

## Aula Prática 02 - Estatística Experimental

### DELINEAMENTO EM BLOCOS CASUALIZADOS

```

> rm(list = ls(all = TRUE))
> set.seed(311)
> Bloc <- c("I", "II", "III")
> Trat <- (seq(1, 9, 1))
> set.seed(13)
> Parcela <- sample(1:length(Trat))
> Bloc

[1] "I"    "II"   "III"

> Trat

[1] 1 2 3 4 5 6 7 8 9

> Parcela

[1] 7 2 3 1 5 6 8 9 4

> for (k in 2:length(Bloc)) Parcela <- c(Parcela, sample(1:length(Trat)))
> Parcela

[1] 7 2 3 1 5 6 8 9 4 1 6 7 8 3 5 2 9 4 8 6 1 4 9 3 7 2 5

> layout <- cbind(expand.grid(Tratamento = Trat, Bloco = Bloc),
+   Parcela)
> layout

```

	Tratamento	Bloco	Parcela
1	1	I	7
2	2	I	2
3	3	I	3
4	4	I	1
5	5	I	5
6	6	I	6
7	7	I	8
8	8	I	9
9	9	I	4
10	1	II	1
11	2	II	6
12	3	II	7
13	4	II	8
14	5	II	3
15	6	II	5
16	7	II	2
17	8	II	9
18	9	II	4

19	1	III	8
20	2	III	6
21	3	III	1
22	4	III	4
23	5	III	9
24	6	III	3
25	7	III	7
26	8	III	2
27	9	III	5

```
> bloco <- c(1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 2, 2, 2, 2, 2, 2, 2, 2,
+ 2, 3, 3, 3, 3, 3, 3, 3, 3, 3)
> trat <- c(1, 2, 3, 4, 5, 6, 7, 8, 9, 1, 2, 3, 4, 5, 6, 7, 8,
+ 9, 1, 2, 3, 4, 5, 6, 7, 8, 9)
> prod <- c(145, 200, 183, 190, 180, 130, 206, 250, 164, 155, 190,
+ 186, 175, 160, 160, 165, 271, 190, 166, 190, 208, 186, 156,
+ 130, 170, 230, 193)
> dados <- as.data.frame(cbind(bloco, trat, prod))
> dados
```

	bloco	trat	prod
1	1	1	145
2	1	2	200
3	1	3	183
4	1	4	190
5	1	5	180
6	1	6	130
7	1	7	206
8	1	8	250
9	1	9	164
10	2	1	155
11	2	2	190
12	2	3	186
13	2	4	175
14	2	5	160
15	2	6	160
16	2	7	165
17	2	8	271
18	2	9	190
19	3	1	166
20	3	2	190
21	3	3	208
22	3	4	186
23	3	5	156
24	3	6	130
25	3	7	170
26	3	8	230
27	3	9	193

```
> names(dados)
```

```
[1] "bloco" "trat" "prod"
```

```
> is.factor(dados$bloco)
```

```
[1] FALSE
```

```
> dados$bloco <- as.factor(dados$bloco)
> is.factor(dados$trat)
```

```
[1] FALSE
```

```
> dados$trat <- as.factor(dados$trat)
```

```
> sb <- with(dados, tapply(prod, bloco, sum))
> sb
```

```
      1      2      3
1648 1652 1629
```

```
> st <- with(dados, tapply(prod, trat, sum))
> st
```

```
      1      2      3      4      5      6      7      8      9
466 580 577 551 496 420 541 751 547
```

```
> media.por.trat <- with(dados, tapply(prod, trat, mean))
> media.por.trat
```

```
      1      2      3      4      5      6      7      8
155.3333 193.3333 192.3333 183.6667 165.3333 140.0000 180.3333 250.3333
      9
182.3333
```

```
> variancia.por.trat <- with(dados, tapply(prod, trat, var))
> variancia.por.trat
```

```
      1      2      3      4      5      6      7      8
110.33333 33.33333 186.33333 60.33333 165.33333 300.00000 500.33333 420.33333
      9
254.33333
```

```
> mod1 <- with(dados, aov(prod ~ bloco + trat))
> names(mod1)
```

```
[1] "coefficients" "residuals"    "effects"      "rank"
[5] "fitted.values" "assign"       "qr"          "df.residual"
[9] "contrasts"    "xlevels"     "call"        "terms"
[13] "model"
```

```
> anova(mod1)
```

