STAT 3690 Lecture Note

Part V: Comparisons of population mean vectors

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Comparisons of population mean vectors

Comparing two population mean vectors (J&W Sec. 6.3)

• Two independent samples following two distributions with equal covariance

$$egin{aligned} &- oldsymbol{X}_{11}, \dots, oldsymbol{X}_{1n_1} \overset{ ext{iid}}{\sim} ext{MVN}_p(oldsymbol{\mu}_1, oldsymbol{\Sigma}) \ &- oldsymbol{X}_{21}, \dots, oldsymbol{X}_{2n_2} \overset{ ext{iid}}{\sim} ext{MVN}_p(oldsymbol{\mu}_2, oldsymbol{\Sigma}) \end{aligned}$$

- Let \bar{X}_i and S_i be the sample mean and sample covariance for the *i*th sample, i = 1, 2.
- Hypotheses $H_0: \boldsymbol{\mu}_1 = \boldsymbol{\mu}_2$ v.s. $H_1: \boldsymbol{\mu}_1 \neq \boldsymbol{\mu}_2$
- Test statistic following LRT

$$T(\mathcal{X}) = (\bar{\boldsymbol{X}}_1 - \bar{\boldsymbol{X}}_2)^{\top} \{ (n_1^{-1} + n_2^{-1}) \mathbf{S}_{\text{pool}} \}^{-1} (\bar{\boldsymbol{X}}_1 - \bar{\boldsymbol{X}}_2) \sim \frac{(n_1 + n_2 - 2)p}{n_1 + n_2 - p - 1} F(p, n_1 + n_2 - p - 1) \text{ under } H_0$$
$$- \mathbf{S}_{\text{pool}} = \frac{(n_1 - 1)\mathbf{S}_1 + (n_2 - 1)\mathbf{S}_2}{n_1 + n_2 - 2}$$

• Reject H_0 at level α when

$$T(\mathcal{X}) \ge \frac{p(n_1 + n_2 - 2)}{n_1 + n_2 - p - 1} F_{1-\alpha, p, n_1 + n_2 - p - 1}$$

• *p*-value

$$1 - F_{F_{1-\alpha,p,n_1+n_2-p-1}} \left[\frac{n_1 + n_2 - p - 1}{p(n_1 + n_2 - 2)} T(\mathcal{X}) \right]$$

```
options(digits = 4)
install.packages(c("dslabs"))
library(dslabs)
data("gapminder")
dataset1 = gapminder[
  !is.na(gapminder$infant_mortality) &
    gapminder$continent == "Africa" &
    gapminder$year == 2012,
    c('infant_mortality', "life_expectancy")]
dataset1 = as.matrix(dataset1)
```

```
dataset2 = gapminder[
  !is.na(gapminder$infant_mortality) &
    gapminder$continent == "Asia" &
    gapminder$year == 2012,
  c('infant_mortality', "life_expectancy")]
dataset2 = as.matrix(dataset2)
n1 <- nrow(dataset1); n2 <- nrow(dataset2); p <- ncol(dataset1)</pre>
(mu_hat1 <- colMeans(dataset1))</pre>
(mu_hat2 <- colMeans(dataset2))</pre>
(S1 <- cov(dataset1))
(S2 <- cov(dataset2))
S_{pool} \leftarrow ((n1 - 1)*S1 + (n2 - 1)*S2)/(n1+n2-2)
(lrt <- t(mu_hat1-mu_hat2) %*%</pre>
  solve((n1^-1 + n2^-1)*S_pool) %*%
  (mu_hat1-mu_hat2))
alpha <- .05
(cri.val \leftarrow (n1+n2-2)*p/(n1+n2-p-1)*qf(1-alpha, p, n1+n2-p-1))
lrt >= cri.val
(p.val = 1-pf((n1+n2-p-1)/(n1+n2-2)/p*lrt, p, n1+n2-p-1))
```

• Report: Testing hypotheses H_0 : in 2012 Asia and Africa shared the identical mean value in both infant mortality and life expectancy v.s. H_1 : otherwise, we carried on the LRT and obtained 87.65 as the value of test statistic and $[6.255, \infty)$ as the corresponding level .05 rejection region. In addition, the p-value was 4.952e-14. So, at the .05 level, there was a strong statistical evidence against H_0 , i.e., we rejected H_0 and believed that in 2012 Asia and Africa didn't share the identical mean value in infant mortality and/or life expectancy.

