

# Stats1 Definitions and Variables

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# Data Management and Descriptive Statistics

## 1. Topics for masters statistics B\_block

- 1.1 Frequency Distribution Presentation, & Central Tendency I
- 1.2 Measures of Dispersion
- 1.3 Data Distribution and Variance
- 1.4 Hypothesis Testing Unit of Analysis
- 1.5 Basic Probability, Inference
- 1.6 Significance Tests

## 1. Topics for masters statistics II B\_block

- 1.1 Chi Square, Expected values & Mean Testing Continued
- 1.2 Analysis of Variance (ANOVA)
- 1.3 Associations Nominal and Ordinal Data, Bivariate Correlation,
- 1.4 Pearson's  $r$  and Spearman's  $Rho$
- 1.5 Bivariate Regression
- 1.6 Multiple Regression

## 1. Variables B\_definition

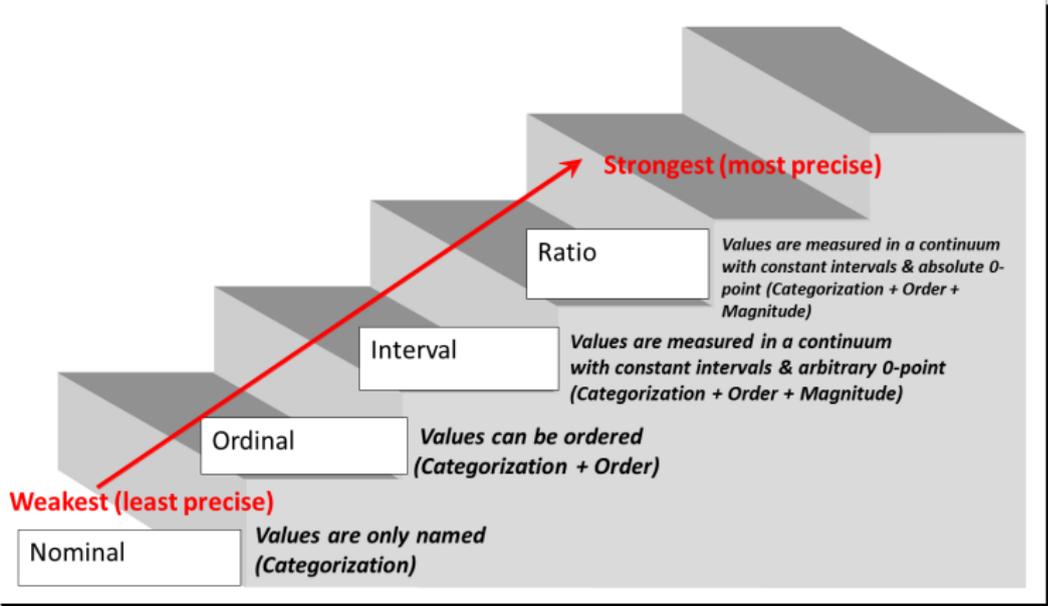
- ▶ **Variables** are characteristics that change from person to person or object to object in a population of interest
  - ▶ Variables can take different values or levels

- ▶ They are called variables because they vary between cases
- ▶ Each variable has a level of measurement.

## 1. Levels of Measurement

B\_block:BMCOL

- ▶ Each variable has a level of measurement.
- ▶ The level of measurement is important because it determines the type of analysis.
- ▶ There are four levels of measurement



## 1. Categorical, Nominal or Qualitative data

B\_block

- ▶ The word nominal means names
- ▶ A nominal variable ONLY describes something
- ▶ The only function is to label and categorize

- ▶ NO INHERENT NUMERIC QUANTITY, NO Ranking of levels or ordering scheme
  - ▶ Numbers can be nominal level if there is no quantity associated with them.
  - ▶ Categories should be distinct, mutually exclusive, and completely exhaustive

## 1. Examples

B\_block

- ▶ Sex
- ▶ Religious Affiliation
- ▶ Race
- ▶ Zip code
- ▶ State

## 1. Ordinal Variables

B\_definition

- ▶ Ordinal variables have a ranking
- ▶ The root of Ordinal is “Ord” for **order**
- ▶ The exact amount of difference is unknown
- ▶ rank in most wanted

