

Re-introduction to openMx

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Starting at the beginning...

- Data preparation
 - The algebra style used in Mx expects 1 line per case/family
 - (Almost) limitless number of families and variables
 - Data needs to be read into R before it can be analysed
 - (the commands to read the data can be nested within the R script)
 - Default missing code is now **NA**

Getting your data into R

- Example data: ozbmi2.txt

Fam	agecat	age	zyg	part	wt1	wt2	ht1	ht2	htwt1	htwt2	bmi1	bmi2
115	0	0.21	1	2	58	57	1.7	1.7	20.0692	19.7232	20.9943	20.8726
121	0	0.24	1	2	54	53	1.6299	1.6299	20.3244	19.9481	21.0828	20.9519
158	0	0.21	1	2	55	50	1.6499	1.6799	20.202	17.7154	21.0405	20.121
172	0	0.21	1	2	66	76	1.5698	1.6499	26.7759	27.9155	23.0125	23.3043
182	0	0.19	1	2	50	48	1.6099	1.6299	19.2894	18.0662	20.7169	20.2583
199	0	0.26	1	2	60	60	1.5999	1.5698	23.4375	24.3418	22.0804	22.3454

- `data<-read.table("ozbmi2.txt", header=T, na.strings = "NA")`
- `head(data)`

```
> head(data)
  fam agecat  age zyg part wt1 wt2  ht1  ht2  htwt1  htwt2  bmi1  bmi2
1 115     0 0.21  1   2  58  57  1.7  1.7 20.0692 19.7232 20.9943 20.8726
2 121     0 0.24  1   2  54  53 1.6299 1.6299 20.3244 19.9481 21.0828 20.9519
3 158     0 0.21  1   2  55  50 1.6499 1.6799 20.202 17.7154 21.0405 20.121
4 172     0 0.21  1   2  66  76 1.5698 1.6499 26.7759 27.9155 23.0125 23.3043
5 182     0 0.19  1   2  50  48 1.6099 1.6299 19.2894 18.0662 20.7169 20.2583
6 199     0 0.26  1   2  60  60 1.5999 1.5698 23.4375 24.3418 22.0804 22.3454
```

Selecting and sub-setting data

- Make separate data sets for the MZ and DZ

```
> mzData <- as.data.frame(subset(data, zyg<3, c(bmi1,bmi2)))
> dzData <- as.data.frame(subset(data, zyg>2, c(bmi1,bmi2)))
> head(dzData)
      bmi1    bmi2
843 21.9642     NA
844 21.8791 21.2112
845 22.2321 22.6044
846 19.8491 20.1743
847 20.1743     NA
848 21.7050 21.2905
```

- Check data is numeric and behaves as expected

```
> cov(mzData, use="complete")
      bmi1    bmi2
bmi1 0.8779390 0.6734489
bmi2 0.6734489 0.8987715
> cov(dzData, use="complete")
      bmi1    bmi2
bmi1 0.8908474 0.2872594
bmi2 0.2872594 0.8657751
> colMeans(mzData, na.rm=TRUE)
      bmi1    bmi2
21.75089 21.73471
> colMeans(dzData, na.rm=TRUE)
      bmi1    bmi2
21.68689 21.88095
```

Common problem

- Problem: data contains a non numeric value

```
> cov(dzData, use="complete")
           bmi1      bmi2
bmi1 0.8908474 0.2872594
bmi2 0.2872594 0.8657751
Warning message:
In cov(dzData, use = "complete") : NAs introduced by coercion
> colMeans(mzData, na.rm=TRUE)
Error in colMeans(mzData, na.rm = TRUE) : 'x' must be numeric
> colMeans(dzData, na.rm=TRUE)
Error in colMeans(dzData, na.rm = TRUE) : 'x' must be numeric
```

20171	1	0.35	5	2	51	79	1.5999	1.7998	19.9219	24.3827	20.9427	22.3571
20188	1	0.37	5	2	53	65	1.5698	1.73	21.5019	21.7181	21.477	21.547
20204	1	0.53	5	1	58	64	1.6299	NA	21.83	NA	A	NA
20390	1	0.37	5	2	64	73	1.6499	1.8298	23.5078	21.7982	22.1013	21.5728
20398	1	0.52	5	2	60	77	1.6299	1.73	22.5827	25.7276	21.8203	22.7329

- Equivalent Mx Classic error - *Uh-oh... I'm having trouble reading a number in D or E format*

Important structural stuff

- openMx has a very fluid and flexible structure
- Each code snippet is being saved as a variable
- We tend to reuse the variable names in our scripts
- This makes it very important to create a new project for each series of analyses
- Remember the project also contains the data so these files can become very large.