Comparing multiple population mean vectors (one-way multivariate analysis of variance (One-way MANOVA), J&W Sec. 6.4)

```
Generalization of two-sample problem

Model: m independent samples, where
* X<sub>11</sub>,..., X<sub>1n1</sub> iid MVN<sub>p</sub>(μ<sub>1</sub>, Σ)
* :

* * X<sub>m1</sub>,..., X<sub>mnm</sub> iid MVN<sub>p</sub>(μ<sub>m</sub>, Σ)
Hypotheses H<sub>0</sub>: μ<sub>1</sub> = ··· = μ<sub>m</sub> v.s. H<sub>1</sub>: otherwise

Alternatively

Model: m independent samples, where
* X<sub>11</sub>,..., X<sub>1n1</sub> iid MVN<sub>p</sub>(μ + τ<sub>1</sub>, Σ)
* :

* * X<sub>m1</sub>,..., X<sub>mnm</sub> iid MVN<sub>p</sub>(μ + τ<sub>m</sub>, Σ)
· Identifiability: ∑<sub>i</sub> τ<sub>i</sub> = 0 otherwise there are infinitely many models that lead to the same data-generating mechanism.
- Hypotheses H<sub>0</sub>: τ<sub>1</sub> = ··· = τ<sub>m</sub> = 0 v.s. H<sub>1</sub>: otherwise

Alternatively

Model: X<sub>ij</sub> = μ + τ<sub>i</sub> + E<sub>ij</sub> with E<sub>ij</sub> iid MVN<sub>p</sub>(0, Σ)
* Identifiability: ∑<sub>i</sub> τ<sub>i</sub> = 0
- Hypotheses H<sub>0</sub>: τ<sub>1</sub> = ··· = τ<sub>m</sub> = 0 v.s. H<sub>1</sub>: otherwise
```

- Sample means and sample covariances
 - Sample mean for the *i*th sample $X_{i} = n_{i}^{-1} \sum_{j} X_{ij}$
 - Sample covariance for the *i*th sample $\mathbf{S}_i = (n_i 1)^{-1} \sum_i (\mathbf{X}_{ij} \bar{\mathbf{X}}_{i\cdot}) (\mathbf{X}_{ij} \bar{\mathbf{X}}_{i\cdot})^{\top}$

 - Grand mean $\bar{X}_{\cdot\cdot\cdot} = \sum_{i} n_i \bar{X}_{i\cdot\cdot} / \sum_{i} n_i = \sum_{ij} X_{ij} / \sum_{i} n_i$ Decomposition of total (corrected) sum of squares and cross products matrix (SSP):

$$\mathbf{SSP}_{t} = \mathbf{SSP}_{w} + \mathbf{SSP}_{b}$$

- * Total (corrected) SSP: $\mathbf{SSP}_{\mathrm{t}} = \sum_{ij} (\boldsymbol{X}_{ij} \bar{\boldsymbol{X}}_{\cdot\cdot}) (\boldsymbol{X}_{ij} \bar{\boldsymbol{X}}_{\cdot\cdot})^{\top} = \mathbf{SSP}_{\mathrm{w}} + \mathbf{SSP}_{\mathrm{b}}$
- * Within-group SSP: $\mathbf{SSP}_{w} = \sum_{i} (\vec{n}_{i} 1) \mathbf{S}_{i} = \sum_{i,j} (X_{ij} \bar{X}_{i}) (X_{ij} \bar{X}_{i})^{\top}$
- * Between-group SSP: $\mathbf{SSP}_{b} = \sum_{i} n_{i} (\bar{\boldsymbol{X}}_{i.} \bar{\boldsymbol{X}}_{..}) (\bar{\boldsymbol{X}}_{i.} \bar{\boldsymbol{X}}_{..})^{\top}$
- ML estimator of $(\mu_1, \ldots, \mu_m, \Sigma)$
 - Unconstrained

*
$$\hat{\boldsymbol{\mu}}_i = \bar{\boldsymbol{X}}_{i\cdot} = n_i^{-1} \sum_j \boldsymbol{X}_{ij}$$

$$* \hat{\Sigma} = (\sum_{i} n_{i})^{-1} \mathbf{SSP}_{w}$$

$$- \text{ Under } H_{0}$$

$$* \hat{\boldsymbol{\mu}}_{i} = \bar{\boldsymbol{X}}.. \text{ for each } i$$

$$* \hat{\Sigma} = (\sum_{i} n_{i})^{-1} \mathbf{SSP}_{t}$$

- Likelihood ratio

$$\lambda = \left\{ \frac{\det(\mathbf{SSP_w})}{\det(\mathbf{SSP_t})} \right\}^{\sum_i n_i/2}$$

monotonic increasing with respect to the Wilk's lambda test statistic

$$\Lambda = \frac{\det(\mathbf{SSP_w})}{\det(\mathbf{SSP_t})} = \frac{\det(\mathbf{SSP_w})}{\det(\mathbf{SSP_w} + \mathbf{SSP_b})}$$

