

Lesson 5:

Statistical Testing

Recap: Bivariate Statistics

- correlation vs. linear regression

Important commands (recap)

COMMAND	EFFECT
<code>plot(var1, var2)</code>	plot scatterplot of two continuous variables
<code>scatterplot()</code>	package <code>car</code> , advances scatterplot
<code>cor(var1, var2)</code>	package <code>stats</code> , calculate correlation between two variables
<code>cor.test()</code>	package <code>stats</code> , test the correlation for significance
<code>rcorr()</code>	package <code>Hmisc</code> , calculates correlation also for more than one variable, adds p values
<code>abline()</code>	add linear regression line to scatterplot
<code>lm()</code>	package <code>stats</code> , calculates linear regression model
<code>summary()</code>	summarizes linear regression model

Statistical Testing

1. continuous variables

- difference between 2 groups: t Test
- difference between > 2 groups: ANOVA

2. categorical variables

- chi-square Test / Fisher exact Test

3. survival data: log-rank Test

1. continuous: 2 groups – t Test

assumptions:

- is data normally distributed?
- load package `stats`

```
> shapiro.test(data$variable)
```

Shapiro-Wilk normality test

```
data: data$cariable
```

```
W = 0.92449, p-value < 2.2e-16
```

if p-value below 0.05 → data is NOT normally distributed!



1. continuous: 2 groups – t Test

assumptions:

- is data normally distributed?

→ YES: t Test (package `stats`)

```
> t.test(group_1$var, group_2$var)
```

```
Welch Two Sample t-test
```

```
data: group_1$var and group_2$var
```

```
t = 0.14342, df = 915.79, p-value = 0.886
```

```
alternative hypothesis: true difference in means is not equal to 0
```

```
95 percent confidence interval:
```

```
-104.3638 120.8202
```

```
sample estimates:
```

```
mean of x mean of y
```

```
1674.243 1666.014
```

1. continuous: 2 groups – Wilcoxon Test

assumptions:

- is data normally distributed?

→ NO: Wilcoxon Test(package stats)

```
> wilcox.test(group_1$var, group_2$var)
```

```
Wilcoxon rank sum test with continuity correction
```

```
data: group_1$var and group_2$var
```

```
W = 108542, p-value = 0.8349
```

```
alternative hypothesis: true location shift is not equal to 0
```

1. continuous: > 2 groups – ANOVA

assumptions:

- is data normally distributed? (`shapiro.test()`)
- are variances equal? → Levene Test (package `car`)

```
> leveneTest(data$variable, data$categories, center = mean)
```

```
Levene's Test for Homogeneity of Variance (center = mean)
```

	Df	F value	Pr(>F)
group	2	2.7311	0.06567 .
	926		

if p-value below 0.05 → groups do NOT have same variance!

```
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

1. continuous: > 2 groups – ANOVA

assumptions:

- is data normally distributed? → YES
- are variances equal? → YES

→ ANOVA

```
> summary(aov(variable ~ categories, data = data))
```

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
categories	2	7540308	3770154	5	0.00692 **
Residuals	926	698256213	754056		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

1. continuous: > 2 groups – ANOVA

assumptions:

- is data normally distributed? → NO
- Kruskal-Wallis Test

```
> kruskal.test(variable ~ categories, data = data)
```

```
Kruskal-Wallis rank sum test
```

```
data: variable by categories
```

```
Kruskal-Wallis chi-squared = 9.2014, df = 2, p-value = 0.01004
```

1. continuous: > 2 groups – ANOVA

assumptions:

- are variances equal? → NO
→ Welch Test

```
> oneway.test(variable ~ categories, data = data)
```

One-way analysis of means (not assuming equal variances)

data: variable and categories

F = 5.0458, num df = 2.0, denom df = 616.8, p-value = 0.006704

2. categorical: χ^2 Test

- use `chisq.test()` from package `stats`

```
> chisq.test(data$cat_1, data$cat_2)
```

Pearson's Chi-squared test with Yates' continuity correction

```
data: data$cat_1 and data$cat_2
```

```
X-squared = 1.511, df = 1, p-value = 0.219
```

2. categorical (small samples): Fisher Test

- for very small samples use Fisher-exact Test:

```
fisher.test(data$cat_1, data$cat_2)
```

3. survival data: Log-rank Test

- use `survdif()` from package `survival`

```
> p=survdif(Surv(time, status == 1) ~ sex, data=test1)
> p
Call:
survdif(formula = Surv(time, status == 1) ~ sex, data = test1)

      N Observed Expected (O-E)^2/E (O-E)^2/V
sex=0 445      215      217  0.01049  0.0202
sex=1 484      237      235  0.00965  0.0202

      chisq= 0 on 1 degrees of freedom, p= 0.9
> pchisq(p$chisq, length(p$n)-1, lower.tail = FALSE)
[1] 0.8871178
> |
```

3. survival data: Log-rank Test

- requires packages `survival` and `survminer`

```
> fit <- survfit(Surv(data$cont_var, data$status == 1) ~ cat_var, data = data)
> summary(fit)
```

```
Call: survfit(formula = Surv(colon$time, colon$status == 1) ~ sex, data = colon)
```