Matrices are the building blocks

```
mxMatrix( type="Lower", nrow=nv, ncol=nv, free=TRUE,  
values=.6, label="a11", name="a" ), #X
```

- Many types eg. type="Lower"
- Denoted by names eg. name="a"
- Size eg. nrow=nv, ncol=nv
- All estimated parameters must be placed in a matrix & Mx must be told what type of matrix it is

Matrices are the building blocks

- Many types

```
mxMatrix( type="Zero", nrow=2, ncol=3, name="a" )
```

```
0 0 0  
0 0 0
```

```
mxMatrix( type="Unit", nrow=2, ncol=3, name="a" )
```

```
1 1 1  
1 1 1
```

```
mxMatrix( type="Ident", nrow=3, ncol=3, name="a" )
```

```
1 0 0  
0 1 0  
0 0 1
```

```
mxMatrix( type="Diag", nrow=3, ncol=3, free=TRUE, name="a" )
```

```
? 0 0  
0 ? 0  
0 0 ?
```

```
mxMatrix( type="Sdiag", nrow=3, ncol=3, free=TRUE, name="a" )
```

```
0 0 0  
? 0 0  
? ? 0
```

```
mxMatrix( type="Stand", nrow=3, ncol=3, free=TRUE, name="a" )
```

```
1 ? ?  
? 1 ?  
? ? 1
```

```
mxMatrix( type="Symm", nrow=3, ncol=3, free=TRUE, name="a" )
```

```
? ? ?  
? ? ?  
? ? ?
```

```
mxMatrix( type="Lower", nrow=3, ncol=3, free=TRUE, name="a" )
```

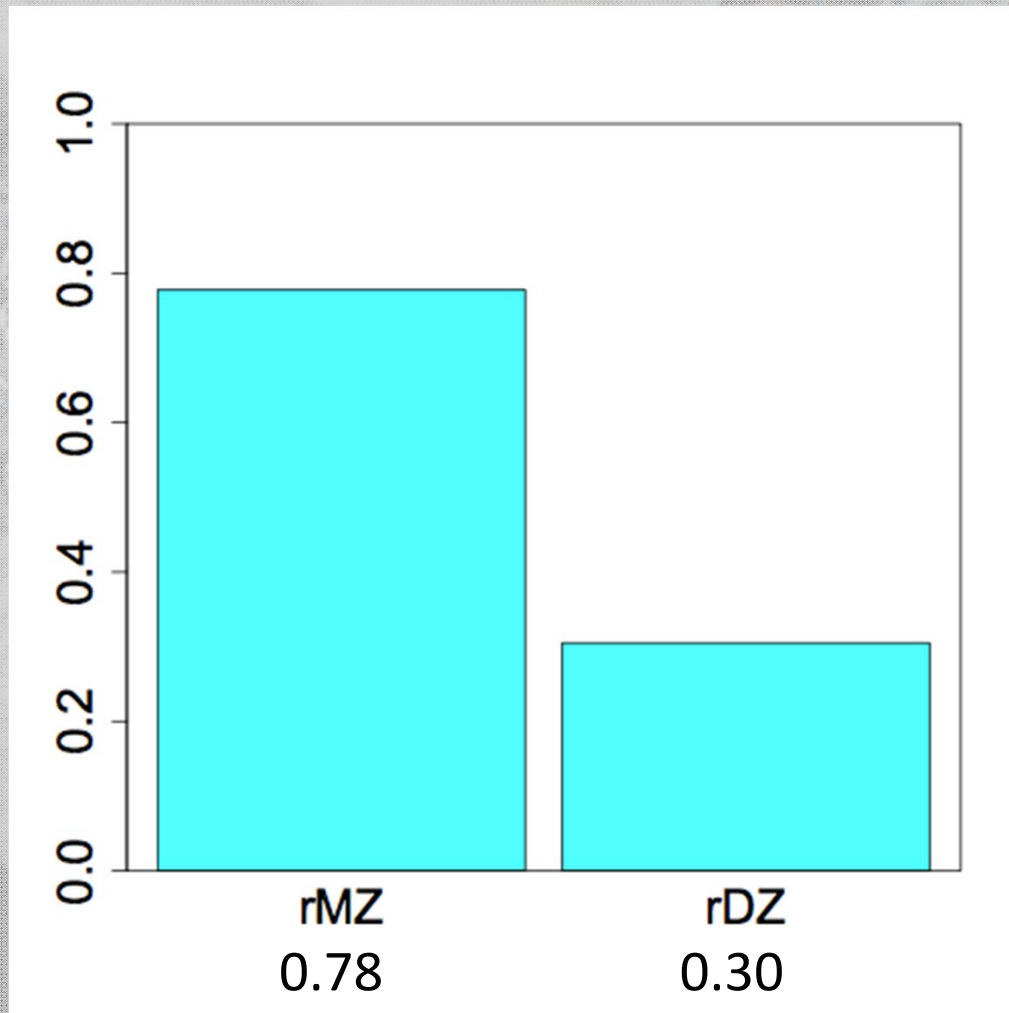
```
? 0 0  
? ? 0  
? ? ?
```

```
mxMatrix( type="Full", nrow=2, ncol=4, free=TRUE, name="a" )
```

```
? ? ? ?  
? ? ? ?
```