- $\Lambda \sim$ Wilk's lambda distribution $\Lambda(\mathbf{\Sigma}, \sum_i n_i m, m-1)$ under H_0
 - * Since $SSP_w \sim W_p(\Sigma, \sum_i n_i m)$ and $SSP_b \sim W_p(\Sigma, m-1)$ under H_0
 - * Bartlett's approximation (when $\sum_{i} n_{i} m$ is large)

$$\{(p+m)/2 - \sum_{i} n_i + 1\} \ln \Lambda \approx \chi^2(p(m-1))$$

- * Rao's approximation (default for R functions manova and car::Manova)
- Reject H_0 when Λ is small (equiv. **SSP**_b is large); specifically, H_0 is rejected when

$$\{(p+m)/2 - \sum_{i} n_i + 1\} \ln \Lambda \ge \chi^2_{1-\alpha,p(m-1)} \quad \text{OR} \quad \Lambda \le \exp\left\{\frac{\chi^2_{1-\alpha,p(m-1)}}{(p+m)/2 - \sum_{i} n_i + 1}\right\}$$

• p-value

$$1 - F_{\chi^2(p(m-1))} \left[\{ (p+m)/2 - \sum_i n_i + 1 \} \ln \Lambda \right]$$

- $F_{\chi^{2}(p(m-1))}$: the cdf of $\chi^{2}(p(m-1))$
- Exercise 5.1: investigating the factors in manufacturing plastic film (see W. J. Krzanowski (1988) Principles of Multivariate Analysis. A User's Perspective. Oxford UP, pp. 381.)

- Three response variables (tear, gloss and opacity) describing measured characteristics of the resultant film
- A total of 20 runs
- One factor RATE (rate of extrusion, 2-level, low or high) in the production test

```
options(digits = 4)
install.packages('car')
tear <- c(
  6.5, 6.2, 5.8, 6.5, 6.5, 6.9, 7.2, 6.9, 6.1, 6.3,
  6.7, 6.6, 7.2, 7.1, 6.8, 7.1, 7.0, 7.2, 7.5, 7.6
gloss <- c(
  9.5, 9.9, 9.6, 9.6, 9.2, 9.1, 10.0, 9.9, 9.5, 9.4,
  9.1, 9.3, 8.3, 8.4, 8.5, 9.2, 8.8, 9.7, 10.1, 9.2
opacity <- c(
 4.4, 6.4, 3.0, 4.1, 0.8, 5.7, 2.0, 3.9, 1.9, 5.7,
  2.8, 4.1, 3.8, 1.6, 3.4, 8.4, 5.2, 6.9, 2.7, 1.9
)
(X <- cbind(tear, gloss, opacity))</pre>
(rate <- factor(gl(2,10,length=nrow(X)), labels=c("Low", "High")))</pre>
# Bartlett's approximation to Wilks lambda distribution
X low <- X[rate == 'Low',]</pre>
X_high <- X[rate == 'High',]</pre>
n <- nrow(X); p <- ncol(X); m <- length(levels(rate))</pre>
(SSPcor = (n-1)*cov(X))
(SSPw \leftarrow (nrow(X_low) - 1)*cov(X_low) + (nrow(X_high) - 1)*cov(X_high))
(Lambda <- det(SSPw)/det(SSPcor))</pre>
(cri.point = exp(qchisq(0.95, p*(m-1))/((p+m)/2-n+1)))
Lambda <= cri.point
(p.val = 1-pchisq(((p+m)/2-n+1)*log(Lambda), p*(m-1)))
# Rao's approximation to Wilks lambda distribution
summary(manova(X ~ rate), test = 'Wilks')
summary(car::Manova(lm(X ~ rate)), test.statistic='Wilks')
```

• Report: Testing hypotheses H_0 : no RATE effect on film characteristics v.s. H_1 : otherwise, we carried on the Wilk's lambda test and obtained 0.4136 as the value of test statistic and $(-\infty, 0.6227]$ as the corresponding level .05 rejection region. In addition, the *p*-value was 0.002227. So, at the .05 level, there was statistical evidence against H_0 , i.e., we rejected H_0 and believed that there was an effect from RATE on film characteristics.

