

Module10

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Exercise 1 - Create a duniform function for the density function.

```
#a. Simple function without vector input.
duniform = function(x,a=0,b=1){
  #Check the bounds of x per the density function.
  if(a <= x && x <= b){
    return(1/(b-a))
  }
  #If the check fails, return 0 instead.
  return(0)
}
```

```
#x<a
duniform(1,2,3)
```

```
## [1] 0
```

```
#a<x<b
duniform(3,2,4)
```

```
## [1] 0.5
```

```
#b<x
duniform(6,7,3)
```

```
## [1] 0
```

```
#b. Duniform with vector input modification
duniformVec = function(x,a=0,b=1){
  output = NULL

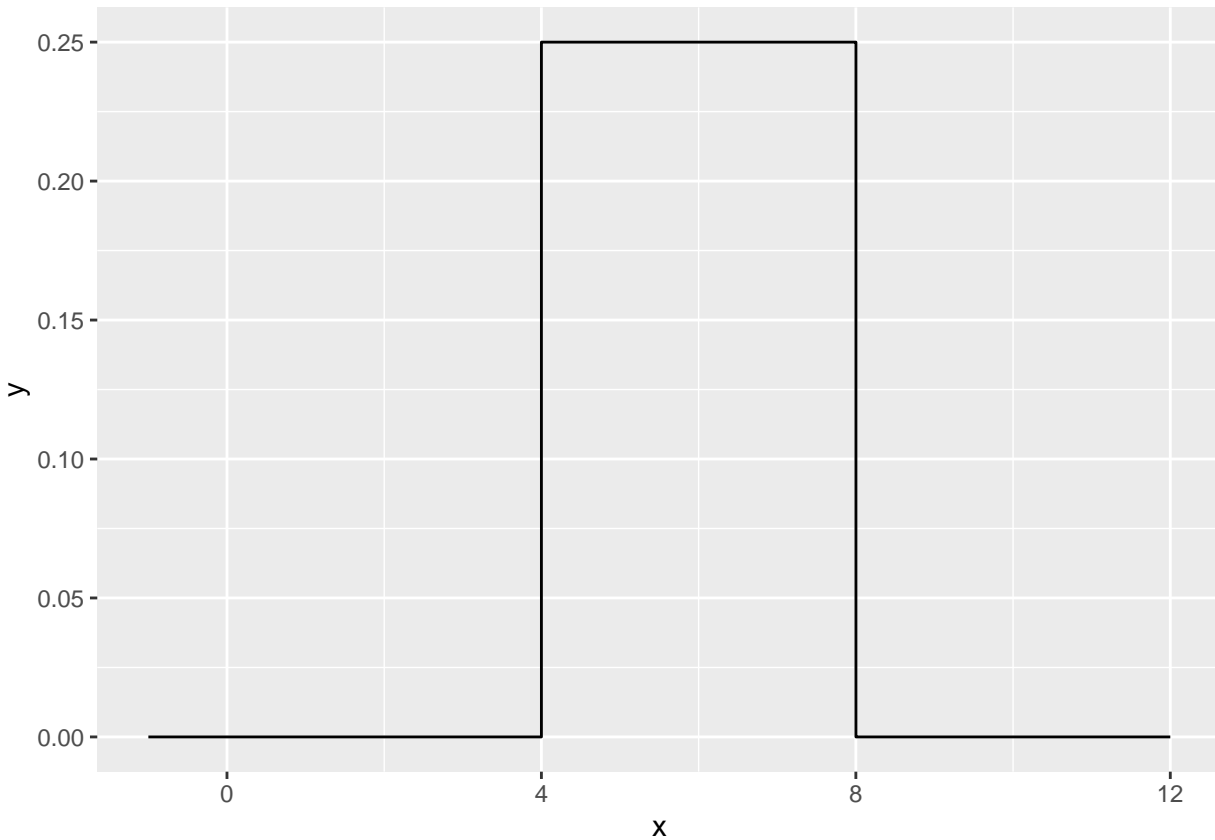
  #Grab indicies that correspond to the length of x.
  for(i in 1:length(x)){
    #Check the bounds of x per the density function.
    if(a <= x[i] && x[i] <= b){
      output[i] = (1/(b-a))
    }
    else{
      output[i] = 0
    }
  }

  #Once we finish looping and adding to output, return the vector
  return(output)
}
```

```

}
#Test code
data.frame( x=seq(-1, 12, by=.001) ) %>%
  mutate( y = duniformVec(x, 4, 8) ) %>%
  ggplot( aes(x=x, y=y) ) +
  geom_step()

