Coursera - Data Science & R Course

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Types of Data Science Questions

Descriptive

Goal: To describe or summarize a set of data

- Early analysis when receiving new data
- Generate simple summaries about the samples and their measurements
- **Not** for generalizing the results of the analysis to a larger population

Exploratory

Goal: To examine the data and find relationships that weren't previously known

- Explore how different variables might be related
- Useful for discovering new connections
- Help to formulate hypotheses and drive the design of future studies and data collection

Inferential

Goal: Use a relatively small sample of data to say something about the population at large

- Provide your estimate of the variable for the population and provide your uncertainty about your estimate
- Ability to accurately infer information about the larger population depends heavily on sampling scheme

Predictive

Goal: Use current and historical data to make predictions about future data

- Accuracy in predictions is dependent on measuring the right variables
- Many ways to build up prediction models with some being better or worse for specific cases

Causal

Goal: See what happens to one variable when we manipulate another variable

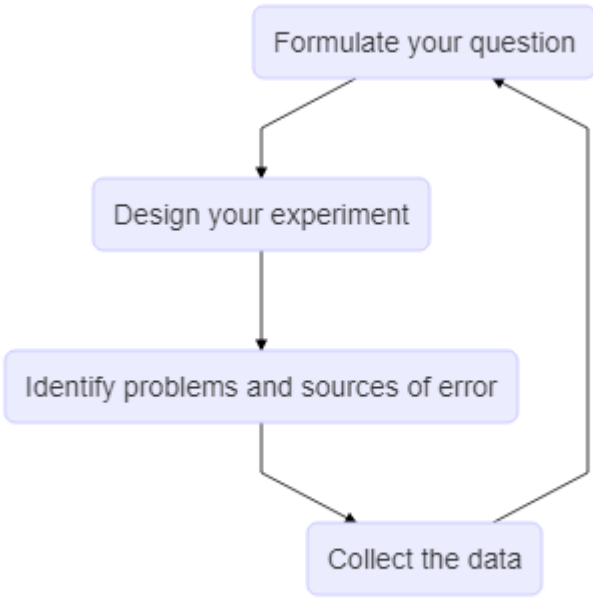
- Gold standard in data analysis
- Often applied to the results on randomized studies that were designed to identify causation
- Usually analyzed in aggregate and observed relationships are usually average effects

Mechanistic

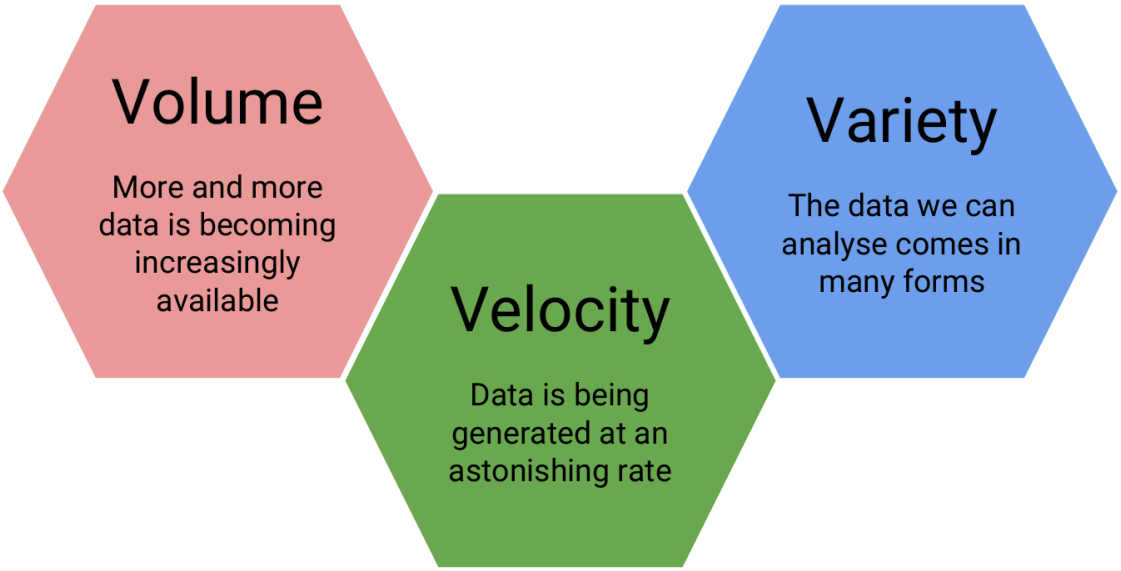
Goal: Understand the exact changes in variables that lead to exact changes in other variables

- Applied to simple situations or those that are nicely modeled by deterministic equations
- Commonly applied to physical or engineering sciences
- Often, the only noise in the data is measurement error

Experimental design



Big Data



Practical R Exercises in swirl

- install.packages("swirl")
- library(swirl)
- install_from_swirl("R Programming")
- swirl()
- getwd()
- setwd()
- ls()
- dir()
- dir.create()
- file.create()
- file.exists()
- file.infor()
- file.path()
- file.rename()

- file.copy()

–

- length(x)
- x <- dim(x,y) » Creates matrix
- attributes(x)

–

- matrix(data, nrow,ncol)
- cbind({vector with row names}, matrix) » Combine columns
- colnames(data.frame) = {vector with column names} » Names of columns
- data.frame({vector with row names}, matrix) » Allows Text and Numbers

Types of Data Science Questions

Control Structures - if/else

if(condition) {

```
  <do something>
```

} else(“condition2”) {

```
  <do something else>
```

}

–

if(condition) {

```
  <do something>
```

{ else if(condition2) {

```
  <do something different>
```

} else {

```
  <do something else>
```

}

Control Structures - for loop

for(i in vect){

```
  <function>[i]
```

}

Control Structures - while

while(finite condition){

```
  <do something>
```

}

Control Structures - repeat, next, break, return

`repeat` is a construct that basically initiates an infinite loop. The only way to exit a `repeat` loop is to call `break`

`next` is basically used in any time of looping construct when you want to skip an iteration.

`return` signals that a function should exit and and return a given value

Functions

Basic format:

f <- function(x, y =

){

x+y

}

User-defined binary operators have the following syntax:

```
  %[whatever]%
```

where [whatever] represents any valid variable name.

Dates and Time

Dates are represented by the `Date` class and times are represented by the `POSIXct` and `POSIXlt` classes. Internally, dates are stored as the number of days since 1970-01-01 and times are stored as either the number of seconds since 1970-01-01 (for `POSIXct`) or a list of seconds, minutes, hours, etc. (for `POSIXlt`).

`strptime()` converts character vectors to `POSIXlt`.

If you want more control over the units when finding the above difference in times, you can use `difftime()`, which allows you to specify a 'units' parameter.

Loop functions

`lapply`: Loop over a list and evaluate a function on each element *e.g.* `lapply(data, fun)`

`sapply`: Same as `lapply` but try to simplify the result

`apply`: Apply the function over the margins of an array *e.g.* `apply(data, dimension, fun)`

`tapply`: Apply a function over the subsets of a vector

`mapply`: Multivariate function of `lapply`