

# Loading Data into R

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# Loading data

- We will use .csv (comma separated values), because most software can both write and read this format
- `somedataset <- read.csv(file.choose())`
- Always check with `str()` that the file has loaded correctly

# Your turn

- Open the Shangri La data in excel, save it as csv, and then load into R.
- Open the baseball data in excel, save it as csv, and then load into R.
- Check that they look OK using `str()`
- Advanced: Open the csv in Word. Try and break the data import, by adding odd characters (try #, , “, ), read `?read.csv` and figure out what's going on.

# Examining variables

- `a`
- `head(a)`
- `summary(a)`
- `str(a)`
- `dim(a)`

# Your turn

- Download data from the United States Cancer Statistics at [http://www.cdc.gov/cancer/npcr/uscs/2004/download\\_data.htm](http://www.cdc.gov/cancer/npcr/uscs/2004/download_data.htm)
- Unzip the archive (use Winzip, e.g.)
- Load ByArea.txt into Excel (2007)

# in Excel

- Replace all ~ by NA. Are there other symbols that should be replaced by NAs?
- Delete all records for “2002-2004”, for “male and female”, and all other fields that represent sums of other fields
- Split rate, upper & lower CI, and count into two columns each according to event type
- Save as comma delimited text file (.csv)

# Switch to R

- # Load data into R  
cancer <- read.csv(file.choose())
- Use  
head(cancer)  
dim(cancer)  
summary(cancer)  
to check that it worked
- Did you catch all the symbols for missing data in Excel?

# Data frames

- A data frame is a list of vectors of the same length
- Create with **data.frame** using named arguments
- `data.frame(a=1:10, b=c(TRUE,FALSE))`
- Created by **read.csv** too

# Extracting subsets

- One of the keys to mastering the R is learning to use the extraction (or subset) operators effectively.

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- By positive integers, select specified
- By negative integers, omit specified
- By logical vector, select T, omit F
- By character vector (by name)

# “Sub”sets can be bigger

- `a <- c(1, 5, 9)`
- `a[c(1,2,3)]`
- `a[c(1,1,1,2,2,3)]`

# [ + logical vectors

- The most complicated to understand, but the most powerful
- Lets you extract a subset defined by some characteristic of the data
- `cancer$Site[cancer$Mortality.Rate > 100]`
- `cancer[cancer$Mortality.Rate > 100,]`

# Updating subsets

- You can take a subset and update the original data
- `a <- 1:4`
- `a[2:3] <- 0`
- `a`
- Very useful with logical subsetting

# Practice

- Select the Race variable in three different ways
- Drop variables Age.Adjusted.Rate, Age.Adjusted.CI.Lower, and Age.Adjusted.CI.Upper from the dataset
- Replace all “~” by NA
- Replace all other symbols for missing data by NA

# More about missings

- $NA + x = NA, NA * x = NA$
- $x == NA$
- **is.na** returns logical vector, for single vector
- **complete.cases** does the same for a data.frame
- Many functions have *na.rm*

# Practice

- Remove all missings from the cancer data.  
Why might this be a problem?
- Remove all records with missing mortality rate.

# Analysing the data

- What questions do we have about the data?
- Write down questions for 1 min, then get together with your neighbor and discuss.
- What data will you need to try to answer your questions? What graphics would you draw in support of your questions?  
Discuss again. Be ready to report.

# Questions about the cancer data

# Report

- Write a short report, which should include:
  - your question
  - your expectation before looking at the data
  - a graphic which answers the question
  - a conclusion based on the graphic
- Print/Email your report (don't forget to put your names on it, too)

# Homework

- Pick one of the “Major Findings” from [http://www.cdc.gov/cancer/npcr/uscs/2004/facts\\_major\\_findings.htm](http://www.cdc.gov/cancer/npcr/uscs/2004/facts_major_findings.htm) and find a graphic which supports this finding
- Write up a paragraph about what else the graphic also shows