EGS 2405- Geostatistics



Basic Statistics and Exploratory Data Analysis (EDA)

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Introduction



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Recall from our previous class

Most **consequential decisions** of any **geostatistical** study
are made early in the exploratory data analysis (EDA) – Which
is our today's topic

Introduction



Survey/research data is structured as;

- Rectangular array (e.g., spreadsheet or database)
- One row/experimental subject and one column for each subject identifier
- Outcome variable, and explanatory variable.

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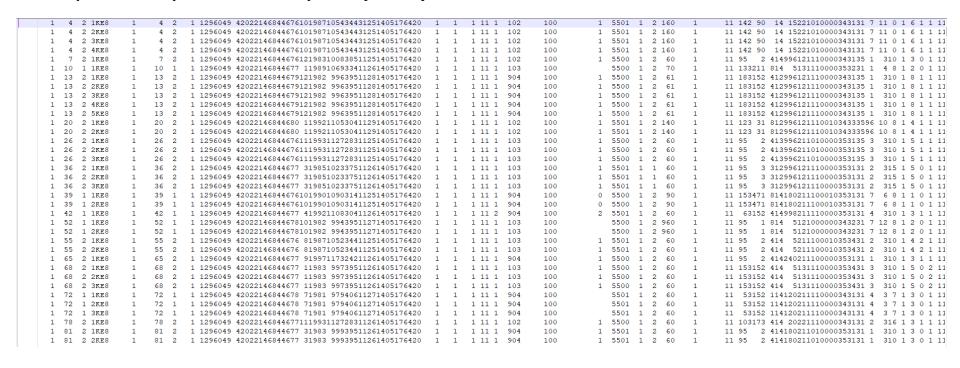
DHS data 2022

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Introduction



- Examining the spread sheet by eye is a tedious, boring and overwhelming process
- Exploratory data analysis (EDA) can be used to aid the situation



What is EDA

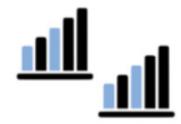


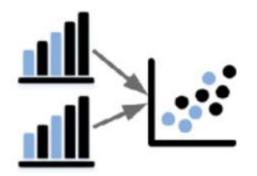
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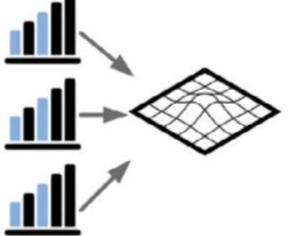
 Approach of analyzing data sets to summarize their main characteristics, often with visual methods.



 EDA helps to gain insight into a dataset before doing any formal statistical modelling/hypothesis testing









Purpose of EDA

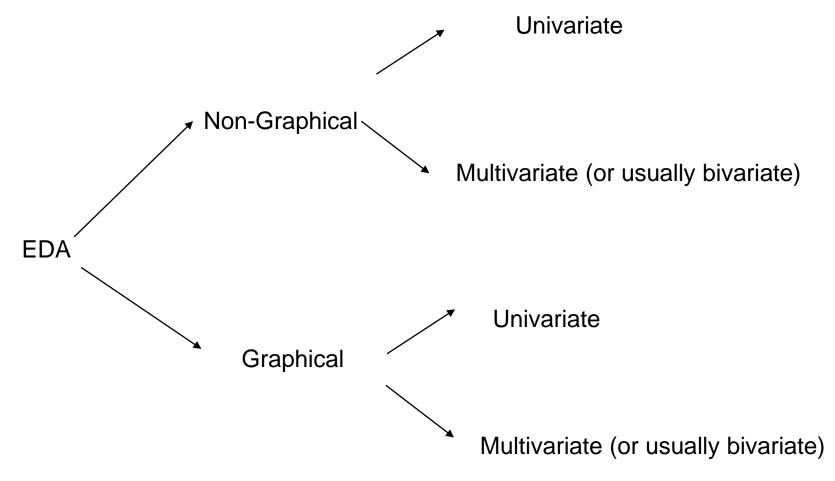


- Uncovers the underlying structure of the dataset
- Identifies important variables
- Detects outliers and anomalies
- Tests underlying assumptions
- Determines relevant variables, their transformations, and interaction among variables with respect to the model to be built.
- Highlights missing data as may be relevant to building desired models.