## 1. Examples

B\_block

- ▶ rating at a restaurant

- ▶ rank in the police department
- ▶ letter grade
- ▶ Service ratings
  - ▶ 1=Poor
  - ▶ 2=Fair
  - ▶ 3=Good
  - ▶ 4=Excellent

## 1. Interval variable

B\_definition

- ▶ *Interval Variables* have inherently numeric values
- ▶ We can talk about the difference between two items

## 1. Interval Variable Gotcha

B\_block

- ▶ Temperature has difference
- ▶ '0' is arbitrary
- ▶ Does not indicate that there is no temperature

## 1. Ratio Variables

B\_block

- ▶ the same as interval but with a *true* 0
- ▶ Number of children
- ▶ Age
- ▶ Prior Arrests

- ▶ Commute time

## 1. Dependent and Independent B\_block

- ▶ Dependent Variable is what you are trying to predict
- ▶ Independent is what you are using
- ▶ There are many names for each

## 1. Synonyms for Dependent B\_block

- ▶ outcome
- ▶ response

## 1. Synonyms for Independent B\_block

- ▶ feature
- ▶ explanatory
- ▶ causal

## 1. Validity B\_definition

- ▶ *Validity*: Addresses the question of whether the variable used actually reflects the concept or theory you seek to examine.
- ▶ *Reliability*: A measure is reliable if it is consistent and stable.
  - ▶ *stable* is if it remains the same when measured in the same group

- ▶ *reliable* is if the same person in different groups will score similarly

## 1. Univariate descriptive statistics describes one variable B\_block

- ▶ Univariate statistics is composed of:
  - ▶ Central Tendency
  - ▶ Measures of Dispersion
  - ▶ Form of the distribution

### 1. What is a statistic?

B\_definition:BMCOL

- ▶ A *statistic* is one number that summarizes many numbers

### 2. Example statistics

B\_example:BMCOL

- ▶ a batting average
- ▶ a shooting percentage
- ▶ an *average* length of stay for pretrial detention
- ▶ a *median* income for a zip code

### 1. Goal of Central Tendency

B\_block

- ▶ We want the single best number that describes **typical** case

- ▶ The measure of central tendency you choose depends on the level of measurement of the variable

## 1. Mode B\_definition

- ▶ The *mode* is the most frequently occurring value
- ▶ It is the only MCT appropriate for Nominal Data

## 1. Median B\_definition

- ▶ The *median* is the value in where half the values are greater and half are less

## 1. Mean B\_definition

- ▶ The *mean* is the sum of the values divided by the number of values

## 1. Calculation example B\_block

```
rdata <- c()  
rdata <- c(2, 3, 3, 4, 5, 5, 5, 6, 6, 6, 7, 7)  
rdata <- c(rdata,7, 7, 8, 8, 8, 9, 9, 10)  
table(rdata)  
table(rdata)[order(table(rdata),decreasing=TRUE)]
```

```
rdata
```

```
 2  3  4  5  6  7  8  9 10  
 1  2  1  3  3  4  3  2  1
```

```
rdata
```

```
 7  5  6  8  3  9  2  4 10  
 4  3  3  3  2  2  1  1  1
```

1. Median is the value that splits the distribution B\_block
  - ▶ into two equal parts
  - ▶ often used with distributions that are skewed or have outliers

```
myMedian <- function(x) {  
  x1 <- x[order(x,decreasing = F)]  
  l <- length(x)  
  if(l %% 2 == 0) {  
return( .5*(x1[.5*l]+x1[.5*l+1]))  
  } else {  
return( x1[(l+1)/2])  
  }  
}
```

1. Do we get the same result for our own median

B\_block

- ▶ function as the one built into R

```
median(rdata)
```

```
myMedian(rdata)
```

```
[1] 7
```

```
[1] 7
```

1. Mean is also known as the average

B\_block

- ▶ Only for Interval or Ratio level data
- ▶ Assumes order and equality of intervals
- ▶ Very sensitive to outliers and skew
- ▶  $\bar{X} = \frac{\sum X}{n}$

```
myMean <- function(x) {  
  myMean <- sum(x)/length(x)  
  return(myMean)  
}
```

```
mean(rdata)
```

```
myMean(rdata)
```