```



```

#c. Microbenchmark the duniformVec function.
microbenchmark::microbenchmark( duniformVec( seq(-4,12,by=.0001), 4, 8), times=100)

```

```

## Unit: milliseconds
##               expr      min       lq      mean  median
##  duniformVec(seq(-4, 12, by = 1e-04), 4, 8) 39.8738 42.69665 45.53388 44.0648
##           uq      max neval
## 46.11045 101.9896   100

```

```

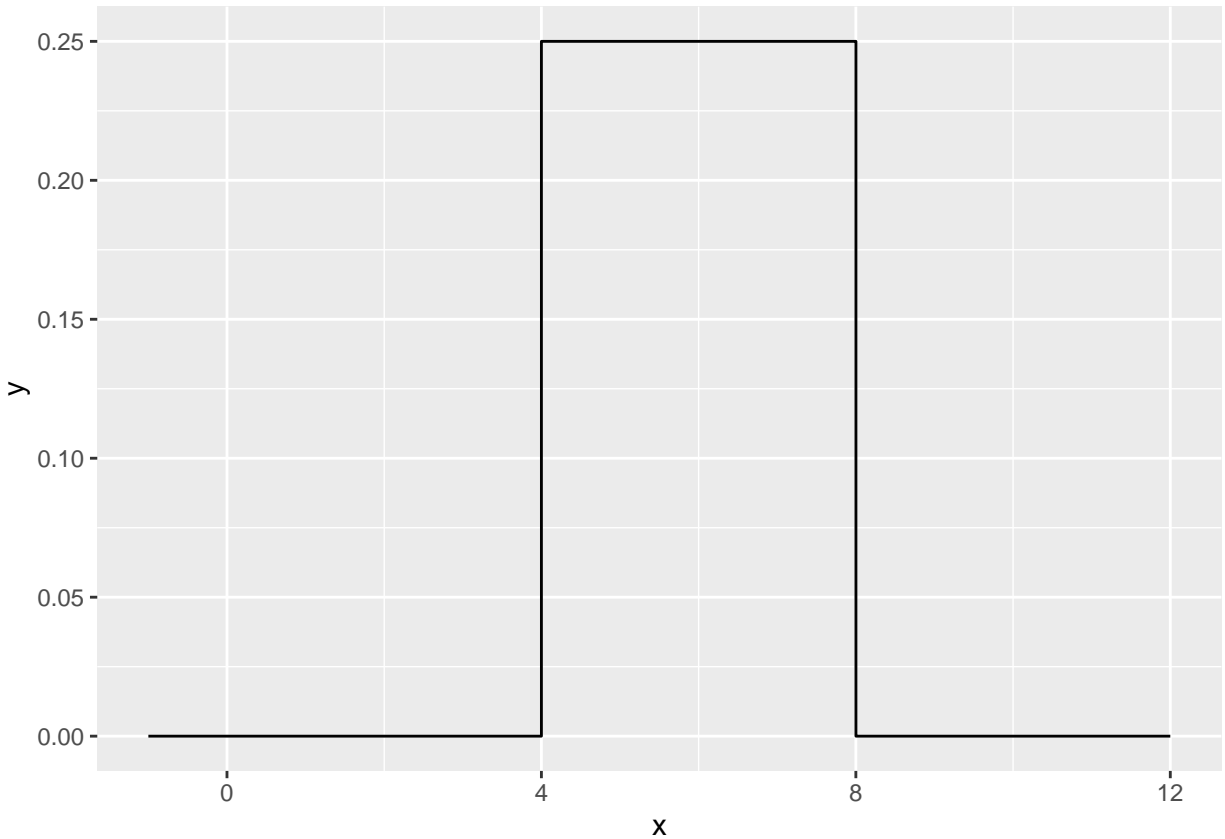
#d. Optimized duniform function using the ifelse command instead of a for loop.
duniformOpt = function(x,a=0,b=1){
  return(ifelse(a<=x & x <= b, 1/(b-a), 0))
}