Classifications of EDA



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Univariate non-graphical EDA helps to:

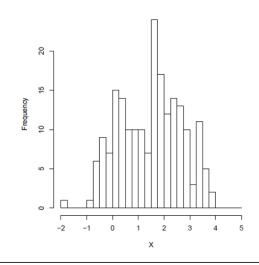
- Better appreciate the "sample distribution"
- Make some tentative conclusions about what population distribution(s) is/are compatible with the sample distribution

Categorical data

Statistic/College	H&SS	MCS	SCS	other	Total
Count	5	6	4	5	20
Proportion	0.25	0.30	0.20	0.25	1.00
Percent	25%	30%	20%	25%	100%

Count, proportion, percentage of students taking different subjects

Quantitative data

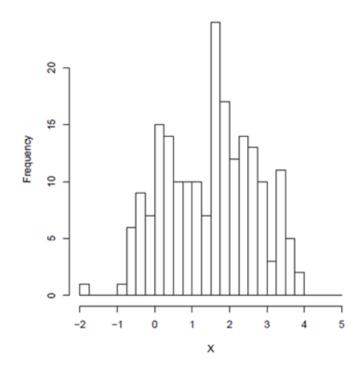


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Univariate Non-graphical EDA

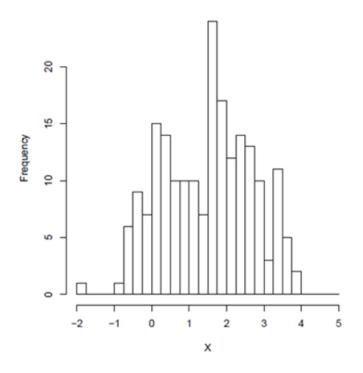
- Sample's distributional characteristics are seen qualitatively in the univariate graphical EDA technique of a histogram
- Think of univariate non-graphical EDA as telling you about aspects of the histogram of the distribution of the variable of interest for example SOC





Univariate Non-graphical EDA

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The central tendency

- Measures of central tendency provide information about where the center of a distribution is located.
- The most commonly used measures of center for numerical data are the
 - ❖ Mean
 - Median
 - ❖ Mode



- The mean is the simple arithmetic average:
- The sum of the values of a variable divided by the number of observations (n)

$$\bar{x} = \frac{\sum_{i=1}^{n} x_i}{n}$$

where

- n is the total number of observations
- x_i is the score of the ith observation
- Σ is the symbol of summation (pronounced sigma)
- \bar{x} is the sample mean value



- The mean is the simple arithmetic average:
- The sum of the values of a variable divided by the number of observations (n)

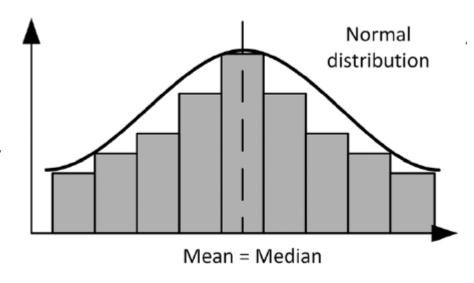
$$\bar{x} = \frac{\sum_{i=1}^{n} x_i}{n}$$
 Equation (2.1)

where

- n is the total number of observations
- x_i is the score of the ith observation
- Σ is the symbol of summation (pronounced sigma)
- \bar{x} is the sample mean value



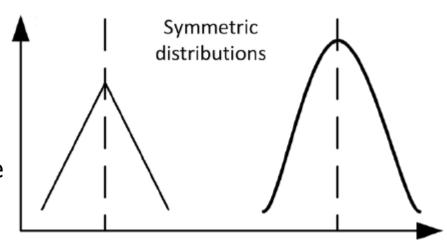
- The median is the value that divides the sorted scores form smaller to larger in half. It is a measure of center.
- Relevant for interval/ratio and ordinal
- The median overcomes the outlier problem
- When n is odd; single median.
 When n is even, there are two
 "middle values,". Average is taken



The median is located in the center of a normal distribution and coincides with the mean