Testing for equality of covariance matrices (J&W Sec. 6.6)

• Model: m independent samples, where

```
- \boldsymbol{X}_{11}, \dots, \boldsymbol{X}_{1n_1} \stackrel{\text{iid}}{\sim} \text{MVN}_p(\boldsymbol{\mu}_1, \boldsymbol{\Sigma}_1)
- \vdots
- \boldsymbol{X}_{m1}, \dots, \boldsymbol{X}_{mn_m} \stackrel{\text{iid}}{\sim} \text{MVN}_p(\boldsymbol{\mu}_m, \boldsymbol{\Sigma}_m)
```

- Hypotheses $H_0: \Sigma_1 = \cdots = \Sigma_m$ v.s. $H_1:$ otherwise
- MLE of $(\boldsymbol{\mu}_1,\ldots,\boldsymbol{\mu}_m,\boldsymbol{\Sigma}_1,\ldots,\boldsymbol{\Sigma}_m)$

$$\begin{aligned} &-\text{ Under } H_0 \\ &* \hat{\boldsymbol{\mu}}_i = \bar{\boldsymbol{X}}_{i\cdot} = n_i^{-1} \sum_j \boldsymbol{X}_{ij} \\ &* \hat{\boldsymbol{\Sigma}}_i = (\sum_i n_i)^{-1} \mathbf{SSP_w} = (\sum_i n_i)^{-1} \sum_{ij} (\boldsymbol{X}_{ij} - \bar{\boldsymbol{X}}_{i\cdot}) (\boldsymbol{X}_{ij} - \bar{\boldsymbol{X}}_{i\cdot})^{\top} \text{ for all } i \\ &-\text{ No restriction on } \boldsymbol{\Sigma}_i \\ &* \hat{\boldsymbol{\mu}}_i = \bar{\boldsymbol{X}}_{i\cdot} = n_i^{-1} \sum_j \boldsymbol{X}_{ij} \\ &* \hat{\boldsymbol{\Sigma}}_i = n_i^{-1} (n_i - 1) \mathbf{S}_i = n_i^{-1} \sum_j (\boldsymbol{X}_{ij} - \bar{\boldsymbol{X}}_{i\cdot}) (\boldsymbol{X}_{ij} - \bar{\boldsymbol{X}}_{i\cdot})^{\top} \end{aligned}$$

• Likelihood ratio

$$\lambda = \prod_{i} \left[\frac{\det\{n_i^{-1}(n_i - 1)\mathbf{S}_i\}}{\det\{(\sum_{i} n_i)^{-1}(\sum_{i} n_i - m)\mathbf{S}_{\text{pool}}\}} \right]^{n_i/2} - \mathbf{S}_{\text{pool}} = (\sum_{i} n_i - m)^{-1}\mathbf{SSP}_{\text{w}}$$

• Box's M test statistic (a modification of LRT)

$$M = -2 \ln \prod_{i} \left(\frac{\det \mathbf{S}_{i}}{\det \mathbf{S}_{\text{pool}}} \right)^{(n_{i}-1)/2}$$

$$- \text{ Under } H_{0}$$

$$(1-u)M \approx \chi^{2}(p(p+1)(m-1)/2)$$

$$* u = \{\sum_{i} (n_{i}-1)^{-1} - (\sum_{i} n_{i}-m)^{-1}\} \{6(p+1)(m-1)\}^{-1} (2p^{2}+3p-1)$$

$$- \text{ Reject } H_{0} \text{ at level } \alpha \text{ when}$$

$$(1-u)M \geq \chi^{2}_{1-\alpha,p(p+1)(m-1)/2}$$

$$- p\text{-value}$$

$$1 - F_{\chi^{2}_{1-\alpha,p(p+1)(m-1)/2}} \{(1-u)M\}$$

Exercise 5.2: factors in producing plastic film (continued)
 Check the equality of covariance matrices for RATE="Low" and RATE="High"

```
install.packages('heplots')
options(digits = 4)
tear <- c(
   6.5, 6.2, 5.8, 6.5, 6.5, 6.9, 7.2, 6.9, 6.1, 6.3,
   6.7, 6.6, 7.2, 7.1, 6.8, 7.1, 7.0, 7.2, 7.5, 7.6
)
gloss <- c(
   9.5, 9.9, 9.6, 9.6, 9.2, 9.1, 10.0, 9.9, 9.5, 9.4,
   9.1, 9.3, 8.3, 8.4, 8.5, 9.2, 8.8, 9.7, 10.1, 9.2
)
opacity <- c(
   4.4, 6.4, 3.0, 4.1, 0.8, 5.7, 2.0, 3.9, 1.9, 5.7,
   2.8, 4.1, 3.8, 1.6, 3.4, 8.4, 5.2, 6.9, 2.7, 1.9
)
(X <- cbind(tear, gloss, opacity))
(rate <- factor(gl(2,10,length=nrow(X)), labels=c("Low", "High")))
result = heplots::boxM(lm(X-rate))
result$p.value</pre>
```