[1] 6.25

[1] 6.25

## 1. Mean, outliers and typical case

B\_block

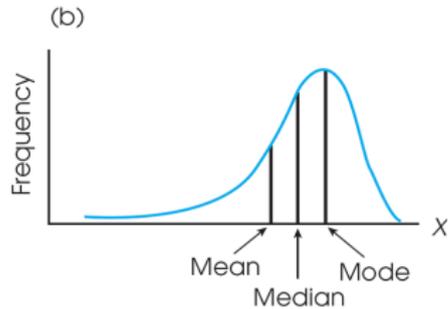
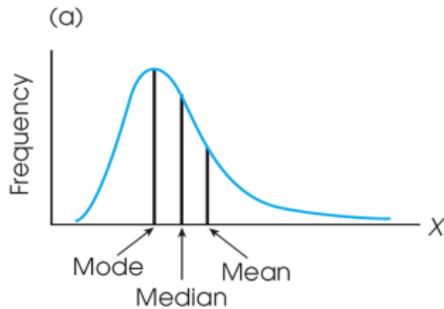
- ▶ the mean is sensitive to outliers
- ▶ outliers can be important
- ▶ crashes (stock market)
- ▶ frequent (arrestees)
- ▶ billionaires in income data
- ▶ You need to understand what you are describing/modeling
- ▶ **textbook answer** use the median in presence of skew and outliers

## 1. Defining skew

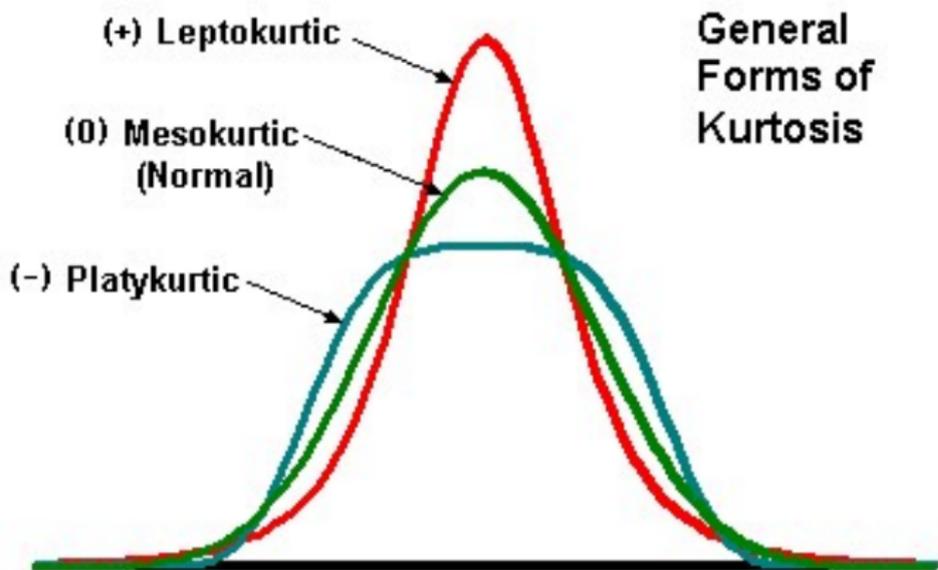
B\_definition

- ▶ A distribution is skew if it is not symmetric
- ▶ If a distribution is skewed it is conventional to use the median not the mean
- ▶ median moves less for a given outlier than the mean

## 1. Skew examples B\_ignoreheading Below are two asymmetrical distributions: (a) positive (right) skew, (b) negative (left) skew.



1. Skew examples B\_ignoreheading In the masters program you are also responsible for recognizing different levels of kurtosis



1. Choose the most appropriate measure of

B\_block

- ▶ central tendency (MCT) for:
  - ▶ SEX [sexr]
  - ▶ SEV MS ARREST CHARGE [asev1]
  - ▶ HOURS PER WEEK WORKED [hrspw]
  - ▶ UCR MS ARREST CHARGE [aucr1]
  - ▶ NUMBER PRIOR MISD CONV [pmis]

1. Recap so far B\_block
- ▶ *level of measurement* limit on the information in a variable
  - ▶ *central tendency* describe the typical case
  - ▶ *measures of dispersion* tell us how typical the typical case is
  
  - ▶ the measures we should use are dictated by the variable's level of measurement