- The mode is the most typical value.
- It implies that the frequency distribution has a single
- peak.
- For a symmetric distribution the mode, the mean and the median are in principle the same.
- For an asymmetric one (mode - median) ~ 2 * (median mode)



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- Measures of spread/variability/variation/diversity or dispersion provide information on how much the values of a variable differ among themselves and in relation to the mean.
- The most common measures are as follows (de Smith 2018):
 - ❖ Range
 - Deviation from the mean
 - Variance
 - Standard deviation
 - Standard distance
 - Percentiles and quartiles

Example with Meuse data



```
install.packages("sp")
library(sp)
data(meuse)
summary(meuse)
head(meuse)
dim(meuse) #To check the data dimension (RowsxColumns)
str(meuse) #To check the structure of the data (Continuous vs categorical)
mean(meuse$cadmium)
median(meuse$copper)
# define mode() function
mode = function() {
 # calculate mode of the copper concentration
 return(names(sort(-table(meuse$copper)))[1])
# call mode() function
mode()
```



Measures of dispersion

 A range is the difference between the largest and smallest values of the variable studied.

Range =
$$xmax-xmin$$
 Equation (2.2)

- **Deviation** from the mean is the subtraction of the mean from each score, **Deviation**= $(x_i \bar{x})$ Equation (2.3)
- The sum of all deviations is zero (sometimes, due to rounding up, the sum is very close to zero)



Measures of dispersion

Variance and standard deviation

The variance of a set of values, which we denote S², is by definition

$$S^2 = \frac{1}{N} \sum_{i=1}^{N} (z_i - \bar{z})^2$$
 Equation (2.4)

- The variance is the second moment about the mean.
- Like the mean, it is based on all of the observations, it can be treated algebraically, and it is little affected by sampling fluctuations.
- It is both additive and positive.
- Its square root is the standard deviation, S.
- To estimate population variance σ^2 from a sample, then N in Equation (2.4) is replaced by N-1.



Measures of dispersion

Coefficient of variation

- The SD is regarded in relative terms (relative variability).
- The CV is particularly useful when you want to compare results from two different surveys or tests that have different measures or values (hence different means)
 - CV = $(SD/\bar{x}) * 100$

Equation (2.5)



Measures of dispersion in R with Meuse data

```
var(meuse$copper)
sd(meuse$copper)

#anotherway of computing the SD
sqrt(var(meuse$copper))

#Coefficient of variance for copper conc in the region
((sd(meuse$copper))/(mean(meuse$copper)))*100
```



Measures of shape describe how values (e.g., frequencies) are distributed across the intervals (bins) and are measured by **skewness** and **kurtosis**.

Skewness

The skewness measures the asymmetry of the observations. It is defined formally from the third moment about the mean:

$$m_3 = rac{1}{N} \sum_{i=1}^{N} (z_i - ar{z})^3$$
 Equation (2.6)



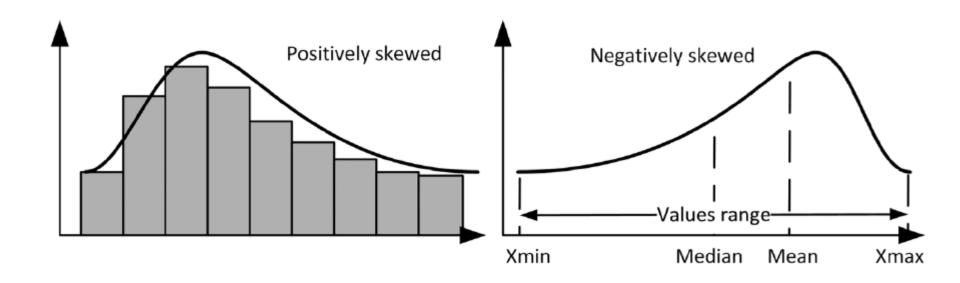
The coefficient of skewness is then

$$g_1 = \frac{m_3}{m_2\sqrt{m_2}} = \frac{m_3}{S^3}$$

Equation (2.7)

- where m2 is the variance and S the standard deviation. Symmetric distributions have g1 = 0. N
- Negative skewness indicates that the mean of the data values is less than the median, and the data distribution is left-skewed.
- Positive skewness would indicate that the mean of the data values is larger than the median, and the data distribution is right-skewed