• Report: Testing hypotheses H_0 : the covariance matrix does not vary with the level of RATE v.s. H_1 : otherwise, we carried on the Box's M test and obtained .6743 as the p-value. So, at the .05 level, there was no strong statistical evidence against H_0 , i.e., we did not reject H_0 and believed that the covariance matrix does not vary with the level of RATE.

Two-way MANOVA (J&W Sec. 6.7)

- Model: $X_{ijk} = \mu + \tau_i + \beta_j + \gamma_{ij} + \mathbf{E}_{ijk}$ with $\mathbf{E}_{ijk} \stackrel{\text{iid}}{\sim} \text{MVN}_p(\mathbf{0}, \mathbf{\Sigma}), i = 1, \dots, m, j = 1, \dots, b, k = 1, \dots, n$
 - τ_i : the main effect of factor 1 at level i $-\beta_i$: the main effect of factor 2 at level j
 - $-\gamma_{ij}$: the interaction effect of factors 1 and 2 when their levels are i and j, respectively
 - Constraints for identifiability: $\sum_i \tau_i = \sum_j \beta_j = \sum_i \gamma_{ij} = \sum_j \gamma_{ij} = 0$
- Decomposition of total (corrected) SSP

$$SSP_{t} = SSP_{m1} + SSP_{m2} + SSP_{2fi} + SSP_{r}$$

- Total (corrected) SSP

$$\mathbf{SSP}_{\mathrm{t}} = \sum_{i=1}^{m} \sum_{j=1}^{b} \sum_{k=1}^{n} (\boldsymbol{X}_{ijk} - \bar{\boldsymbol{X}}_{\cdot\cdot\cdot}) (\boldsymbol{X}_{ijk} - \bar{\boldsymbol{X}}_{\cdot\cdot\cdot})^{\top}$$

* \bar{X} ... = $(mbn)^{-1} \sum_{i,j,k} X_{ijk}$ - SSP for main effect of factor 1

$$\mathbf{SSP}_{m1} = \sum_{i=1}^{m} bn(\bar{\boldsymbol{X}}_{i\cdots} - \bar{\boldsymbol{X}}_{\cdots})(\bar{\boldsymbol{X}}_{i\cdots} - \bar{\boldsymbol{X}}_{\cdots})^{\top}$$

* $\bar{\boldsymbol{X}}_{i\cdots} = (bn)^{-1} \sum_{j,k} \boldsymbol{X}_{ijk}$ - SSP for main effect of factor 2

$$\mathbf{SSP}_{\mathrm{m2}} = \sum_{j=1}^{b} mn(\bar{\boldsymbol{X}}_{\cdot j \cdot} - \bar{\boldsymbol{X}}_{\cdot \cdot \cdot})(\bar{\boldsymbol{X}}_{\cdot j \cdot} - \bar{\boldsymbol{X}}_{\cdot \cdot \cdot})^{\top}$$

* $\bar{X}_{.j.} = (mn)^{-1} \sum_{i,k} X_{ijk}$ - SSP for 2-factor-interaction (2fi)

$$\mathbf{SSP}_{2\mathrm{fi}} = \sum_{i=1}^{m} \sum_{j=1}^{b} n(\bar{\boldsymbol{X}}_{ij.} - \bar{\boldsymbol{X}}_{i..} - \bar{\boldsymbol{X}}_{.j.} + \bar{\boldsymbol{X}}_{...})(\bar{\boldsymbol{X}}_{ij.} - \bar{\boldsymbol{X}}_{i..} - \bar{\boldsymbol{X}}_{.j.} + \bar{\boldsymbol{X}}_{...})^{\top}$$

* $\bar{\boldsymbol{X}}_{ij.} = n^{-1} \sum_{k} \boldsymbol{X}_{ijk}$ - SSP for residual

$$\mathbf{SSP}_{\mathrm{r}} = \sum_{i=1}^{m} \sum_{j=1}^{b} \sum_{k=1}^{n} (oldsymbol{X}_{ijk} - ar{oldsymbol{X}}_{ij\cdot}) (oldsymbol{X}_{ijk} - ar{oldsymbol{X}}_{ij\cdot})^{ op}$$