1. Open the Rikers 1989 Data Set from Blackboard B\_block
- ▶ Choose the most appropriate MCT for:
    - ▶ SEX [sexr]
    - ▶ SEV MS ARREST CHARGE [asev1]
    - ▶ HOURS PER WEEK WORKED [hrspw]
    - ▶ UCR MS ARREST CHARGE [aucr1]
    - ▶ NUMBER PRIOR MISD CONV [pmis]

1. MCT R code B\_block

```
library(foreign)
myRikers <- read.spss(file="rikers1989.sav", to.data.frame=TRUE)
head(rikers[1:3,1:4])
```

Warning message:

```
In read.spss(file = "rikers1989.sav", to.data.frame = TRUE)
```

```
rikers1989.sav: Unrecognized record type 7, subtype 18
```

	caseid		aucr1	asev1	acd1
1	10	USE/POSS OTHER DRUGS	A	MISD	DRU
2	46		ROBBERY C	FELONY HARM TO PERS & PP	
3	56	USE/POSS OTHER DRUGS	A	MISD	DRU

1. Two distributions can have the same mean B\_block

- ▶ but very different spread of values
- ▶ the amount of variation is very important
- ▶ there can be a little, there can be a lot

1. More names for spread B\_block

- ▶ variation, dispersion
- ▶ Variation is important because the typical case can be misleading
- ▶ think of the crash or Bill Gates

1. Central Tendency vs. Variability

- ▶ Central Tendency shows typical case
- ▶ *Measures of dispersion* show spread of values around the typical case

## 1. small data examples

B\_block

```
d1 <- c( 7, 6, 3, 3, 1)
d2 <- c( 3, 4, 4, 5, 4)
d3 <- c( 4, 4, 4, 4, 4)
c(mean(d1), median(d1), sd(d1))
c( mean(d2), median(d2), sd(d2))
c( mean(d3), median(d3), sd(d3))

[1] 4.00000 3.00000 2.44949
[1] 4.0000000 4.0000000 0.7071068
[1] 4 4 0
```

## 1. Nominal variables

B\_block

- ▶ proportion in modal category
- ▶ Index of Qualitative Variation

$$IQV = \frac{K(100^2 - \sum_{i=1}^K p_i^2)}{100^2(K - 1)}$$

1. Measures of dispersion for ordinal variables B\_block
- ▶ proportion in modal category
  - ▶ Index of Qualitative Variation

1. Measures of dispersion for interval/ratio variables B\_block
- ▶ range
  - ▶ variance
  - ▶ standard deviation

1. Proportion in modal category formula B\_block

$$\frac{Number_{modal}}{Total N} * 100$$

1. Proportion in modal category code B\_block

```
rdata <- c(1,3,3,3,3,5)
freqTable <- table(rdata)
ordFreqTable <- freqTable[order(freqTable,decreasing = T)]
propOrdFTable <- prop.table(ordFreqTable)
100*propOrdFTable[1]
```

## 1. Computing IQV

B\_block

Fear of Crime	resp
Not Concerned at All	3
A Little Concerned	4
Quite Concerned	6
Very Concerned	20

## 1. Computing IQV in R

B\_block

```
myCr <- c(rep(1,3),rep(2,4),rep(3,6), rep(4,20))
```

```
myFreq <- table(myCr)
```

```
myFreq
```

```
myCr
```

```
1  2  3  4
```

```
3  4  6 20
```

## 1. continuing the calculation of IQV

```
K <- length(myFreq)
myPropFreq <- prop.table(myFreq)
sqProp <- apply(X=myPropFreq,MARGIN = 1,FUN = function(x)
sumSqProp <- sum(sqProp)
IQV <- (K/(K-1))*(1-sumSqProp)
IQV

[1] 0.7689011
```

## 1. The style of this

B\_block

- ▶ is called imperative
- ▶ if we want to make it so we can use it we make it a function

```
IQV <- function(myCr) {
  myFreq <- table(myCr)
  K <- length(myFreq)
  myPropFreq <- prop.table(myFreq)
  sqProp <- apply(X=myPropFreq,MARGIN = 1,
  FUN = function(x){return(x^2)})
  sumSqProp <- sum(sqProp)
  IQV <- (K/(K-1))*(1-sumSqProp)
```

```
    return(IQV)
}
```