## Analysis of Variance Table

Response: prod

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
bloco	2	33.6	16.78	0.0666	0.9358
trat	8	22981.3	2872.67	11.4114	2.637e-05 ***
Residuals	16	4027.8	251.74		

---

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

```
> res <- mod1$residuals
> var.por.trat <- with(dados, tapply(res, trat, var))
> F <- max(variancia.por.trat)/min(variancia.por.trat)
> F
```

[1] 15.01

```
> with(dados, bartlett.test(prod ~ bloco + trat))
```

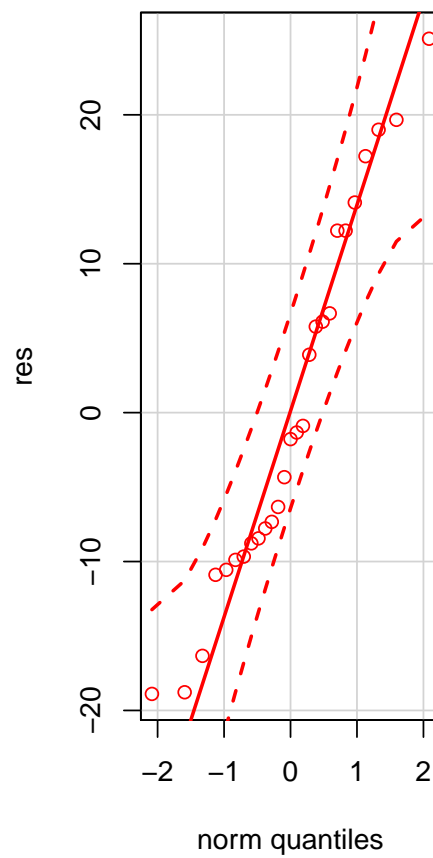
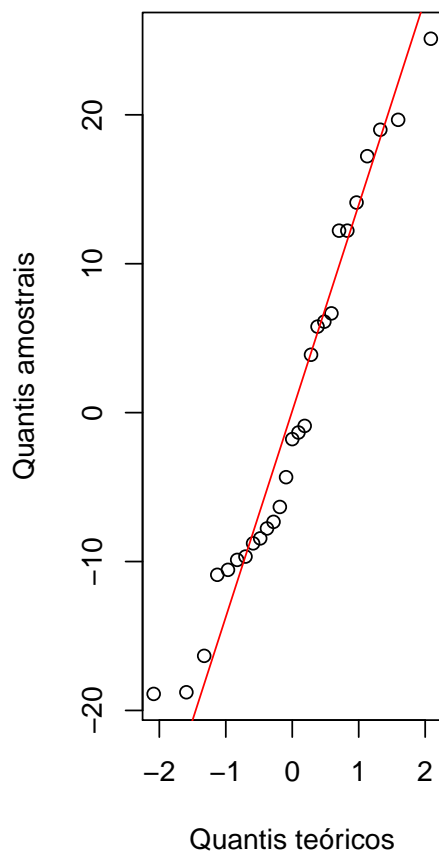
Bartlett test of homogeneity of variances

data: prod by bloco by trat

Bartlett's K-squared = 0.3244, df = 2, p-value = 0.8503

```
> require(car)
> par(mfrow = c(1, 2))
> qqnorm(res, xlab = "Quantis teóricos", ylab = "Quantis amostrais")
> qqline(res, col = "red")
> qqPlot(res)
```

