	0,75490
0,7	0,75804	0,76115	0,76424	0,76730	0,77035	0,77337	0,77637	0,77935	0,78230	0,78524
0,8	0,78814	0,79103	0,79389	0,79673	0,79955	0,80234	0,80511	0,80785	0,81057	0,81327
0,9	0,81594	0,81859	0,82121	0,82381	0,82639	0,82894	0,83147	0,83398	0,83646	0,83891
1,0	0,84134	0,84375	0,84614	0,84849	0,85083	0,85314	0,85543	0,85769	0,85993	0,86214
1,1	0,86433	0,86650	0,86864	0,87076	0,87286	0,87493	0,87698	0,87900	0,88100	0,88298
1,2	0,88493	0,88686	0,88877	0,89065	0,89251	0,89435	0,89617	0,89796	0,89973	0,90147
1,3	0,90320	0,90490	0,90658	0,90824	0,90988	0,91149	0,91309	0,91466	0,91621	0,91774
1,4	0,91924	0,92073	0,92220	0,92364	0,92507	0,92647	0,92785	0,92922	0,93056	0,93189
1,5	0,93319	0,93448	0,93574	0,93699	0,93822	0,93943	0,94062	0,94179	0,94295	0,94408
1,6	0,94520	0,94630	0,94738	0,94845	0,94950	0,95053	0,95154	0,95254	0,95352	0,95449
1,7	0,95543	0,95637	0,95728	0,95818	0,95907	0,95994	0,96080	0,96164	0,96246	0,96327
1,8	0,96407	0,96485	0,96562	0,96638	0,96712	0,96784	0,96856	0,96926	0,96995	0,97062
1,9	0,97128	0,97193	0,97257	0,97320	0,97381	0,97441	0,97500	0,97558	0,97615	0,97670
2,0	0,97725	0,97778	0,97831	0,97882	0,97932	0,97982	0,98030	0,98077	0,98124	0,98169
2,1	0,98214	0,98257	0,98300	0,98341	0,98382	0,98422	0,98461	0,98500	0,98537	0,98574

Solution

2. Sought: $P(X < 0) =$

$$= [\text{standardize}] = P\left(Z < \frac{0 - 12}{8,6}\right) \approx P(Z < -1,40)$$

$$= [\text{substitute forms! draw!}] = P(Z > 1,40) = 1 - P(Z < 1,40)$$

$$= 1 - F_Z(1,40)$$

$$= [\text{use the table}] = 1 - 0,91924 = 0,08076$$

Answer: approximately 8,1 % chance

3. Sought: $P(X = 12)$

= 0 (Remember: the area of a line is zero)



Summary

- Continuous random variables
 - can take any value in an interval S_X
 - S_X can be bounded or unbounded ($\pm\infty$)
 - cumulative distribution functions, $F_X(x) = P(X \leq x)$
 - events = subintervals of S_X
 - The values of the distribution function $f_X(x)$ are not probabilities.
 - Probability = the area under $f_X(x)$ within the subintervals
- Normal distribution, $X \sim N(\mu_X, \sigma_X^2)$
 - standardized normal distribution $Z \sim N(0, 1)$
 - calculation of probabilities using the table

Next time

Two or more random variables at once

- Bivariate probability distributions
- Conditioning
- Covariance and correlation
- We save section 5.4 for L8
 - approximate a binomial with a normal distribution