## Today

- Recap & questions from homework
- Functions
- Coding a descent algorithm (training)
  - using optim()

### Git skills

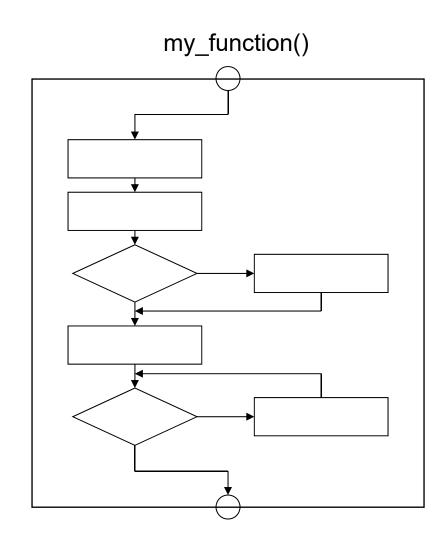
- git amend
- .gitignore
- gitgui
- gitk

### Fit ecological model

- Natural process culture of data science
- Paramecium logistic growth
- Parameters: r, K, N(0)
- Grid search

### Programming: functions

- A function encapsulates an algorithm
- Functions break a program down into modules
- Modularized programs are easier to write, debug, maintain, and modify
- Functions make algorithms easier to reuse



### Making a function in R

?"function" – only the bare bones

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?"function" – only the bare bones

```
function_name <- function(arguments) {
    expression
    return(object)
}
diff_two_nums <- function(x, z) {
    y <- x - z
    return(y)
}</pre>
```

### Making a function in R

?"function" – only the bare bones

```
function name <- function(arguments) {</pre>
     expression
     return (object)
diff two nums <- function(x, z) {
     y <- x - z
     return(y)
                            Objects listed in the arguments or defined
                            in the function can only be seen inside the
                            function. These are called local variables.
                            Concept: scope.
```

### Scope

- See examples in functions.R
- Good programming practice: avoid global variables
  - Define local variables by including in argument list or initializing within the function
  - Global variables make programs harder to maintain and debug

### Make a function

```
function_name <- (arguments) {
    expression
    return(object)
}</pre>
```

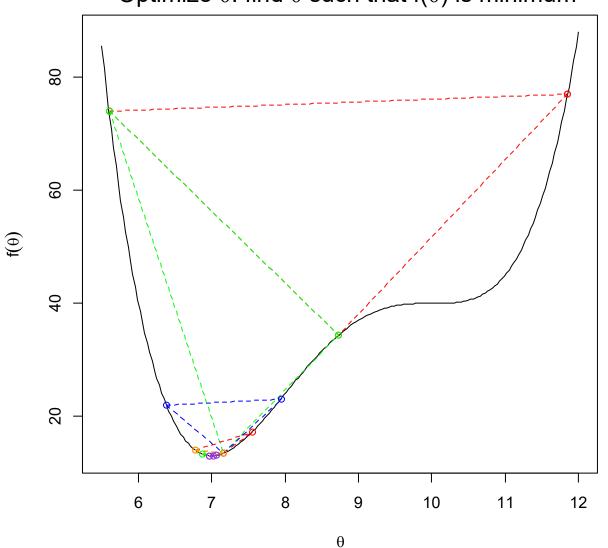
#### Exercise:

Make a function to calculate the linear model given the model parameters and a vector of x data. In other words, turn the following into a function:

$$y < - b_0 + b_1 * x$$

## Descent algorithms

Optimize  $\theta$ : find  $\theta$  such that  $f(\theta)$  is minimum

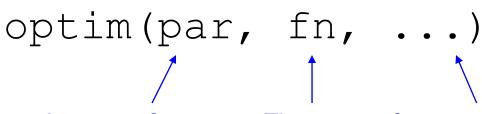


### Narrowing in:

keep changing parameters in the direction that leads to lower SSQ

# optim()

- Has various descent and Monte Carlo methods
- Nelder-Mead algorithm is default (method="Nelder-Mead")



A vector of parameters to be estimated

The name of the function to be minimized

Other arguments to pass to the function. Usually the data plus any parameters not to be estimated.

# Training models: general recipe

- 1) biology function
  - complex mechanistic to abstract pattern
- 2) error function
  - e.g. SSQ: distance of the model from the data sum((observed - predicted) ^ 2)
- 3) optimize
  - find biology parameters that minimize the error
- This recipe is the same no matter how complicated the process model or error function

## Code (train\_ssq\_optim.R)

```
Biology function (linear)
linmod <- function(b 0, b 1, x) {</pre>
    y < -b 0 + b 1 * x
    return(y)
            Parameters are first argument
                                           Response data
                                               Auxiliary data
Error function (SSQ)
ssq linmod <- function(p, y, x) {
                                                         Use the biology
    y pred <- linmod(b_0=p[1], b_1=p[2], x)</pre>
                                                            function to get
    e <- y - y_pred
                                                            predicted values
    ssq \leftarrow sum(e^2)
    return (ssq)
                      Compare predicted
                                            "Unpack" the parameters
                     to the data
                                            (self documenting)
Use optim to optimize error function
par <- c(b 0 start, b 1 start) Starting values for parameters
fit <- optim(par, ssq_linmod, y=data$y, x=data$x)</pre>
                        Need ""= sign
```