1. There are three common measures of variability B\_block
  - ▶ range
  - ▶ variance
  - ▶ standard deviation

1. Calculating range B\_block
  - ▶  $Range = X_{maximum} - X_{minimum}$

```
myData<-c(35,60,80,93,98)
myRange <- max(myData)-min(myData)
myRange
range(myData)
```

```
[1] 63
```

```
[1] 35 98
```

1. Advantages of the range B\_block:BMCOL
  - ▶ easy to compute
  - ▶ interval includes all data

## 2. Disadvantages of the range

- ▶ crude because:
  - ▶ it relies on only two points
  - ▶ it ignores N-2 points
  - ▶ it is very sensitive to outliers

## 1. should never be used as the sole measure

B\_block

- ▶ More examples

```
d1 <- c(10,39,39,40,40,40,40,41,41,47)
```

```
d2 <- c(39,39,40,40,40,40,41,41,41,88)
```

```
d3 <- c(39,39,40,40,40,40,41,41,41,42)
```

```
myRanges<-c(max(d1),max(d2),max(d3))-c(min(d1),min(d2),min(d3))
```

```
myRanges
```

```
[1] 37 49 3
```

## 1. Variance and standard deviation

B\_block

- ▶ appropriate for Interval/Ratio data
- ▶ not used for nominal or ordinal data
- ▶ there is a one-to-one map from variance to standard deviation

```
[1] 10 39 39 40 40 40 40 41 41 47
```

```
[1] 100.0111
```

```
[1] 100.0111
```

## 1. Formulas for sample variance

B\_block:BMCOL

$$\text{var}(x) = \frac{\sum_{i=1}^N (x_i - \bar{x})^2}{N}$$

```
d1
```

```
var(d1)
```

```
md1 <- mean(d1)
```

```
sumSq <- unlist(lapply(X=(d1-md1),function(x){return(x^2)
```

```
myVar <- sum(sumSq)/(length(d1)-1)
```

```
myVar
```

## 2. Formulas for standard deviation

B\_block:BMCOL

$$\text{sd}(x) = \sqrt{\text{var}(x)}$$

```
sd(d1)
```

```
sqrt(myVar)
```

```
[1] 10.00056
```

```
[1] 10.00056
```

## 1. Sample vs. Population Standard Deviation B\_block

- ▶ They both use the difference between the mean ( $\mu$  or  $\bar{X}$ ) and each observation (individual X value)
- ▶ The sample uses “n-1” instead of N to adjust for sampling error
- ▶ Sample statistics are used to estimate population parameters
- ▶ Samples tend to have less variation than the populations
- ▶ n-1 is used to adjust for this by inflating a value we know to be artificially low
  
- ▶ long mathematical derivation **not going to do it**

## 1. Example of univariate analysis B\_block

- ▶ Number of arrests (1,5,7,8,9)

```
arrestData <- c(1,5,7,8,9)
mean(arrestData)
sd(arrestData)
```

## 1. Insight into standard deviation B\_block

- ▶ The reason why the sum of the deviations from the mean is always zero:
  - ▶ The mean serves as a balance point for the distribution

- ▶ The total distance of individual scores above the mean is exactly equal to the total distance of those below the mean
- ▶ The result is the positive and negative numbers cancelling each other out
- ▶ The solution is to get rid of the negative signs
- ▶ Squaring the deviations makes every number positive

1. Drawbacks of standard deviation B\_block:BMCOL

- ▶ It takes into account all values in the distribution
- ▶ Values far from the mean are given extra weight because deviations from the mean are squared

2. Drawbacks of standard deviation B\_block:BMCOL

- ▶ Standard deviation and variance are sensitive to outliers (extreme values)

- ▶ a display useful in summarizing distribution
- ▶ Goes back to Arthur Bowley's work in early 1900's
- ▶ Popularized by John Tukey in 1977's *Exploratory Data Analysis*
  
- ▶ class scores

Class Scores:

75	8	36	36	55
55	42	36	50	42
100	83	27	58	55
55	27	83	17	58
82	27	92	50	42

```
myScores<-c(75,8,36,36,55,55,42,36,50,42,100,83,27,58,55,55)
myOrdScores <- myScores[order(myScores)]
stem(myScores,scale=1)
```

```
stem(myScores,scale=2)
```

## 1. Topics for next time

- ▶ Probability
- ▶ The Gaussian Distribution and the “Normal” Curve
- ▶ Risk of Error