

Resampling Techniques and their Application

-Class 10-

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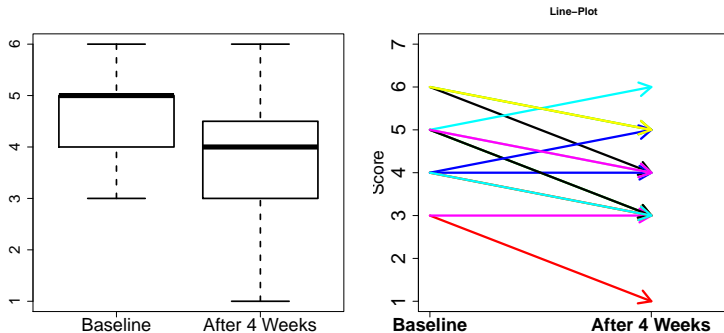
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Example

- Panic disorder longitudinal trial
 - Specific physical exercise therapy
 - $n = 15$ patients
 - Response: CGI-score **before and after** 4 weeks of treatment
 - 0 = patient is healthy, . . . , 6 = patient is critically ill



Example: Discussion

- What is the research question?
- What is the data scale?
- Are means appropriate?

Statistical Model

- $\mathbf{X}_k = (X_k, Y_k)', k = 1, \dots, n$
- $X_k \sim F_1, Y_k \sim F_2$
- F_i is any distribution
 - How to define an effect? ("Difference between the samples")
 - Means are not defined
 - Other options?

Nonparametric Effects

- $\mathbf{X}_k = (X_k, Y_k)', k = 1, \dots, n$
- $X_k \sim F_1, Y_k \sim F_2$
- Effect: Difference of the Medians
 - $\theta = \text{Median}(X) - \text{Median}(Y)$
 - Crude Measure
 - What else could we measure?

```
x=c(6,3,5,4,5,3,4,5,5,4,  
6,4,4,5,6)
```

```
y=c(4,1,3,4,6,3,3,4,3,3,  
5,5,3,4,5)
```

```
xM<-median(x)
```

```
xY<-median(y)
```

```
theta<- xM-xY
```

Nonparametric Effects

- $\mathbf{X}_k = (X_k, Y_k)', k = 1, \dots, n$
- $X_k \sim F_1, Y_k \sim F_2$
- Effect: $\% \{X \text{ is smaller than } Y\}$
 - $p = \text{Proportion of } X < Y$
 - Issue: Equal observations
 - $\%(X < Y) + \%(X > Y) + \%(X = Y) = 100\%$
 - Reasonable:

$$\%(X < Y) + \frac{1}{2}\%(X = Y)$$

```
#X<Y +1/2(X=Y)
```

```
grid1=expand.grid(x,y)
prop1 = (grid1[,1]<grid1[,2]) +
1/2*(grid1[,1]==grid1[,2])
p=mean(prop1)
p
```

```
#Y<X+1/2(X=Y)
prop2 = (grid1[,2]<grid1[,1]) +
1/2*(grid1[,2]==grid1[,1])
q=mean(prop2)
q
p+q
```

Statistical Model

- $\mathbf{X}_k = (X_1, Y_k)', k = 1, \dots, n$
- $X_k \sim F_1, Y_k \sim F_2$
- Effect and hypothesis
 - Relative marginal effect $p = P(X_1 < Y_2) + 0.5P(X_1 = Y_2)$
 - Interpretation
 - If $p < \frac{1}{2}$: X tends to be larger than Y
 - If $p = 1/2$: No tendency to smaller or larger values
 - Hypothesis:

$$H_0 : p = 1/2$$

Relative Effects: Other Examples

- Two samples: What is the chance, data in the first are smaller than in the second?
- What is the probability that men are taller than women?
- What is the probability, blood pressure under treatment is higher than under control?
- What is the probability, that scores after 4 weeks of treatment are smaller than at the beginning?
- **The relative effect thus measures, whether scores are smaller/larger between the groups**

Rel. Effect: Point Estimator

- **Ranks of the data**

- Sort the data
- Smallest number gets rank 1, largest gets rank $2n$
- Mid-ranks are used to adjust for ties
- In R: rank(x)

```
x <-c(5,3,2,1,4)
rank(x)
```

Two samples:

```
x=c(6,3,5,4,5,3,4,5,5,4,
6,4,4,5,6)
```

```
y=c(4,1,3,4,6,3,3,4,3,3,
5,5,3,4,5)
```

```
xy=c(x,y)
rank(xy)
```

