# Repeated Measurements in Clinical Studies

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**Exercise Session One** 



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#### Materials for the Session

- The course materials contain 5 sets of exercises, each has two parts
  - Part one: guided tour through an exercise with a complete solution
  - Part two: just an assignment without a solution
- All sas files for today are located in the directory //Day one
- → We will focus on Part two, leaving Part one as a home exercise
- → Please find
  - Exercise one Part two
  - exercise1\_part2\_code.sas



#### Goals of This Exercise Session

- (1) Getting familiar with the SAS system (loading and plotting data, summary statistics etc.)
- (2) Understanding the complexity of repeated measurements data
- (3) Performing data exploration to get a better feel for the data
- (4) Trying out simple methods to analyze longitudinal data
- (5) Understanding the limitations of such simple approaches



#### The Hipfracture Data

- --- 60 patients recovering after a hip fracture operation
- → 5 examinations at days 1, 3, 5, 8 and 12 after the operation
- OUTCOME Mini-Mental State Examination score (MMSE): number of correct answers to 30 questions
- **→ BASELINE CHARACTERISTICS** 
  - age: continuous
  - neuro-status: binary (0/1 for neuro-psychiatric/non-neuro-psychiatric before the operation)



# The Objective of the Investigator

How does the MMSE score **evolve** over time and how this **evolution** depends on the neuro-status and age?



### Getting the Feel for the Data

- → Data exploration is a necessary pre-modeling step!
  - Check for missing values
  - Check for outliers
  - Computing simple summary statistics (mean, standard deviation)
  - Deviations from model assumptions (normality, homoscedasticity)

:



## Getting Started with SAS

- Scripting language to perform statistical analysis
- --- Many different ways to load the data

```
data hipfracture;
input NEURO MMSE1 MMSE3 MMSE5 MMSE8 MMSE12 IDNR AGE;
cards;
         2.8
                  2.8
                            2.8
                                      2.6
                                               2.5
                                                                   74
         25
                  25
                            23
                                      27
                                                                   67
         2.6
                  2.9
                            29
                                      2.7
                                                                   67
         24
                   20
                            26
                                      29
                                               25
                                                         59
                                                                   83
         2.6
                   2.8
                            2.5
                                      30
                                               2.7
                                                                   77
                                                         60
run;
```

→ This is a horizontal data format



### **Summary Statistics**

→ How many neuro-psychiatric patients are in the data?

```
proc freq data=hipfracture;
  table NEURO;
run;
```

Compute means of the responses and age over time for the two neuro groups

```
proc sort data=hipfracture;
by neuro;
run;

proc means data=hipfracture Q1 median mean  std Q3 maxdec=2 n;
by NEURO;
var MMSE1 MMSE3 MMSE5 MMSE8 MMSE12 AGE;
run;
```



### Plot Histograms

- → Check for symmetry/skewness in the MMSE distribution.
- Normality in the responses would be nice.

```
proc univariate data=hipfracture;
var NEURO MMSE1 MMSE3 MMSE5 MMSE8 MMSE12 AGE;
histogram MMSE1 MMSE3 MMSE5 MMSE8 MMSE12 / normal;
run;
```

Separate histograms for the two NEURO groups?



# **Summary Statistics**

- → How many missing observations there are in the data for each response variable?
- Can we already tell something about the effect of NEURO on the mean evolution of MMSE over time?
- → How about the effect of AGE (a continuous variable)?



#### **Summary Statistics**

Categorize AGE based on quartiles and get the summary statistics of MMSE in the 4 categories

```
data hipfracture;
set hipfracture;
agecat=.;
if (AGE<72) then agecat=1:
if (AGE>=72) and (AGE<80) then agecat=2;
if (AGE>=80) and (AGE<84) then agecat=3;
if (AGE>=84) then agecat=4;
run;
proc sort data=hipfracture;
by agecat;
run;
proc means data=hipfracture Q1 median mean std Q3 maxdec=2;
by agecat:
var NEURO MMSE1 MMSE3 MMSE5 MMSE8 MMSE12 AGE;
                                                       Erasmus MO
run;
```

## Conversion to the Long Format

- → We proceed by visualizing MMSE evolution over time
- → It is easier to convert the data in the long format
- We use a SAS macro makelong (alternative ways in Part One of Exercise 1)

```
proc sort data=hipfracture;
by IDNR;
run;

%makelong(data=hipfracture,
out=hip_long,
id=IDNR,
root=MMSE,
copy=NEURO AGE agecat,
measurement=TIMECLSS)

proc print data=hip_long
;
run;
```



## Plotting the Mean Evolution

Mean MMSE evolution over time, overall and in relation to AGE and NEURO

```
proc gplot data=hip_long;
goptions reset=all ftext=swiss;
plot MMSE*TIMECLSS;
symbol c=blue i=stdlmjt l=1 w=2;
run; quit;

proc gplot data=hip_long;
plot MMSE*TIMECLSS=NEURO;
symbol1 c=red i=stdlmjt l=1 w=2;
symbol2 c=blue i=stdlmjt l=1 w=2;
run; quit;
```

→ Does MMSE decrease over time? Is the behavior of MMSE different in the two NEURO groups?



### Plotting the Individual Evolution

Individual MMSE evolution over time (separately based on NEURO)

- Are the individual curves wiggly? Does the wigglyness differ in the two NEURO groups?
- → Which NEURO group has a higher "within-subject" variability?



## **Exploring the Correlation**

- → Repeated measurements are likely to be correlated!
- Pearson correlation matrix (separately for the two NEURO groups)
- Again we use the horizontal data format

```
proc sort data=hipfracture;
by neuro;
run;

proc corr data=hipfracture;
by NEURO;
var MMSE1 MMSE3 MMSE5 MMSE8 MMSE12;
run;
```

Are there higher correlations in the NEURO group with less wiggly individual profiles?

### Simple Methods for Longitudinal Data

#### (a) Separate analysis for every day

→ What is an appropriate model for the analysis?

```
proc glm data = hipfracture;
model MMSE1 = AGE NEURO;
run;
proc glm data = hipfracture;
model MMSE3 = AGE NEURO:
run;
proc glm data = hipfracture;
model MMSE5 = AGE NEURO;
run;
proc glm data = hipfracture;
model MMSE8 = AGE NEURO;
run;
proc glm data = hipfracture;
model MMSE12 = AGE NEURO;
run;
```

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what are the conclusions and disadvantages of this approach?

## Simple Methods for Longitudinal Data

#### (b) Analysis of Increments

Linear regression outcome: difference MMSE12-MMSE1

```
data hipfracture;
set hipfracture;
increment=MMSE12-MMSE1;
run;
proc glm data = hipfracture;
model increment = AGE NEURO;
run;
```

- → What is the conclusion?
- → What are the disadvantages of this approach?



### Simple Methods for Longitudinal Data

- (c) Analysis of an Endpoint with an Additional Covariate
- → Covariates: NEURO, AGE, MMSE1

```
proc glm data = hipfracture;
model MMSE12 =MMSE1 AGE NEURO;
run;
```

- Are the results consistent with analysis of increments?
- How does the interpretation of the model change as compared to the analysis of increments?



#### Summary

- (1) We learned simple SAS commands for data exploration
  - proc freq
  - proc univariate
  - proc means
  - proc corr
- (2) We learned how to convert data into a long format
- (3) We learned how to plot mean and individual evolutions
  - proc sgplot
- (4) We learned simple methods for repeated measurements
  - proc glm
- (5) We learned that more elaborate methods are needed to make an efficient use of the data!

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# Thank you for your attention!

Questions?

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