Positively and Negatively skewed distributions



Kurtosis is obtained from the fourth moment about the mean

$$m_4 = \frac{1}{N} \sum_{i=1}^{N} (z_i - \bar{z})^4$$
 Equation (2.8)

The coefficient of Kurtosis is given by

$$g_2 = \frac{m_4}{m_2^2} - 3 = \frac{m_4}{(S^2)^2} - 3$$
 Equation (2.9)



- Kurtosis, from the graphical inspection perspective, is the degree of the peakedness or flatness of a distribution.
- A zero kurtosis (g=0) indicates a near-normal distribution peakedness.
- A negative kurtosis indicates a more flat distribution (lower than normal)
 (g<0)
- A positive kurtosis reveals a distribution with a higher peak than the normal distribution (g>0)

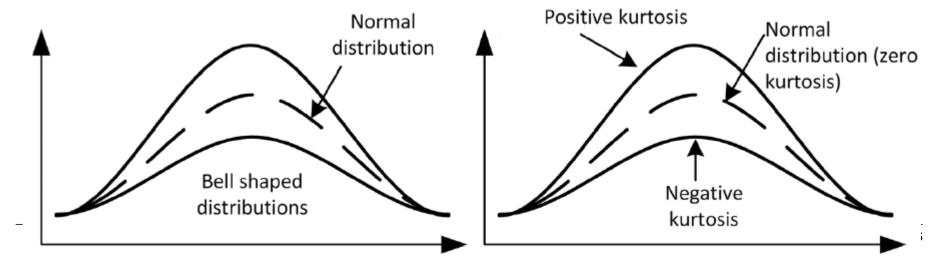
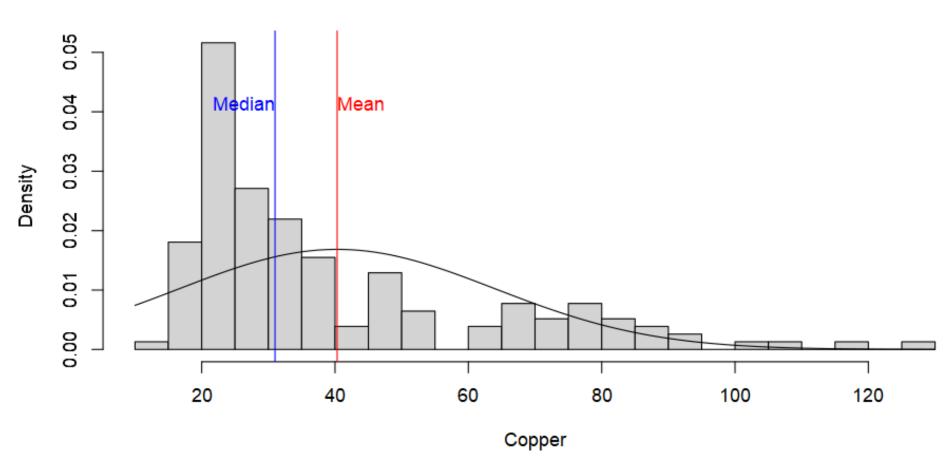


Illustration of dispersion in R with Meuse data



Normal distribution curve over histogram



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Measures of shape in R with Meuse data



```
#Measures of shape
library(pacman)
pacman::p_load(e1071)
kurtosis(meuse$copper)
hist(meuse$copper, prob=TRUE,breaks=20,main="Normal distribution
curve over histogram", xlab= "Copper")
pacman::p_load(e1071)
pacman::p_load(e1071)
curve(dnorm(x, mean=mean(meuse$copper), sd=sd(meuse$copper)),
add=TRUE)
abline(v=mean(meuse$copper), col="red")
text(mean(meuse$copper),0.04,"Mean", col = "red", adj = c(0, -.1))
abline(v=median(meuse$copper), col="blue")
text(median(meuse$copper),0.04,"Median", col = "blue", adj = c(1, -.1))
```



Univariate graphical EDA

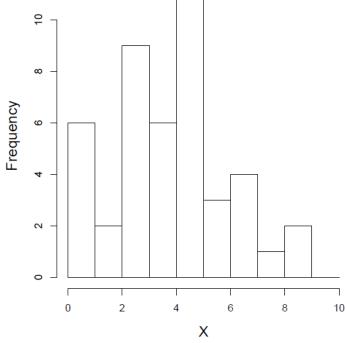
It involves visualizing graphically at the distribution of the sample

Graphical methods are more qualitative and involve a degree of

subjective analysis.

