# Assignment 6: Using Sobol with an ODE

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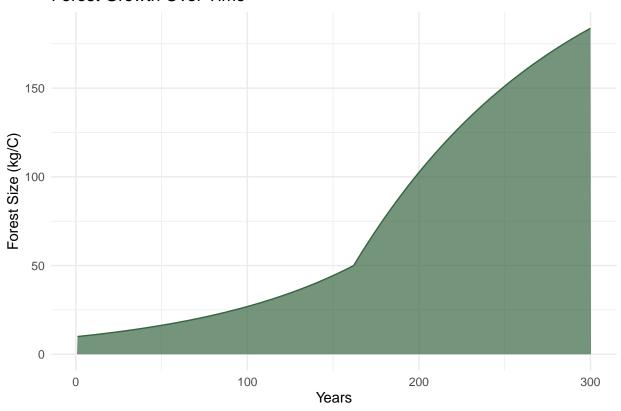
### Establishing the Carbon Model

```
source("calculate_forest_growth.R")
calculate_forest_growth
## function (time, C, params)
##
       if (C < params$thresh) {</pre>
           carbon_change = params$r * C
##
##
       }
##
       else {
##
           carbon_change = params$g * (1 - C/params$k)
##
##
       return(list(carbon_change))
```

### Running the Model without Sobel

# Graphing the Results

## Forest Growth Over Time



### Running Model with Sobol Sensitivity Analysis

### Initiating parameters

```
# setting sobel inputs
num_params <- 1500
sd_percent <- 0.10

### ------
### parameter values</pre>
```

```
k <- rnorm(mean = params$k,
           sd = (params$k * sd_percent),
           n = num_params)
r <- rnorm(mean = params$r,
           sd = (params$r * sd_percent),
           n = num_params)
g <- rnorm(mean = params$g,
           sd = (params$g * sd_percent),
           n = num_params)
thresh <- rnorm(mean = params$thresh,</pre>
           sd = (params$thresh * sd_percent),
           n = num_params)
sample1 = cbind.data.frame(k = k,
                       r = r,
                       g = g,
                       thresh = thresh)
k <- rnorm(mean = params$k,
           sd = (params$k * sd_percent),
           n = num_params)
r <- rnorm(mean = params$r,
           sd = (params$r * sd_percent),
           n = num_params)
g <- rnorm(mean = params$g,
           sd = (params$g * sd_percent),
           n = num_params)
thresh <- rnorm(mean = params$thresh,</pre>
           sd = (params$thresh * sd_percent),
           n = num_params)
```

```
# changing any negative values to zero in both samples
sample1 <- sample1 |>
    map_df(pmax, 0.0)

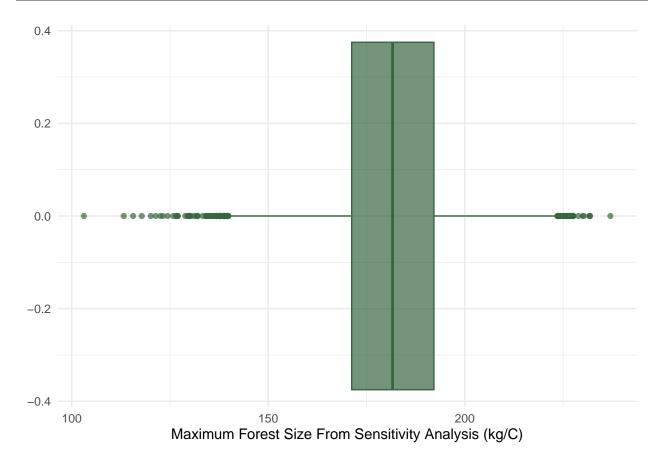
sample2 <- sample2 |>
    map_df(pmax, 0.0)
```

#### Creating sobol parameter object

#### Conducting sensitivity analysis

# Sensitivity Analysis - Indices & Plot

#### Graphing the sensivitity results



#### Reporting sobel indices

#### max\_growth\$S

```
## original bias std. error min. c.i. max. c.i.
## k 0.35858225 -0.0036924134 0.02651246 0.309576342 0.4163448
## r 0.34102945 -0.0017966077 0.02462932 0.291323904 0.3886687
## g 0.19734455 -0.0023145924 0.02468385 0.150804583 0.2503948
## thresh 0.05872827 -0.0009866682 0.02635791 0.005379591 0.1060137
```

```
# adding row names to second order
rownames(max_growth$T) <- c("k", "r", "g", "thresh")
print("Second Order Indices")</pre>
```

## [1] "Second Order Indices"

#### max\_growth\$T

```
## original bias std. error min. c.i. max. c.i.
## k 0.37032319 -1.096166e-03 0.01857377 0.33168616 0.40524311
## r 0.38535542 1.256730e-03 0.02008718 0.34226370 0.42336110
## g 0.20945947 1.350349e-04 0.01078753 0.18738093 0.22846209
## thresh 0.06356204 -2.312842e-05 0.00373331 0.05586691 0.07096621
```

#### Discussion

Looking at the first order index, it appears that both carrying capacity and the exponential growth rate is most sensitive to the output of the model with a relatively small confidence interval that does not include 0. This could mean that the output from rate at which forest grow when below the threshold canopy closure is most sensitive to change and in the context of climate change, with more severe storms and heavy rain instances contrasted with extreme drought, this model will could not accurately portray forest growth and there could be a decrease in growth rate. Also, depending on the conditions, warming temperature could lead to more extreme growing rates when less than the threshold canopy. This rapid growth could have negative implications for understory flora and biodiversity.