

Statistical Applications using



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Agenda

Statistical Applications using R

- Basic Statistics
- Correlation
- Linear Regression
- Multiple Regression
- 2 Sample T-Test
- 1 Sample T-Test
- ANOVA
- Clustering

Mean Syntax

```
mean(x, trim = 0, na.rm = FALSE, ...)
```

trim is used to drop some observations from both end of the sorted vector

na.rm is used to remove the missing values from the input vector.

Eg:

```
mean(x)
```

```
mean(x,trim=0.3)
```

Median Syntax

```
median(x, na.rm = FALSE)
```

x is the input vector.

na.rm is used to remove the missing values from the input vector.

Summary

summary(x)

Min, 1st Q, Median, Mean, 3rd Q, Max

Correlation

It provides a measure of strength and direction of linear relationship between two variables

Correlation Example - MTCars

```
corMat <- cor(mtcars[,c(1,3,6)])  
round(corMat,2)
```

	mpg	disp	wt
mpg	1.00	-0.85	-0.87
disp	-0.85	1.00	0.89
wt	-0.87	0.89	1.00

Linear Regression

Linear Regression

- Finding a straight line that best describes the data
- The best fit line is then used for making prediction

Linear Regression

Fitting linear model

```
m<-lm(y ~ x,data)
```

Intercept and Slope

```
coef(m)
```

```
summary(m)
```

Prediction

```
p <- predict(m, data.frame(x = ..))
```

LR Example - MTCars

```
m <- lm(mpg ~ wt, data = mtcars)
```

```
coef(m)
```

(Intercept)	wt
-------------	----

37.285126	-5.344472
-----------	-----------

```
summary(m)
```

```
p <- predict(m, data.frame(wt = 3))
```

Linear Regression - MTCars

```
plot(mgp~wt,data=mtcars)
```

```
abline(m)
```

Challenge: Linear Regression

1. Do a linear regression for the quake dataset - mag vs stations
2. Predict the quake mag when there are 100 stations receiving the quake signal.

Time : 5 mins

Multiple Regression

```
m<-lm(y ~ x1+x2+x1*x2...,data)
```

```
coef(m)
```

```
summary(m)
```

```
p<- predict(m, data.frame(x1=..,x2=..,...))
```

Multiple Regression - MTCars

```
m <- lm(mpg ~ wt+hp, data = mtcars)
```

```
coef(m)
```

(Intercept)	wt	hp
37.22	-3.87	-0.03

```
summary(m)
```

```
p<- predict(m, data.frame(wt = 3, hp = 200))
```

MR with Interaction - MTCars

```
m <- lm(mpg ~ wt*hp, data = mtcars)
```

```
coef(m)
```

(Intercept)	hp	wt	hp*wt
49.80	-0.12	-8.21	0.02

```
summary(m)
```

```
p<- predict(m, data.frame(wt = 3, hp = 200))
```


Logistics Regression

Logistics regression is a nonlinear regression that apply to response (y) that is binary

```
m<-glm(y ~ x1+x2+x3...,data)
```

```
coef(m)
```

```
summary(m)
```

```
p<- predict(m, data.frame(x1=...,x2=...,...))
```

Logistics Regression - MTCars

```
m <- glm(am ~ mpg+wt+hp, data = mtcars)
```

```
coef(m)
```