### Normal Q-Q Plot



```
> shapiro.test(res)
```

Shapiro-Wilk normality test

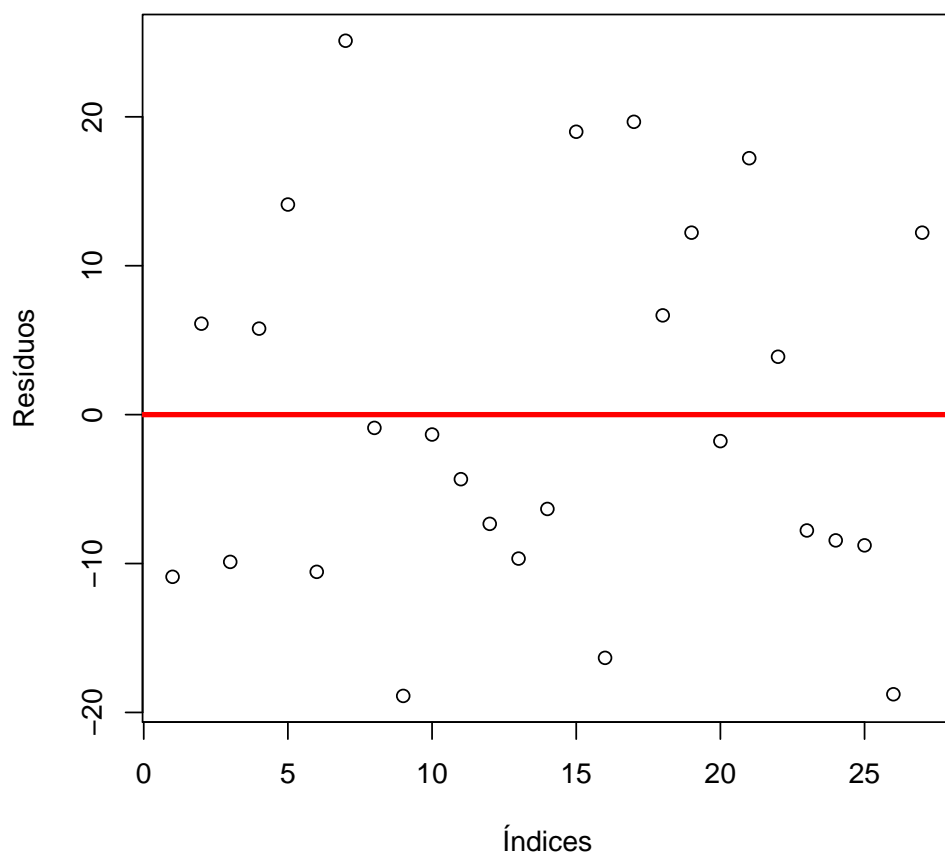
data: res

W = 0.9476, p-value = 0.1873

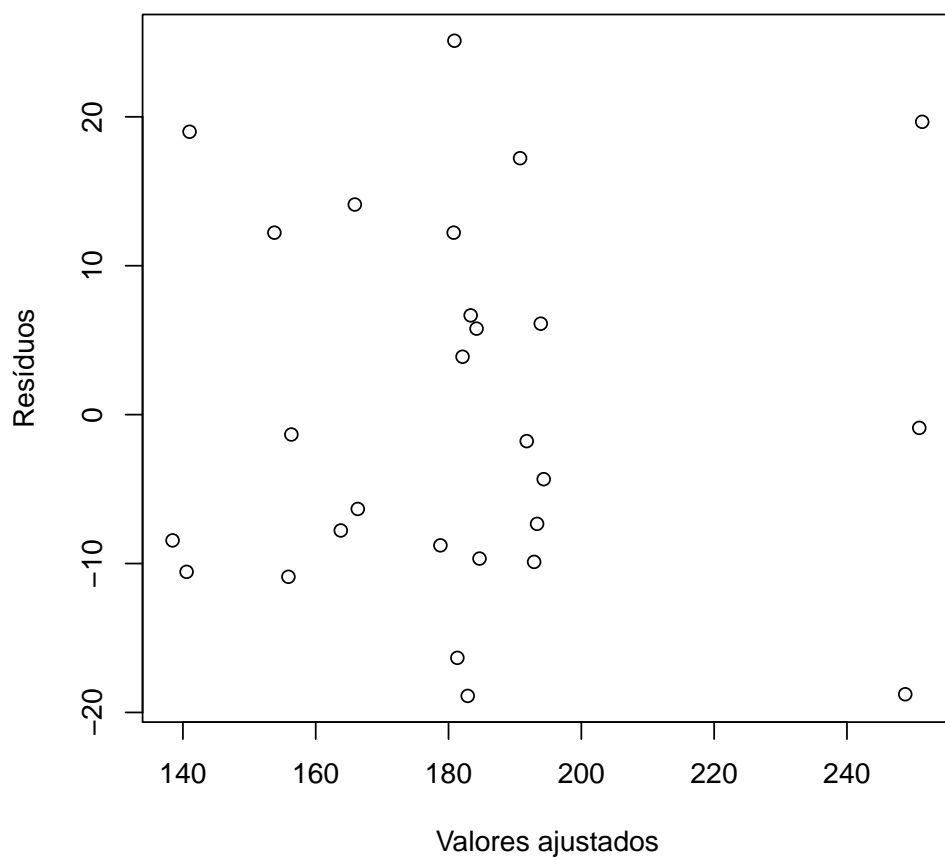
```
> par(mfrow = c(1, 1))
```

```
> plot(mod1$res, xlab = "Índices", ylab = "Resíduos")
```

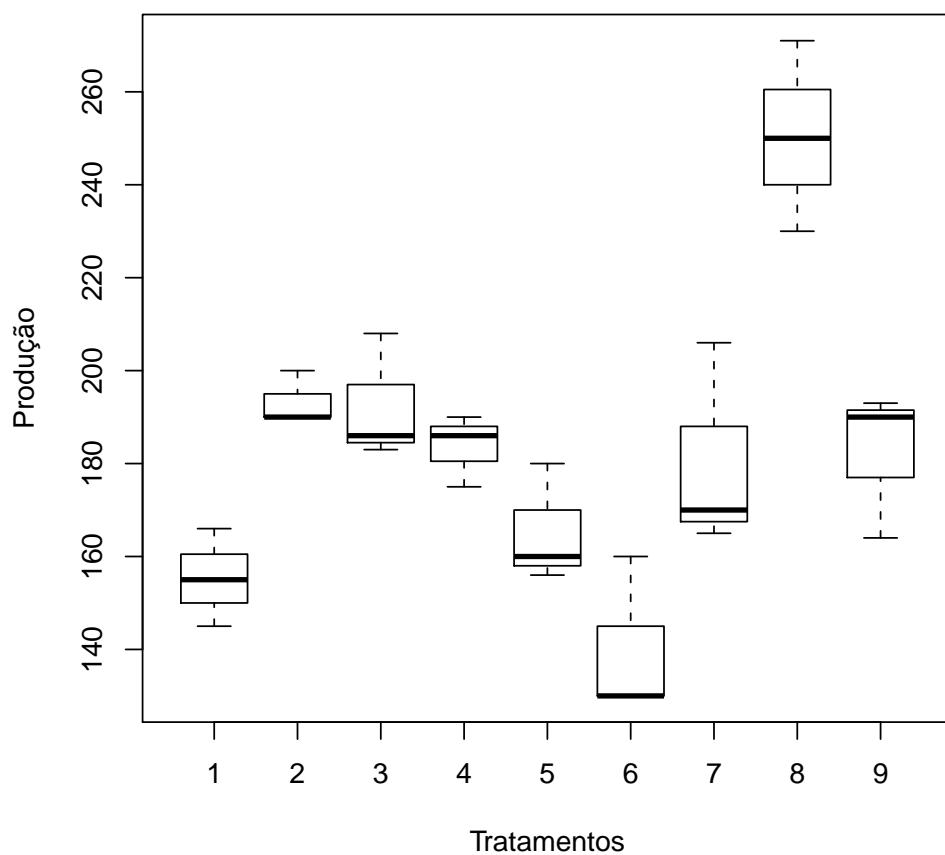
```
> abline(h = 0, col = "red", lwd = 3)
```



```
> plot(fitted(mod1), res, xlab = "Valores ajustados", ylab = "Resíduos")
```



```
> boxplot(prod ~ trat, xlab = "Tratamentos", ylab = "Produção",
+         data = dados)
```



```
> mod1$coeff
```

```
(Intercept)      bloco2      bloco3      trat2      trat3      trat4
155.8888889    0.4444444   -2.1111111   38.0000000   37.0000000   28.3333333
      trat5      trat6      trat7      trat8      trat9
10.0000000  -15.3333333   25.0000000   95.0000000   27.0000000
```

```
> b1 <- c(0, mod1$coeff[2:3])
```

```
> tr <- c(0, mod1$coeff[4:11])
```

```
> b1
```

```
      bloco2      bloco3
0.0000000    0.4444444   -2.1111111
```

```
> tr
```

```
      trat2      trat3      trat4      trat5      trat6      trat7      trat8
0.00000    38.00000    37.00000    28.33333    10.00000   -15.33333    25.00000    95.00000
      trat9
27.00000
```



```
> bltr <- rep(bl, 3) * rep(tr, rep(3, 9))
> ttna <- update(mod1, . ~ . + bltr)
> anova(ttna)
```