Histograms

The only one of these techniques
 that makes sense for categorical
 data is the histogram (basically
 just a bar plot of the tabulation of
 the data).



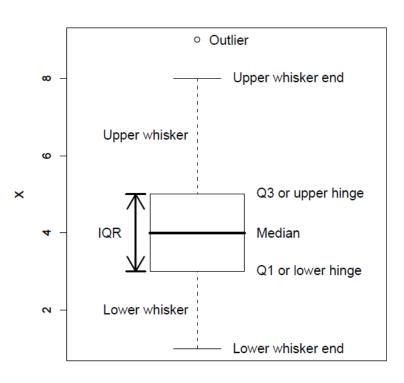
 Histograms are one of the best ways to quickly learn a lot about your data, including central tendency, spread, modality, shape and outliers.



Univariate graphical EDA

Box-plots

- Boxplots are very good at presenting information about the central tendency, symmetry and skew, as well as outliers
- They can be misleading about aspects such as multimodality. One of the best uses of boxplots is in the form of sideby-side boxplots



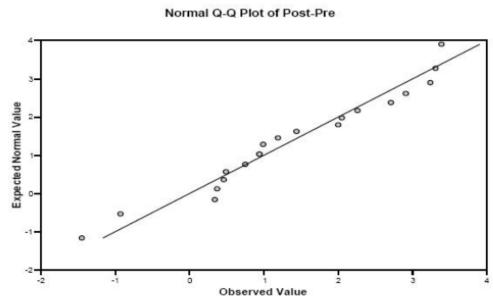
 Boxplots show robust measures of location and spread as well as providing information about symmetry and outliers



Univariate graphical EDA

Quantile-normal plots

- QN plot or more generality the or QQ plot.
- Used to see how well a sample
 of data of size n matches a
 Gaussian distribution with
 mean and variance equal to the
 sample mean and variance.
- The quantile-normal plot we can detect left or right skew, positive or negative kurtosis, and bimodality.



Quantile-Normal plots allow detection of non-normality and diagnosis of skewness and kurtosis.



Multivariate non-graphical EDA

 Multivariate non-graphical EDA techniques generally show the relationship between two or more variables in the form of either crosstabulation or statistics.

Subject ID	Age Group	Sex
GW	young	F
JA	middle	F
TJ	young	M
JMA	young	M
JMO	middle	F
JQA	old	F
AJ	old	F
MVB	young	M
WHH	old	F
JT	young	F
JKP	middle	M

Cross-tabulation for sample data

Age Group / Sex	Female	Male	Total
young	2	3	5
middle	2	1	3
old	3	0	3
Total	7	4	11

Sample Data for Cross-tabulation



Multivariate non-graphical EDA

Covariance and correlation

- The joint dispersion of two variables, z1 and z2, is termed as covariance C1;2.
- Covariance for a finite set of observations can be expressed as:

$$C_{1,2} = \frac{1}{N} \sum_{i=1}^{N} \{ (z_1 - \bar{z_1})(z_2 - \bar{z_2}) \}$$
 Equation (2.10)

• where $\overline{z_1}$ and $\overline{z_2}$ are the means of the two variables. This expression is analogous to the variance of a finite set of observations.



Covariance and correlation

- Covariance is affected by the scales on which the properties have been measured.
- This makes comparisons between different pairs of variables and sets of observations difficult unless measurements are on the same scale
- Pearson product-moment correlation coefficient, or simply the correlation coefficient, is often preferred. It refers specifically to linear correlation and it is a dimensionless value



Multivariate non-graphical EDA

The correlation coefficient, is given as.

