















Multiplicity issues

- Co-primary
 - ullet Test both hypotheses simultaneously at overall significance level α
- Hierarchy:
 - Test first hypothesis at significance level α
 - \blacksquare If first hypothesis is rejected then test second hypothesis at significance level α
 - Else: second hypothesis is not rejected

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Simulation of single study

```
# assumption
group_size=100
base_a1c=8.25
sd_a1c=0.5
effect_B=-0.2
effect_C=-0.6
r=0.6
sd_noise=0.02
# start with 3*group_size subjects
# for each of them simulate alc at month 0 and month 6, assuming no effect
# covariance matrix of alc measures at months 0 and 6
var_a1c=sd_a1c^2
covmat=matrix(c(var_alc, r*var_alc, r*var_alc, var_alc), nrow=2)
pats=mvrnorm(n=3*group_size, mu=rep(base_a1c,2), Sigma=covmat)
pats=as.data.frame(pats)
names(pats)=c("month0", "month6")
pats$study_group=""
```

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```
# set study groups and effects
range=1:group_size
pats$study_group[range]="A" # control group, no effect, except noise
pats$month6[range]=pats$month6[range]+rnorm(group_size, mean=0, sd=sd_noise)

# group B
range=(group_size+1):(2*group_size)
pats$study_group[range]="B"
pats$month6[range]=pats$month6[range]+rnorm(group_size, mean=effect_B, sd=sd_noise)

# sensors+model group with specified effect
range=(2*group_size+1):(3*group_size)
pats$study_group[range]="C" # sensors+model group, effect=effect!
pats$month6[range]=pats$month6[range]+rnorm(group_size, mean=effect_C, sd=sd_noise)

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```

```
Analysis of simulated study
  # ancova + ci
  \label{eq:model} $$ \bmod = \lim(month6 \sim month0 + study\_group, data=pats) $$ model\_emeans=emmeans(model, "study\_group") $$
  model_contrast=contrast(model_emeans,
                                list("AB" = c(1, -1, 0),
"AC" = c(1, 0, -1))
  model_contrast=as.data.frame(model_contrast)
 > model_emeans
  study_group emmean
                           SE df lower.CL upper.CL
                8.34 0.0399 296 8.26 8.42
                  8.11 0.0399 296
                                         8.04
                                                    8.19
                  7.66 0.0399 296
                                         7.58
                                                    7.74
 Confidence level used: 0.95
 > model_contrast
         rast estimate SE df t.ratio p.value
AB 0.2258135 0.05641718 296 4.002566 7.926214e-05
   contrast estimate
          AC 0.6786655 0.05635753 296 12.042145 1.891710e-27
 2
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```

```
Testing the hypotheses
# parallel testing
adjusted_pvalues=p.adjust(p=model_contrast$p.value, method="hochberg")
reject1=as.numeric(adjusted_pvalues[1]<0.05)</pre>
reject2=as.numeric(adjusted_pvalues[2]<0.05)</pre>
# hierarchical testing
if(model\_contrast\$p.value[1]>0.05)\{\\
  reject1=0
  reject2=0
} else {
  reject1=1
  if(model_contrast$p.value[2]>0.05){
    reject2=<mark>0</mark>
  } else {
    reject2=1
  }
}
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```