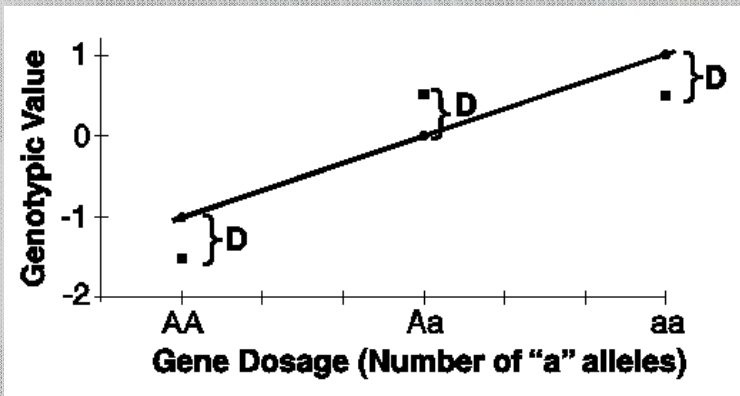
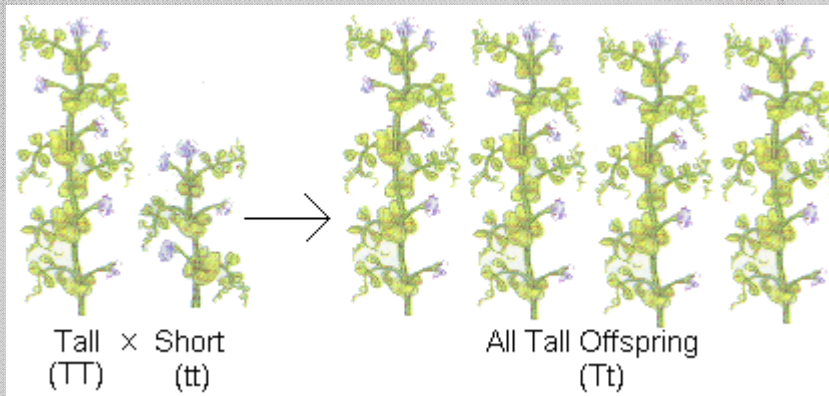

Yesterday we ran an ADE Model

- Why?



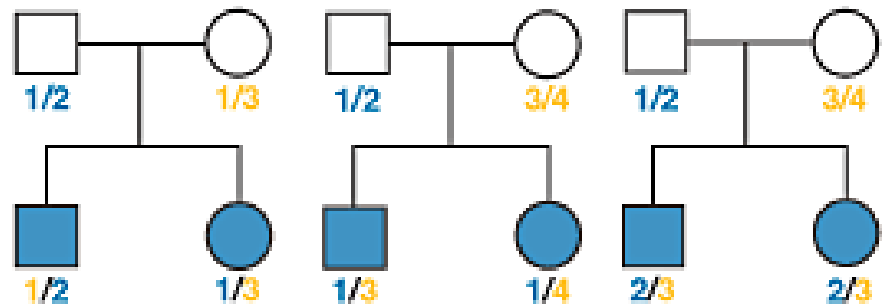
What is D again?

- Dominance refers to non-additive genetic effects resulting from interactions between alleles at the same locus or different loci (epistasis)



What is D again?