#### Analysis of Variance Table

Response: prod

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
bloco	2	33.6	16.78	0.0654	0.9370
trat	8	22981.3	2872.67	11.1974	4.458e-05 ***
bltr	1	179.6	179.56	0.6999	0.4159
Residuals	15	3848.2	256.55		

---

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

```
> q <- qtkey(0.95, 9, 16)
> delta <- q * sqrt(251.7/sqrt(3))
> tcm.tu <- TukeyHSD(mod1)
> tcm.tu
```

Tukey multiple comparisons of means  
95% family-wise confidence level

Fit: aov(formula = prod ~ bloco + trat)

\$bloco

	diff	lwr	upr	p adj
2-1	0.4444444	-18.85487	19.74376	0.9980554
3-1	-2.1111111	-21.41043	17.18820	0.9571497
3-2	-2.5555556	-21.85487	16.74376	0.9379209

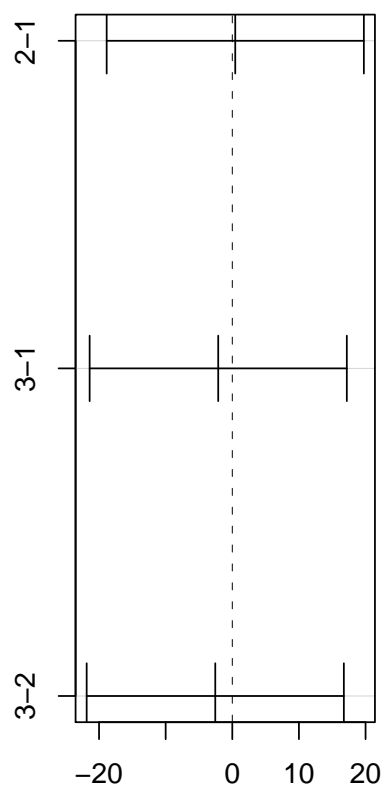
\$trat

	diff	lwr	upr	p adj
2-1	38.000000	-8.085796	84.085796	0.1520249
3-1	37.000000	-9.085796	83.085796	0.1728150
4-1	28.333333	-17.752463	74.419129	0.4559717
5-1	10.000000	-36.085796	56.085796	0.9962223
6-1	-15.333333	-61.419129	30.752463	0.9489958
7-1	25.000000	-21.085796	71.085796	0.6053536
8-1	95.000000	48.914204	141.085796	0.0000460
9-1	27.000000	-19.085796	73.085796	0.5143733
3-2	-1.000000	-47.085796	45.085796	1.0000000
4-2	-9.666667	-55.752463	36.419129	0.9969942
5-2	-28.000000	-74.085796	18.085796	0.4703201
6-2	-53.333333	-99.419129	-7.247537	0.0172692
7-2	-13.000000	-59.085796	33.085796	0.9799785
8-2	57.000000	10.914204	103.085796	0.0099947
9-2	-11.000000	-57.085796	35.085796	0.9929220
4-3	-8.666667	-54.752463	37.419129	0.9985839
5-3	-27.000000	-73.085796	19.085796	0.5143733
6-3	-52.333333	-98.419129	-6.247537	0.0200347
7-3	-12.000000	-58.085796	34.085796	0.9877062
8-3	58.000000	11.914204	104.085796	0.0086074
9-3	-10.000000	-56.085796	36.085796	0.9962223