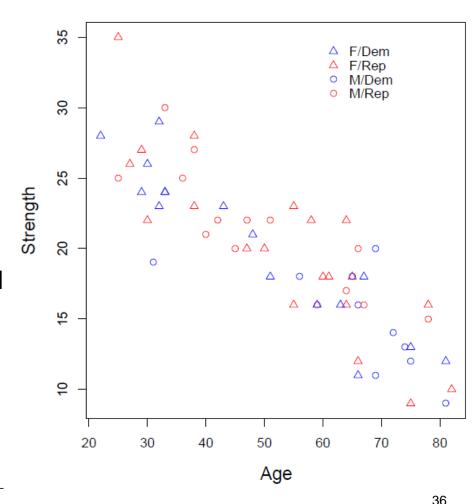
$$\rho = \frac{C_{1,2}}{S_1 S_2}$$
 Equation (2.11)

- CC ranges between 1 and -1.
- If units with large values of one variable also have large values of the other then the two variables are positively correlated, ρ>0;
- if the large values of the one are matched by small values of the other then the two are **negatively correlated**, ρ < 0.
- If $\rho=0$ then there is **no linear relation**.



Multivariate graphical EDA

- For two quantitative variables, the basic graphical EDA technique is the scatterplot
- It has one variable on the x-axis, and one on the y-axis and a point for each case of the dataset.
- If one variable is explanatory and the other is outcome, it is a very, very strong convention to put the outcome on the y (vertical) axis



BEGIS 3



- Normal distribution (also called Gaussian distribution) is the most commonly used distribution in statistics, as many physical phenomena are normally distributed (e.g., human weight and height).
- In a normal distribution, the values of a variable are more likely to be
 closer to the mean, while larger or smaller scores have low probabilities
 of occurring
- Many environmental variables, such as of the soil, are distributed in a way that approximates the normal distribution



 Normal distribution is defined for a continuous random variable X in terms of the probability density function (pdf), f(x), as

$$f(x|\mu,\sigma) = \frac{1}{\sigma\sqrt{2\pi}}e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2}$$
 Equation (2.12)

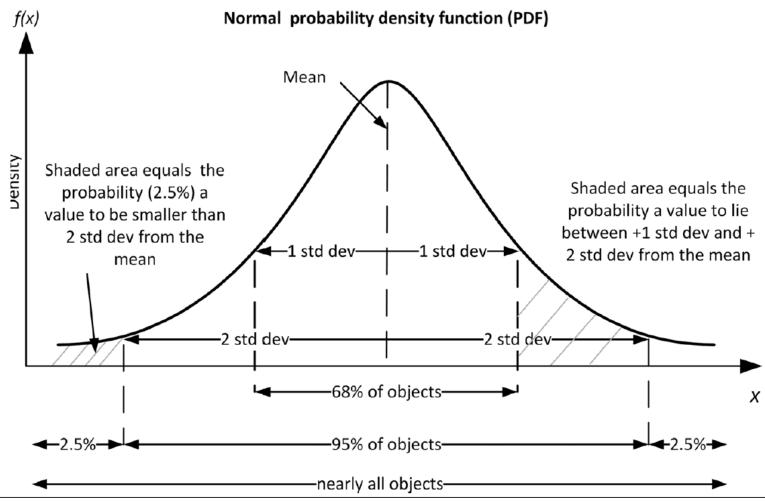
where

 μ is the mean of the population σ is the population standard deviation σ^2 is the population variance σ is the value of the variable



- Normal distribution is continuous and symmetrical, with its peak at the mean of the distribution
- The ordinate f(x) at any given value of x is the probability density at x
- The total area under the curve is 1, the total probability of the distribution.
- The area under any portion of the curve, say between z1 and z2,
 represents the proportion of the distribution lying in that range