- DZ twins/full siblings share
 - ~50% of their segregating DNA &
 - for ~25% loci they share not only the genotype but also the parental origin of each allele



- DZ twins/full siblings share

This is where the .5A comes from

- ~50% of their segregating DNA &
- for ~25% loci they share not only the genotype but also the parental origin of each allele

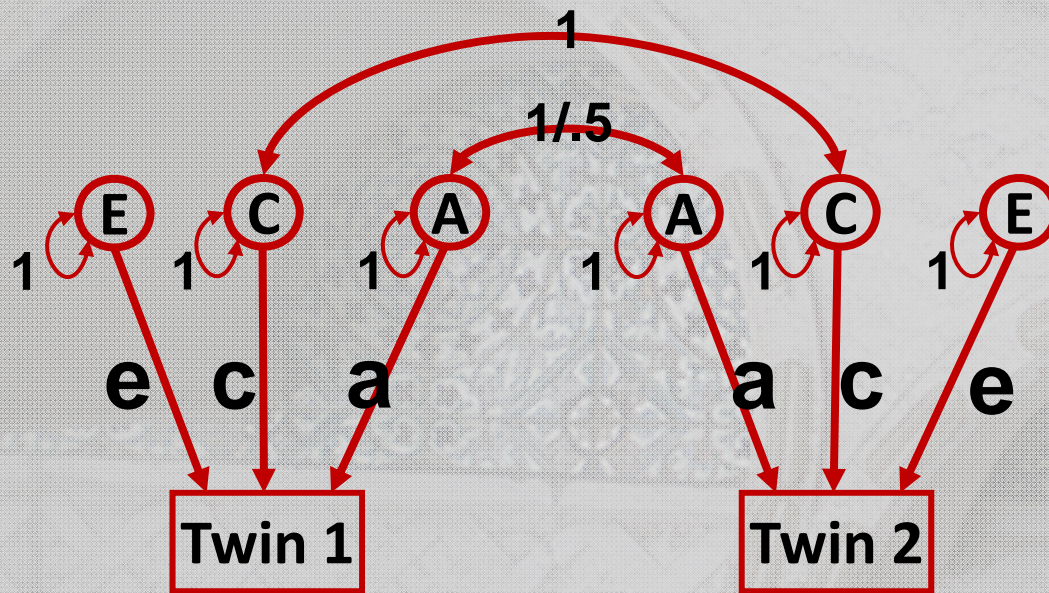
Consider a mating between mother AB x father CD:

		Sib1			
		AC	AD	BC	BD
Sib2	AC	2	1	1	0
	AD	1	2	0	1
	BC	1	0	2	1
	BD	0	1	1	2

This is where the .25D comes from

IBD 0 : 1 : 2 = 25% : 50% : 25%

Today we will run an ACE model



MZ

$$a^2 + c^2 + e^2$$

$$a^2 + c^2$$

$$a^2 + c^2$$

$$a^2 + c^2 + e^2$$

DZ

$$a^2 + c^2 + e^2$$

$$.5a^2 + c^2$$

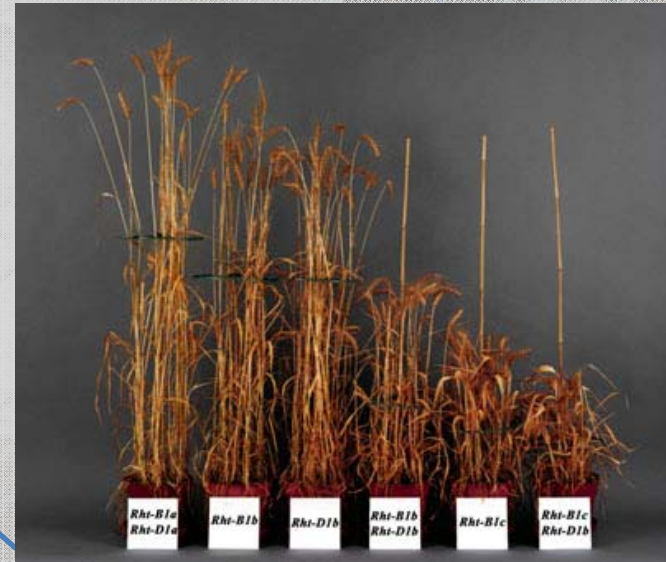
$$.5a^2 + c^2$$

$$a^2 + c^2 + e^2$$

Today we will run an ACE model

Additive genetic effects

- Why is the coefficient for DZ pairs .5?
- Average genetic sharing between siblings/DZ twins