(Intercept)	mpg	wt	hp
0.195756590	0.036308683	-0.338756898	
	0.003891913		

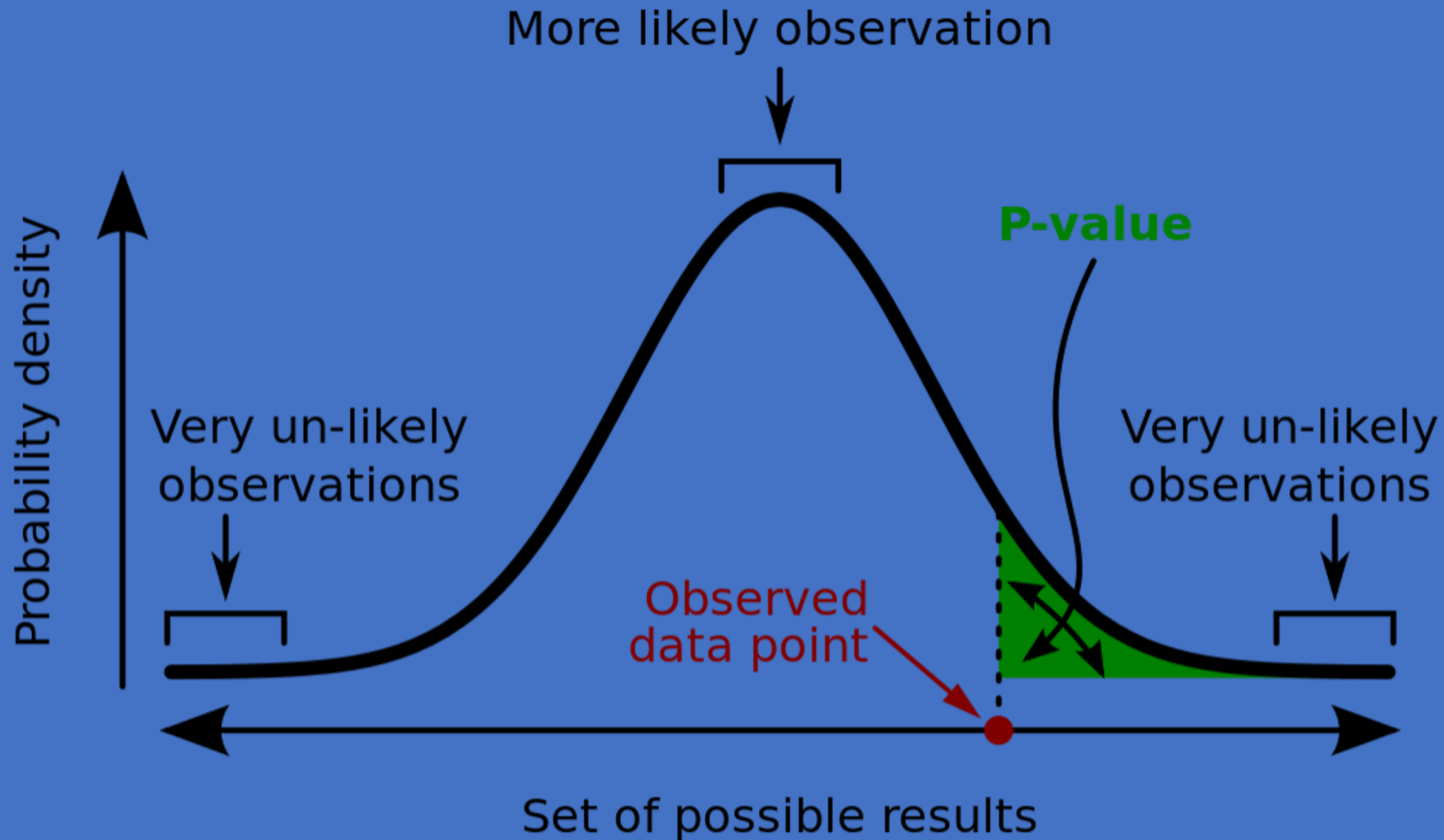
```
summary(m)
```

Hypothesis Testing

Hypothesis Testing

- Assume Null Hypothesis is true - No difference or no effect
- Compute the p-value
- If the p-value is small, then reject Null Hypothesis, else do not reject Null Hypothesis

P-Value



A **p-value** (shaded green area) is the probability of an observed (or more extreme) result assuming that the null hypothesis is true.

Guideline for p-value

If p value $> .10 \rightarrow$ “not significant”

If p value $\leq .10 \rightarrow$ “marginally significant”

If p value $\leq .05 \rightarrow$ “significant”

If p value $\leq .01 \rightarrow$ “highly significant.”

2 Sample t-Test (2 Sided)

```
data(sleep)
```

```
extra<-sleep$extra
```

```
group<-sleep$group
```

```
t.test(extra~group,data=sleep)
```

2 Sample t-Test (One Sided)

```
data(sleep)
```

```
extra<-sleep$extra
```

```
group<-sleep$group
```

```
t.test(extra~group,data=sleep,,alternative="less")
```


1 Sample t-Test

```
t.test(extra ~ group, sleep, paired=TRUE)
```

Challenge: t-Test

Do a 2 sample t-test to compare the performance of chickwts feed - casein vs horsebean

Time : 5 mins

ANOVA

- t-test can do test 2 groups.
- When there are more than 2 groups then use ANOVA to test if there are statistically significant difference between the means.
- ANOVA does this by analysing the variance in the data set.
- ANOVA is similar to multiple regression

ANOVA

```
m<-aov(y~x1+x2+x1*x2....,data)  
summary(m)
```

ANOVA Example - Chickwts

```
m<-aov(weight~feed,data=chickwts)  
summary(m)
```

Alternatively can use Linear Regression Method

```
m <- lm(weight~feed,data=chickwts)  
summary(m)
```

Challenge: ANOVA

Perform an ANOVA to determine any difference between the test scores of 3 teaching methods

Method A	Method B	Method C
79	71	82
86	77	68
94	81	70
89	83	76