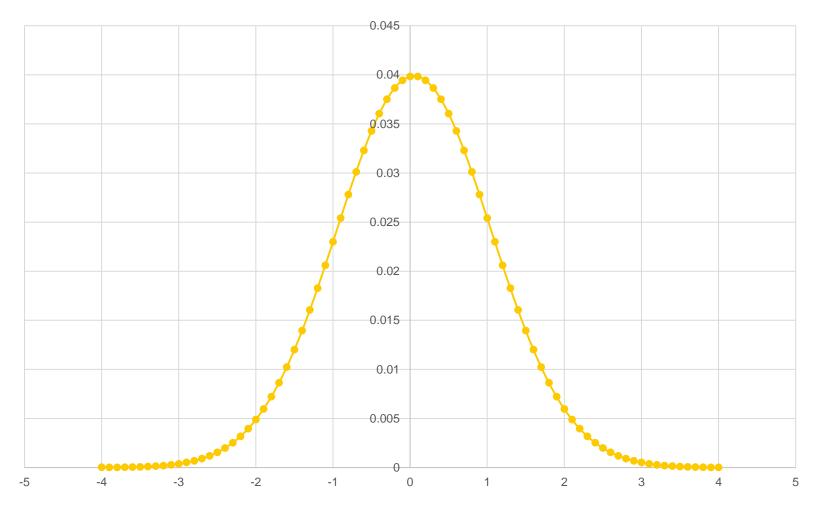




- Slightly more than two-thirds of the distribution lies within one standard deviation of the mean, i.e. between μ - σ AND μ + σ ;
- About 95% lies in the range μ -2 σ AND μ + 2 σ ;
- About 99.73% lies within three standard deviations of the mean μ 3σ AND μ + 3σ ;

The normal distribution- Illustration using excel

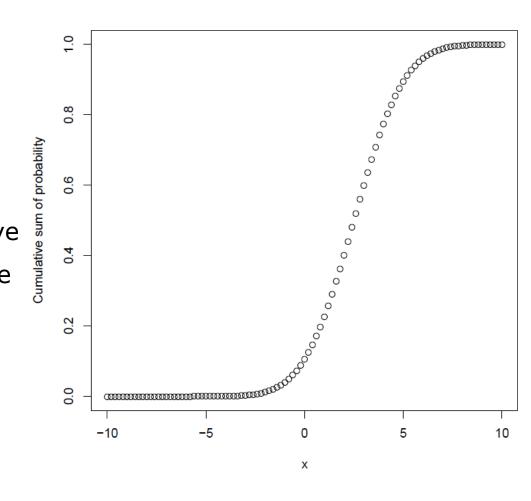




The Cumulative distribution function



- Pdf can be represented as a cumulative distribution
- In this representation the normal distribution is characteristically sigmoid
- The main use of the cumulative distribution function is that the probability of a values being less than a specified amount can be deduced



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Data transformation



- To overcome the difficulties arising from departures from normality observations can be transformed to a new scale on which the distribution is more nearly normal.
- Further analysis can be done on the transformed data, and if necessary transform the results to the original scale at the end.

Logarithmic transformation



The geometric mean of a dataset is given as

$$\bar{g} = \left\{\prod_{i=1}^{N} z_i\right\}^{\frac{1}{N}}, \quad \text{Equation (2.13)}$$

Therefore, the log transformation is

$$\log \bar{g} = \frac{1}{N} \sum_{i=1}^{N} \log z_i, \text{ Equation (2.14)}$$

• The logarithm may be either natural (ln) or common (log10). If by transforming the data Z_i = 1,2, 3......N, we obtain log z with a normal distribution then the variable is said to be **lognormally distributed**.

Square root transformation



- Taking logarithms will often normalize, or at least make symmetric, distributions that are strongly positively skewed, i.e. have g1 > 1.
- Less pronounced positive skewness can be removed by taking square roots:

$$r=\sqrt{z}$$
.

Equation (2.15)

Angular transformation



- This is sometimes used for proportions in the range 0 to 1, or 0 to 100 if expressed as percentages.
- If p is the proportion then define

$$\phi = \sin^{-1} \sqrt{p}.$$

Equation (2.16

• The desired transform is the angle whose sine is \sqrt{p}

Logit transformation



If, as above, p is a proportion (0 , then its logit is

$$l = \ln \frac{p}{1 - p}$$
 Equation (2.17)

- Note that the limits 0 and 1 are excluded; otherwise I would either go to -∞ or +∞.
- If you have proportions that include 0 or 1 then you must make some little adjustment to use the logit transformation.

Thank you for your attention! Questions?