Point Estimator

- Combine the two samples in a joint sample
- $X_1, \dots, X_n, Y_1, \dots, Y_n$
- Assign Ranks: $R_{x1}, \dots, R_{xn}, R_{y1}, \dots, R_{yn}$
- Compute

$$\bar{R}_{x.} = \frac{1}{n} \sum_{k=1}^n R_{xk} \quad \text{"Mean of the first"}$$

$$\bar{R}_{y.} = \frac{1}{n} \sum_{k=1}^n R_{yk} \quad \text{"Mean of the second"}$$

- Estimator:

$$\hat{p} = \frac{1}{2n} (\bar{R}_{y.} - \bar{R}_{x.}) + 1/2$$

```
n<-length(x)
```

```
xy<-c(x,y)  
rxy <- rank(xy)
```

```
mRx <-mean(rxy[1:n])  
mRy <-mean(rxy[(n+1):(2*n)])
```

```
phat<-1/(2*n)*(mRy-mRx)+1/2
```

Variance Estimator

- Need to estimate the variance of \hat{p}
- Needed: Ranks and internal ranks per sample
- Ranks internal (only x): $R_{x1}^{(x)}, \dots, R_{xn}^{(x)}$
- Ranks internal (only y): $R_{y1}^{(y)}, \dots, R_{yn}^{(y)}$
- Compute

$$Z_{xk} = \frac{1}{n} * (R_{xk} - R_{xk}^{(x)})$$

$$Z_{yk} = \frac{1}{n} * (R_{yk} - R_{yk}^{(y)})$$

$$D_k = Z_{xk} - Z_{yk}$$

- Estimator:

$$\hat{\sigma}^2 = \frac{1}{n-1} \sum_{k=1}^n (D_k - \bar{D})^2$$

```
xy<-c(x,y)
rxy <- rank(xy)
rx <-rank(x)
ry <-rank(y)

Z1k <- 1/n*(rxy[1:n]-rx)
Z2k <- 1/n*(rxy[(n+1):(2*n)]-ry)
Dk <- Z1k-Z2k
```

```
sigmahat <-var(Dk)
```

Test Procedures: Munzel's Test

- Test Statistic

$$T = \sqrt{n} \frac{(\hat{p} - 1/2)}{\hat{\sigma}}$$

- p-values and critical values are computed from a t -distribution with $n - 1$ degrees of freedom
- $(1 - \alpha)100\%$ Confidence Intervals:

$$\hat{p} \pm t_{1-\alpha/2}(n-1)\hat{\sigma}/\sqrt{n}$$

- Is the method applicable when samples are small?

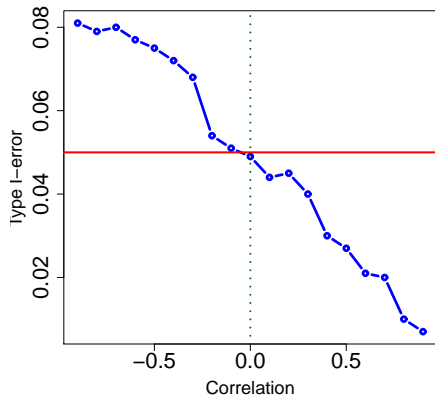
```
T=sqrt(n)*(phat-1/2)/sqrt(sigmahat)
pvalue=2*min(pt(T,n-1),1-pt(T,n-1))
```

```
crit <- qt(0.975,n-1)
SE <- sqrt(sigmahat)/sqrt(n)
```

```
Lower = phat -crit*SE
Upper = phat +crit*SE
```

Simulation: Munzel's test (1999)

- Type-1 error simulation ($\alpha = 5\%$, $n_{sim} = 10,000$)
- $\mathbf{X}_k = (X_{1k}, X_{2k}) \sim N(\mathbf{0}, \mathbf{V})$, $k = 1, \dots, 10$
- \mathbf{V} : compound symmetric (variance = 1, covariance = ρ)