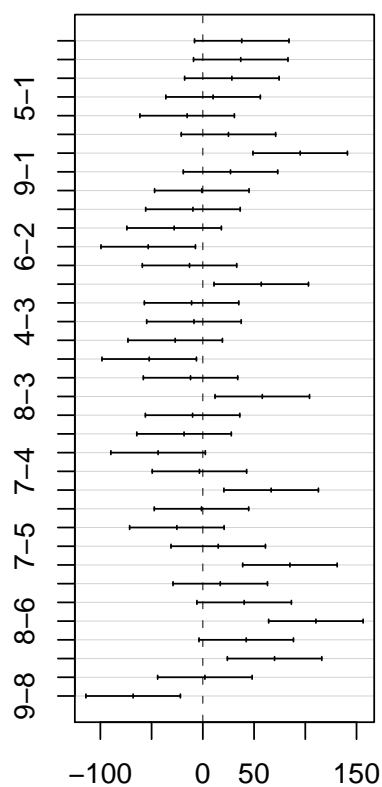
5-4	-18.333333	-64.419129	27.752463	0.8763516
6-4	-43.666667	-89.752463	2.419129	0.0705323
7-4	-3.333333	-49.419129	42.752463	0.9999989
8-4	66.666667	20.580871	112.752463	0.0023716
9-4	-1.333333	-47.419129	44.752463	1.0000000
6-5	-25.333333	-71.419129	20.752463	0.5900630
7-5	15.000000	-31.085796	61.085796	0.9546944
8-5	85.000000	38.914204	131.085796	0.0001740
9-5	17.000000	-29.085796	63.085796	0.9134401
7-6	40.333333	-5.752463	86.419129	0.1116698
8-6	110.333333	64.247537	156.419129	0.0000069
9-6	42.333333	-3.752463	88.419129	0.0849582
8-7	70.000000	23.914204	116.085796	0.0014541
9-7	2.000000	-44.085796	48.085796	1.0000000
9-8	-68.000000	-114.085796	-21.914204	0.0019490

```
> par(mfrow = c(1, 2))
> plot(tcm.tu)
```

**95% family-wise confidence level**      **95% family-wise confidence level**



Differences in mean levels of bloco



Differences in mean levels of trat

```
> require(laercio)
> LTukey(mod1, "trat", conf.level = 0.95)
```

# TUKEY TEST TO COMPARE MEANS

Confidence level: 0.95  
 Dependent variable: prod  
 Variation Coefficient: 8.69116 %

Independent variable: trat

Factors	Means	
8	250.333333333333	a
2	193.333333333333	b
3	192.333333333333	b
4	183.666666666667	bc
9	182.333333333333	bc
7	180.333333333333	bc
5	165.333333333333	bc
1	155.333333333333	bc
6	140	c

> LTukey(mod1, "trat", conf.level = 0.99)

# TUKEY TEST TO COMPARE MEANS

Confidence level: 0.99  
 Dependent variable: prod  
 Variation Coefficient: 8.69116 %

Independent variable: trat

Factors	Means	
8	250.333333333333	a
2	193.333333333333	b
3	192.333333333333	b
4	183.666666666667	b
9	182.333333333333	b
7	180.333333333333	b
5	165.333333333333	b
1	155.333333333333	b
6	140	b