		Sib1			
		AC	AD	BC	BD
Sib 2	AC	2	1	1	0
	AD	1	2	0	1
	BC	1	0	2	1
	BD	0	1	1	2

MZ

$$a^2+c^2+e^2$$

$$a^2+c^2$$

$$a^2+c^2$$

$$a^2+c^2+e^2$$

DZ

$$a^2+c^2+e^2$$

$$.5a^2+c^2$$

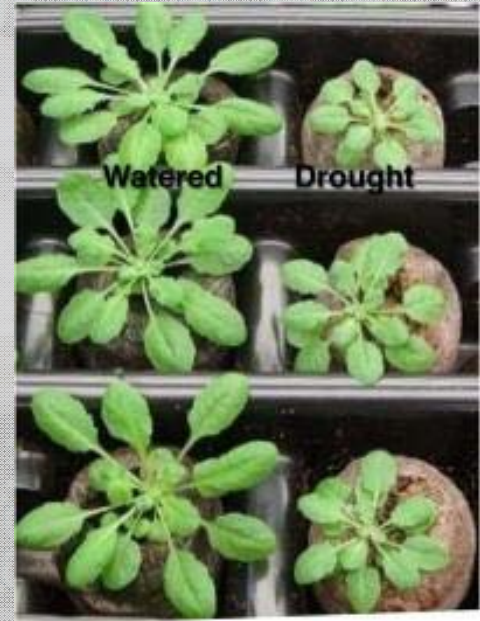
$$.5a^2+c^2$$

$$a^2+c^2+e^2$$

Today we will run an ACE model

Common environmental effects

- Coefficient =1 for MZ and DZ pairs
- Equal environment assumption – for all the environmental influences THAT MATTER there is ON AVERAGE no differences in the degree of environmental sharing between MZ and DZ pairs



MZ

$$a^2+c^2+e^2$$

$$a^2+c^2$$

$$a^2+c^2$$

$$a^2+c^2+e^2$$

DZ

$$a^2+c^2+e^2$$

$$.5a^2+c^2$$

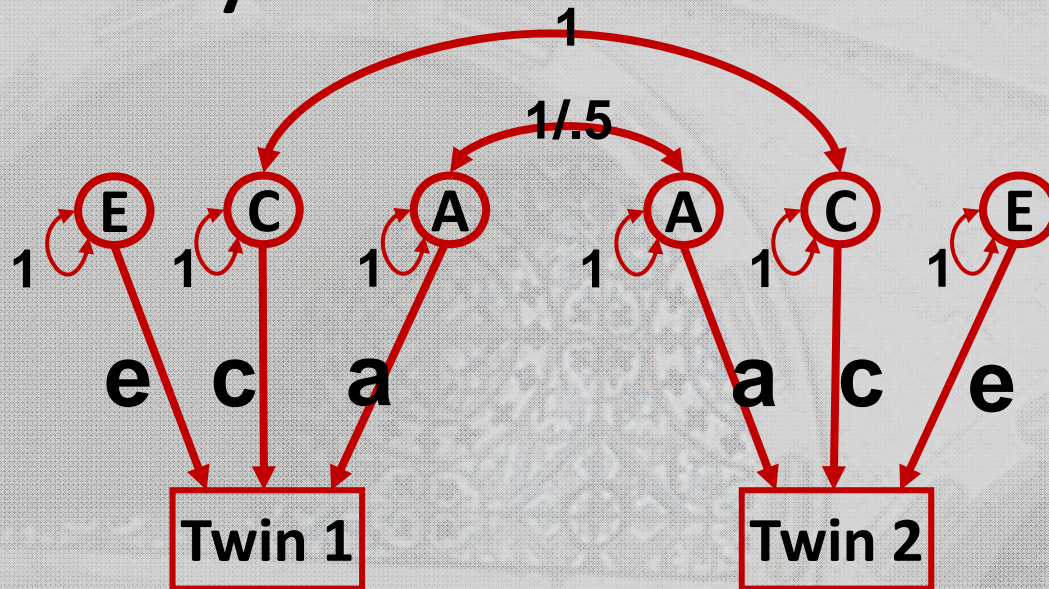
$$.5a^2+c^2$$

$$a^2+c^2+e^2$$

Today we will run an ACE model

- Open RStudio
- `faculty/sarah/tues_morning`
- Copy everything

Today we will run an ACE model



```
pathA    <- mxMatrix( type="Full", nrow=nv, ncol=nv, free=TRUE,
                      values=.6, label="a11", name="a" )
pathC    <- mxMatrix( type="Full", nrow=nv, ncol=nv, free=TRUE,
                      values=.6, label="c11", name="c" )
pathE    <- mxMatrix( type="Full", nrow=nv, ncol=nv, free=TRUE,
                      values=.6, label="e11", name="e" )

meanG    <- mxMatrix( type="Full", nrow=1, ncol=ntv, free=TRUE,
                      values= 20, label="mean", name="expMean" )
```