Improve Munzel's Test

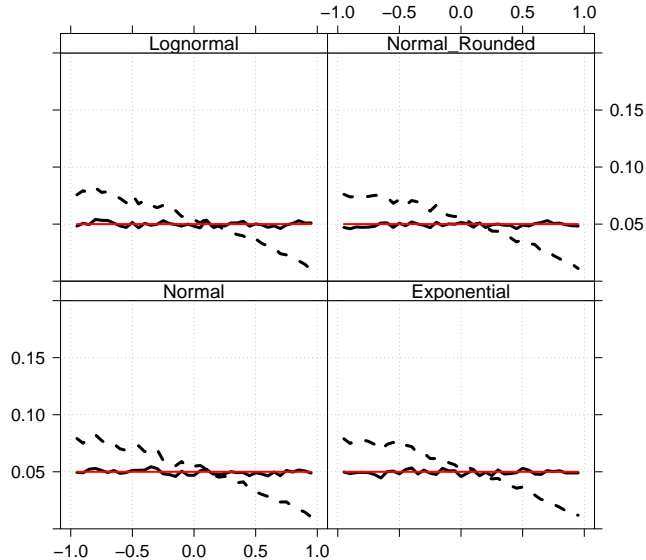
- How?
 - Explore Resampling Methods
 - For example, Wild-Bootstrap Approach
 - Permute data per subject
- Procedure
 - **Permute randomly X_{1k} and X_{2k} within each pair X_k**
 - Assign new ranks for permuted data
 - Compute $T^* = \sqrt{n}(\hat{p}^* - 1/2)/\hat{\sigma}^*$
 - Repeat this step several times
 - CI: $\hat{p} \pm z_{1-\alpha/2}^* \hat{\sigma} / \sqrt{n}$
- Project: Explore other resampling methods
- Note: Permuting data overall is not applicable in this model

Studentized Permutation Test - Theory

- Theoretically shown
 - Procedure is asymptotically exact under non-exchangeable data
 - Exact under exchangeability
 - Confidence intervals are asymptotically exact
 - Both procedures have equal power (asymptotically)
 - Reference: Konietzschke and Pauly (2012), A studentized permutation test for the Nonparametric Behrens-Fisher Problem in Paired Data

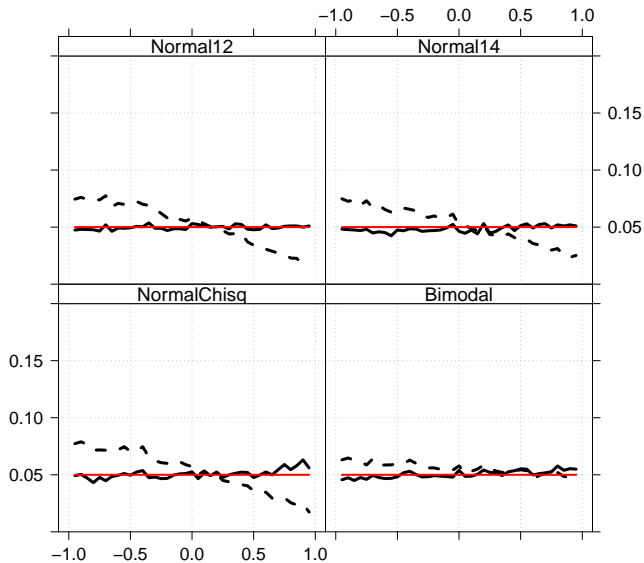
Simulations - Exchangeable data

Type - I Error = 5% (n=10)



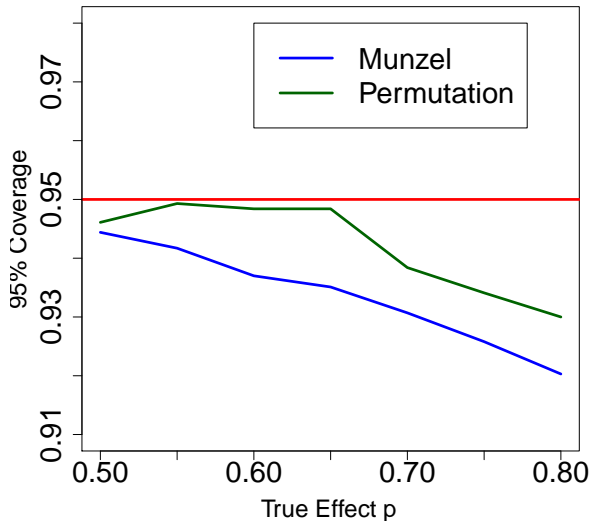
Simulations - Non Exchangeable data

Type - I Error = 5% (n=10)



Simulations - 95%-Coverage Probability

n=7; bivariate normal with rho=0



Example: Evaluation

- Point estimate $\hat{p} = .29$
 - Scores tend to smaller after 4 weeks than under baseline
- Hypothesis $H_0 : p = 1/2 \rightarrow p - value = .0006$
- 95%-CI: $[0.16; 0.43]$
- Munzel's test: similar results

Patient	PGI		Ranks	
k	Baseline	Week 4	Baseline	Week 4
1	6	4	28.5	14.0
2	3	1	5.5	1.0
3	5	3	22.5	5.5
4	4	4	14.0	14.0
5	5	6	22.5	28.5
6	3	3	5.5	5.5
7	4	3	14.0	5.5
8	5	4	22.5	14.0
9	5	3	22.5	5.5
10	4	3	14.0	5.5
11	6	5	28.5	22.5
12	4	5	14.0	22.5
13	4	3	14.0	5.5
14	5	4	22.5	14.0
15	6	5	28.5	22.5