```

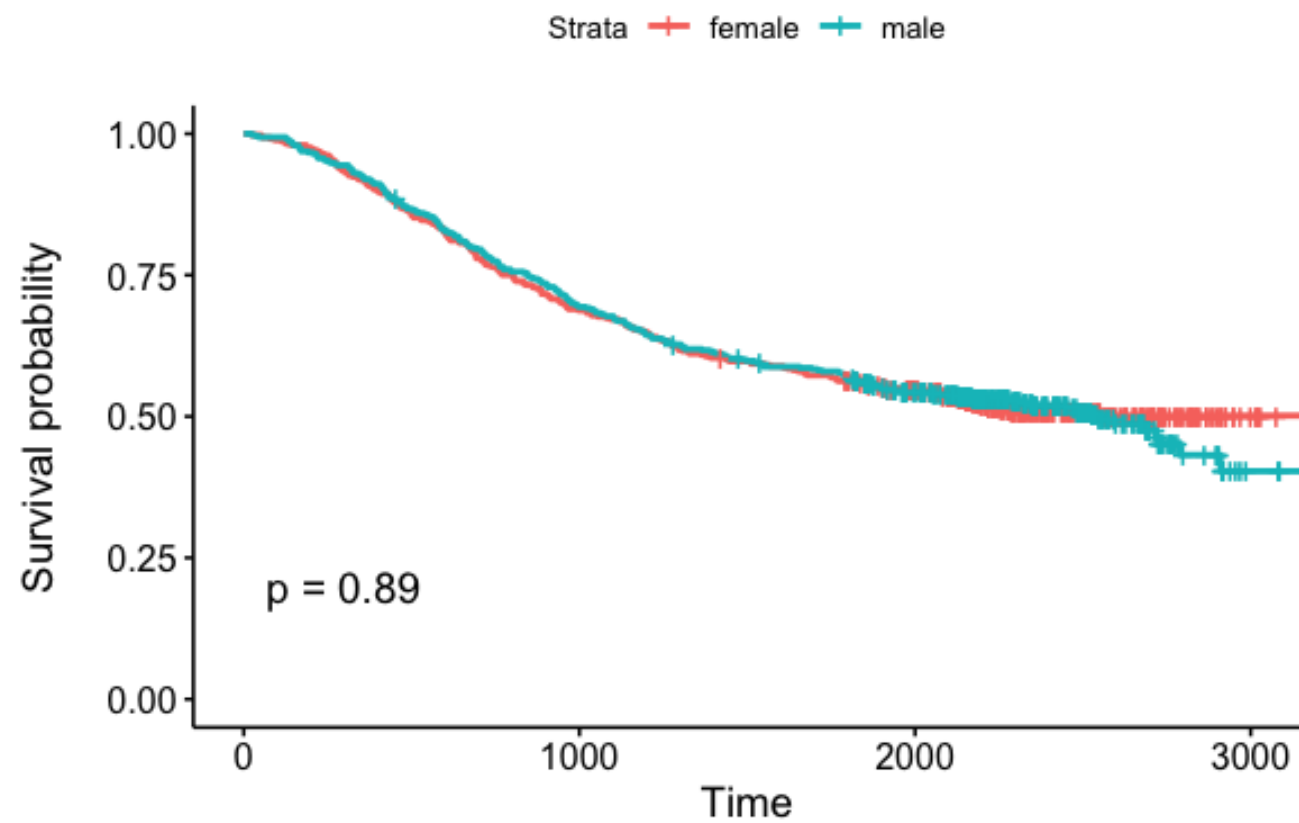
               cat_var=0
cont_var n.risk n.event survival std.err lower 95% CI upper 95% CI
    23      445      1   0.998  0.00224    0.993      1.000
    52      444      1   0.996  0.00317    0.989      1.000
    56      443      1   0.993  0.00388    0.986      1.000
...

```

3. survival data: Log-rank Test

- requires packages `survival` and `survminer`

```
> fit <- survfit(Surv(data$cont_var, data$status == 1) ~ cat_var, data = data)
> ggsurvplot(fit, data = colon,
              pval = TRUE,
              risk.table = TRUE,
              legend.labs = c("female", "male"))
```



Number at risk

Strata	female	445	306	218	10
	male	484	335	238	7
		0	1000	2000	3000

Time

Important commands

COMMAND	PACKAGE	EFFECT
<code>shapiro.test(data\$variable)</code>	stats	tests variable for normal distribution
<code>t.test(group_1\$var, group_2\$var)</code>	stats	tests two groups with normally distributed data for equality
<code>wilcox.test(group_1\$var, group_2\$var)</code>	stats	tests two groups with not normally distributed data for equality
<code>summary(aov(variable ~ categories, data = data))</code>	stats	tests more than two groups split by categories for equality
<code>leveneTest(data\$variable, data\$categories, center = mean)</code>	car	tests a variable split into groups by categories for variance homogeneity
<code>oneway.test(variable ~ categories, data = data)</code>	stats	tests more than two groups where variances are unequal for equality
<code>kruskal.test(variable ~ categories, data = data)</code>	stats	tests more than two groups that are not normally distributed for equality
<code>chisq.test(data\$cat_1, data\$cat_2)</code>	stats	tests two categorical variables
<code>fisher.test(data\$cat_1, data\$cat_2)</code>	stats	tests two categorical variables with small n
<code>Surv(data\$cont_var, data\$status == 1)</code>	survival	
<code>survfit(survfunction ~ cat_var, data = data)</code>	survival	
<code>ggsurvplot(fit, data = colon, pval = TRUE, risk.table = TRUE, legend.labs = c("female", "male"))</code>	survminer	