- Testing interaction
 - Hypotheses $H_0: \gamma_{11} = \cdots = \gamma_{mb} = \mathbf{0}$ v.s. $H_1:$ otherwise
 - Wilk's lambda test statistic

$$\Lambda = \frac{\det \mathbf{SSP}_r}{\det(\mathbf{SSP}_r + \mathbf{SSP}_{2fi})}$$

* Under H_0 , by Bartlett's approximation

$$[\{p+1-(m-1)(b-1)\}/2-mb(n-1)]\ln\Lambda\approx\chi^2((m-1)(b-1))$$

- Reject H_0 at level α when

$$[\{p+1-(m-1)(b-1)\}/2-mb(n-1)]\ln\Lambda \ge \chi^2_{1-\alpha,(m-1)(b-1)}$$

- p-value

$$1 - F_{\chi^2((m-1)(b-1))}([\{p+1-(m-1)(b-1)\}/2 - mb(n-1)]\ln\Lambda)$$

- Testing main effects
 - Testing factor 1 main effects
 - * Hypotheses $H_0: \boldsymbol{\tau}_1 = \cdots = \boldsymbol{\tau}_m = \mathbf{0}$ v.s. $H_1:$ otherwise
 - * Wilk's lambda test statistic

$$\Lambda = \frac{\det \mathbf{SSP}_r}{\det(\mathbf{SSP}_r + \mathbf{SSP}_{m1})}$$

· Under H_0 , by Bartlett's approximation

$$[\{p+1-(m-1)\}/2-mb(n-1)]\ln\Lambda \approx \chi^2(m-1)$$

* Reject H_0 at level α when

$$[\{p+1-(m-1)\}/2-mb(n-1)]\ln\Lambda \ge \chi^2_{1-\alpha,m-1}$$

* p-value

$$1 - F_{\chi^2(m-1)}([\{p+1-(m-1)\}/2 - mb(n-1)]\ln\Lambda)$$

- Testing factor 2 main effects
 - * Hypotheses $H_0: \beta_1 = \cdots = \beta_b = \mathbf{0}$ v.s. $H_1:$ otherwise
 - * Wilk's lambda test statistic

$$\Lambda = \frac{\det \mathbf{SSP}_r}{\det(\mathbf{SSP}_r + \mathbf{SSP}_{26})}$$

· Under H_0 , by Bartlett's approximation

$$[\{p+1-(b-1)\}/2-mb(n-1)]\ln\Lambda\approx\chi^2(b-1)$$

* Reject H_0 at level α when

$$[\{p+1-(b-1)\}/2-mb(n-1)]\ln\Lambda \geq \chi^2_{1-\alpha,b-1}$$

* p-value

$$1 - F_{\chi^2(b-1)}([\{p+1-(b-1)\}/2 - mb(n-1)]\ln\Lambda)$$

- Exercise 5.3: factors in producing plastic film (continued)
 - One more factor ADDITIVE (amount of an additive, 2-level, low or high) in the production test

```
options(digits = 4)
tear <- c(
   6.5, 6.2, 5.8, 6.5, 6.5, 6.9, 7.2, 6.9, 6.1, 6.3,
   6.7, 6.6, 7.2, 7.1, 6.8, 7.1, 7.0, 7.2, 7.5, 7.6
)
gloss <- c(
   9.5, 9.9, 9.6, 9.6, 9.2, 9.1, 10.0, 9.9, 9.5, 9.4,
   9.1, 9.3, 8.3, 8.4, 8.5, 9.2, 8.8, 9.7, 10.1, 9.2
)
opacity <- c(
   4.4, 6.4, 3.0, 4.1, 0.8, 5.7, 2.0, 3.9, 1.9, 5.7,
   2.8, 4.1, 3.8, 1.6, 3.4, 8.4, 5.2, 6.9, 2.7, 1.9</pre>
```

```
(X <- cbind(tear, gloss, opacity))
(rate <- factor(gl(2,10,length=nrow(X)), labels=c("Low", "High")))
(additive <- factor(gl(2,5,length=nrow(X)), labels=c("Low", "High")))
summary(manova(X ~ rate*additive), test = 'Wilks')
summary(car::Manova(lm(X ~ rate*additive)), test.statistic='Wilks')</pre>
```