Implementation

- Number of possible permutations: 2^n
- So, for small n , all possible permutations can be performed
- How?
- Need a matrix that contains all possible placement changes within each pair
- Notation

$$\mathbf{x}_k = (X_k, X_{k+n})', k = 1, \dots, n$$

- $n = 3$

$$\mathbf{x}_k = \begin{pmatrix} X_1 & X_4 \\ X_2 & X_5 \\ X_3 & X_6 \end{pmatrix} \Rightarrow \mathbf{x}_k = (X_1, X_2, X_3, X_4, X_5, X_6)'$$

Implementation- II

- Now, generate a matrix...
 - that has 1 and $(n + 1)$ how often in the first row and in which order?
 - that has 2 and $(n + 2)$ how often in the second row and in which order?
 - that has 3 and $(n + 3)$ how often in the second row and in which order?
 -

Implementation- III

 $n=4$ [illegible]

```

permuall<-function(nsim, rho, n){
#-----Daten einlesen-----#
n1<-n+1
n2<-2*n
x<-matrix(0, ncol=nsim, nrow=n2)
for (h in 1:nsim){
x11<-rnorm(n)
x22<-rho*x11+sqrt(1-rho^2)*(rnorm(n))
x[,h]<-c(x11,x22)}

#-----Brunner-Munzel test-----#
tcrit<-qt(0.975,n-1)
x1<-x[1:n,]
x2<-x[n1:n2,]
rx<-apply(x, 2, rank)
rx1<-rx[1:n,]
rx2<-rx[n1:n2,]
rix1<-apply(x1, 2, rank)
rix2<-apply(x2, 2, rank)
BM1<-1/n*(rx1-rix1)
BM2<-1/n*(rx2-rix2)
BM3<-BM1-BM2
BM4<-1/(2*n)*(rx1 - rx2)
pd<-colMeans(BM2)
m<-colMeans(BM3)
v<-(colSums(BM3^2)-n*m^2)/(n-1)
v0<-(v==0)
v[v0]<-1/n
T<-sqrt(n)*(pd-1/2)/sqrt(v)

```



```
#-----Studentized Permutation Test---#
```

```
nperm<-2^n
```

```
if(nperm <10000){
```

```
  p<-0
```

```
  for (i in 1:n){
```

```
    a<-rep(c(rep(c(i,i+n),nperm/(2^i)),rep(c(i+n,i),nperm/(2^i)))),2^(i-1))
```

```
    p<-rbind(p,a)
```

```
  }
```

```
  p<-p[2:(n+1),]
```

```
  P<-matrix(p,ncol=nperm)}
```

```
if (nperm >=10000){
```

```
  nperm=10000
```

```
  P<-matrix(0,nrow=(2*n),ncol=nperm)
```

```
  for (h in 1:nperm){
```

```
    P[,h]<-c(t(apply(cbind(1:n,(n+1):(2*n)),1,sample))))}
```

```

#-----Beginn der Simulationsschleife----#
BM=PERM=c()
for (s in 1:nsim){
  xs<-x[,s]
  rs<-rx[,s]
  #-----Permutationstest-----#
  xperm<-matrix(xs[P],nrow=n2,ncol=nperm)
  rxperm<-matrix(rs[P],nrow=n2,ncol=nperm)
  xperm1<-xperm[1:n,]
  xperm2<-xperm[n1:n2,]
  rxperm1<-rxperm[1:n,]
  rxperm2<-rxperm[n1:n2,]
  riper1<-apply(xperm1,2,rank)
  riper2<-apply(xperm2,2,rank)
  BMperm2<-1/n*(rperm2-riperm2)
  BMperm3<-1/n*(rperm1-riperm1)-BMperm2
  pdperm<-colMeans(BMperm2)
  mperm3<-colMeans(BMperm3)
  vperm3<-(colSums(BMperm3^2)-n*mperm3^2)/(n-1)
  vperm30<-(vperm3==0)
  vperm3[vperm30]<-1/n
  Tperm<-sqrt(n)*(pdperm-1/2)/sqrt(vperm3)
  p1perm<-mean(Tperm<=T[s]); p2perm<-mean(Tperm>=T[s])
  pperm<-2*min(p1perm,p2perm)
  PERM[s]<-(pperm<0.05)}
ergebnis<-data.frame(nsim=nsim,
  nperm=nperm,n=n,rho=rho,BM= mean(abs(T)>t crit),
  PERM=mean(PERM))
ergebnis}
permua11(100,0,7)

```