A couple dozen functions suffice to carry out your work in Introduction to Statistical Modeling. This sheet provides the names of functions, a review of formula syntax, and some examples of use.

## Help

```
help()
apropos()
?
??
example()
```

### **Arithmetic**

Basic arithmetic is very similar to a calculator.

```
# basic ops: + - * / ^ ( )
log()
exp()
sqrt()
log10()
abs()
```

# Randomization/Iteration

```
do()  # mosaic
sample() # mosaic augmented
resample() # with replacement
shuffle() # mosaic
```

# **Graphics**

```
bwplot()
xyplot()
densityplot()
histogram()
plotFun() # mosaic
```

## **Numerical Summaries**

These functions have a formula interface to match plotting.

```
mean() # mosaic augmented
median() # mosaic augmented
sd() # mosaic augmented
var() # mosaic augmented
tally() # mosaic
qdata() # mosaic
pdata() # mosaic
IQR()
```

# Model Building and Inference

```
mm()  # mosaic
lm()  # linear models
glm()  # for logistic models
resid()
fitted()
confint()
anova()
summary()
makeFun() # mosaic
listFun() # devel
```

# Interactive

For classroom use.

```
mLM() mLineFit() mLinAlgebra()
mCI() mHypTest() mPower()
```

### Formulas for Models

```
response ~ a+b # main effects
response ~ a*b # interaction, too
```

Do not use | or groups=.

#### Common forms:

All cases the same:
response ~ 1
Main effects intercept
response ~ X + Y

Exclude intercept (Rarely used. Be careful!)

response  $\sim$  -1 + X + Y

Main effects and interaction:

response ~ X \* Y

Pure interaction (Rarely used.)

response ~ X:Y

Polynomial terms:

response ~ poly(X,2)

Random model vectors (pedagogical) response ~ rand(2)

## **Data and Variables**

```
fetchData() # mosaic
names()
head()
levels()
subset()
with() # operate on data
transform() # new var in data
factor() # categorical vars
merge()
rank()
```

## Formulas for Graphs & Numerics

Plotting (e.g. xyplot, densityplot, bwplot) and simple numerics (e.g. tally, mm) use formulas in the following ways:

y: is y-axis variable. Leave blank for densityplots

x: is x-axis variable

z: conditioning variable (separate panes in graphs)

groups: conditioning variable (overlaid in graphs)

For other things  $y \sim x \mid z$  can usually be interpreted to mean "y depends on x separately for each z."

# **Common Example Datasets**

Can be used directly with data=:

```
Galton # heights
CPS85 # wages
KidsFeet
Marriage
SAT
```

Read in with fetchData():

```
utils = fetchData("utilities.csv")
alder = fetchData("alder.csv")
grades = fetchData("grades.csv")
courses = fetchData("courses.csv")
# Load software in development:
fetchData("m155development.R")
```

### Quick Look at a Data Frame

```
cps = fetchData("CPS85")
names(cps) nrow(cps) summary(cps)
head(cps) sample(cps,size=5)
```

# **Tallying**

A simple count of the number in each level tally( ~ sex, data=CPS85)

A two-way table of counts

tally( ~ sex + married, data=CPS85) Conditional proportions: A | B means "A conditioned on B".

tally( ~ sex | married, data=CPS85)

Different from ~married|sex.

#### New Dataframe Variable

```
g = fetchData("Galton")
```

Add a variable named mid

g = transform(g,
 mid=(father+1.08\*mother)/2)

names(g) #confirm that it's there

### **Subsets**

... Random subset subset (CPS85, size=4)

# **Data from Google Spreadsheets**

In Google, choose File/Publish to the Web. Get link to the published data as CSV, sheet 1. Copy the link

mydat=fetchGoogle("https://docs.google...")

### **Distributions**

Simple distribution

```
densityplot( ~ age, data=CPS85 )
```

Overlaying two (or more) groups

```
densityplot( ~ age | sex, data=CPS85 )
bwplot( age ~ sector, data=CPS85 )
```

### **Scatter Plots**

```
xyplot( wage ~ age, data=CPS85 )
```

groups= and | work as with densityplot().

## **Plotting Model Values**

```
mod = lm(wage ~ educ+sex,data=CPS85 )
xyplot(fitted(mod) ~ educ,data=CPS85 )
xyplot(wage+fitted(mod) ~ educ,data=CPS85)
```

## **Extract Model Information**

```
mod = lm(wage~educ+sex,data=CPS85)
coef(mod)
fitted(mod)
resid(mod)
f = makeFun(mod) # model function
```

### P's and Q's

Want to find the value that separates the lower 30% from the higher 70%:

```
qdata(0.3, wage, data=CPS85 )
```

Have a value and want to find what fraction of the cases are at or below the value:

```
pdata(10, wage, data=CPS85 )
```

#### Randomization

#### Sample

```
mysamp=sample(CPS85,size=100)
```

#### Resample:

lm(wage~educ,data=resample(CPS85))

#### Shuffle (for hypothesis testing)

lm(wage~shuffle(educ),data=CPS85)

#### Probability distributions:

```
rnorm(10,mean=25,sd=2)
rbinom(10,prob=.5,size=40)
rpois(10,lambda=50) # events per period
rexp(10,rate=0.01) # 1/ave time btw events
```

### **Confidence Intervals**

#### ... via "normal theory"

```
mod = lm( wage ~ educ, data=CPS85 )
confint(mod)
```

See also summary (mod).

#### ... via bootstrapping

```
s = do(500)*
lm(wage~educ, data=resample(CPS85))
sd(s) # standard error
```

See also confint(s)

## Quantitative $\rightarrow$ Categorical

## Sums of Squares, Dot Products

```
sum( fitted(mod)^2 )
sum( resid(mod)^2 )
with( data=CPS85, sum(wage^2))
sum(fitted(mod)*resid(mod)) # dot prod
```

## Something is Wrong

```
run = fetchData("repeat-runners.csv")
mean( net, data=run )
```

Some of the data was missing, thus the NA. The FIX:

```
options(na.rm=TRUE)
mean( net, data=run )
```

# Can't Find Something Here?

Send a note to kaplan@macalester.edu