Today we will run an ACE model

```
COVA      <- mxAlgebra( a %**% t(a), name="A" )
COVC      <- mxAlgebra( c %**% t(c), name="C" )
COVE      <- mxAlgebra( e %**% t(e), name="E" )
```

```
covMZ     <- mxAlgebra(
  rbind( cbind(A+C+E , A+C),
         cbind(A+C   , A+C+E)), name="expCovMZ" )
```

```
covDZ     <- mxAlgebra(
  rbind( cbind(A+C+E, 0.5%x%A+C),
         cbind(0.5%x%A+C , A+C+E)), name="expCovDZ" )
```

MZ

$$a^2+c^2+e^2$$

$$a^2+c^2$$

$$a^2+c^2$$

$$a^2+c^2+e^2$$

DZ

$$a^2+c^2+e^2$$

$$.5a^2+c^2$$

$$.5a^2+c^2$$

$$a^2+c^2+e^2$$

Data	
dzData	351 obs. of 2 variables
mzData	569 obs. of 2 variables
twinData	3808 obs. of 12 variables
Values	

```

# Data objects for Multiple Groups
dataMZ  <- mxData( observed=mzData, type="raw" )
dataDZ  <- mxData( observed=dzData, type="raw" )

# Objective objects for Multiple Groups
objMZ   <- mxFIMLObjective( covariance="expCovMZ",
                             means="expMean", dimnames=selVars )

objDZ   <- mxFIMLObjective( covariance="expCovDZ",
                             means="expMean", dimnames=selVars )

```

To fit a model to data, the differences between the observed covariance matrix and model-implied expected covariance matrix are minimized.

Objective functions are functions for which free parameter values are chosen such that the value of the objective function is minimized.

`mxFIMLObjective()` uses full-information maximum likelihood to provide maximum likelihood estimates of free parameters in the algebra defined by the covariance and means arguments.

```
parameters <- list( pathA, pathC, pathE, covA, covC, covE )  
modelMZ <- mxModel( parameters, meanG, covMZ, dataMZ, objMZ,  
                    name="MZ" )  
modelDZ <- mxModel( parameters, meanG, covDZ, dataDZ, objDZ,  
                    name="DZ" )  
minus2ll <- mxAlgebra( MZ.objective + DZ.objective,  
                      name="m2LL" )  
obj <- mxAlgebraObjective( "m2LL" )  
aceModel <- mxModel( "ACE", parameters, modelMZ, modelDZ,  
                    minus2ll, obj )
```

This models requires
path parameters,
means, covariance,
data and objectives


Automatic naming – you don't
need to predefine this

```
# Run ACE model  
aceFit <- mxRun(aceModel)  
aceSumm <- summary(aceFit)  
aceSumm
```

Submodels

```
# Run AE model
aeModel  <- mxModel( aceFit, name="AE" )
aeModel  <- omxSetParameters( aeModel, labels="c11",
                              free=FALSE, values=0 )

aeFit    <- mxRun(aeModel)
round(aeFit@output$estimate,4)
```



Pickup the previously prepared model
Edit as required
Rerun and compare

Saving your output

- Save the R workspace
 - On closing click yes
 - Very big
 - Saves everything
- Save the fitted model
 - Equivalent to save in classic Mx
 - `save(univACEFit, file="test.omxs")`
 - `load("test.omxs")` – need to load OpenMx first

What to report

- Summary statistics
 - Usually from a simplified ‘saturated’ model
- Standardized estimates
 - Easier to conceptualise
 - ie 40% of the phenotypic variance vs a genetic effect of 2.84
 - Can easily be returned to original scale if summary statistics are provided

What to report

- Path coefficients
 - Very important in multivariate analyses
 - Gives a much clearer picture of the directionality of effects
- Variance components/proportion of variance explained
- Genetic correlations

General Advice/Problem solving

- Scripting styles differ
- Check the sample description
- Learn to love the webpage
- Comments are your friends

Bus shelter on the road to
Sintra (Portugal)