```

```

#Test code
data.frame( x=seq(-1, 12, by=.001) ) %>%
  mutate( y = duniformOpt(x, 4, 8) ) %>%
  ggplot( aes(x=x, y=y) ) +
  geom_step()

```



```
#Microbenchmark
microbenchmark::microbenchmark( duniformOpt( seq(-4,12,by=.0001), 4, 8), times=100)
```

```
## Unit: milliseconds
##              expr      min       lq      mean  median
## duniformOpt(seq(-4, 12, by = 1e-04), 4, 8) 2.9553 3.1346 4.800938 3.35935
##              uq      max neval
## 5.32975 59.2713   100
```

```
#d. The function with the ifelse command ran much faster than the for loop function. To me, I like the .
# as that is what I am used to. However, ifelse takes much of the work out of writing a for loop explic
# to write and analyze.
```

Exercise 2 - Test default values of duniform

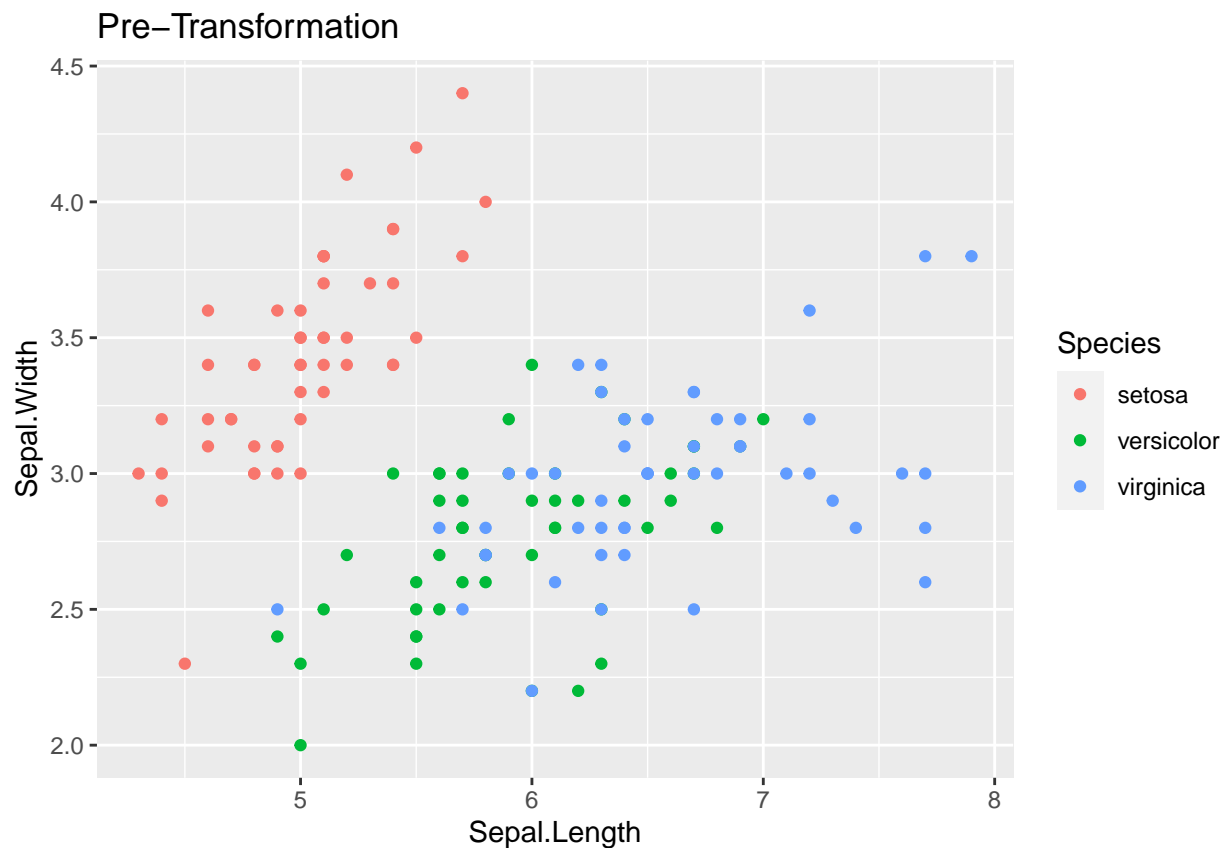
```
duniform(0)
```

```
## [1] 1
```

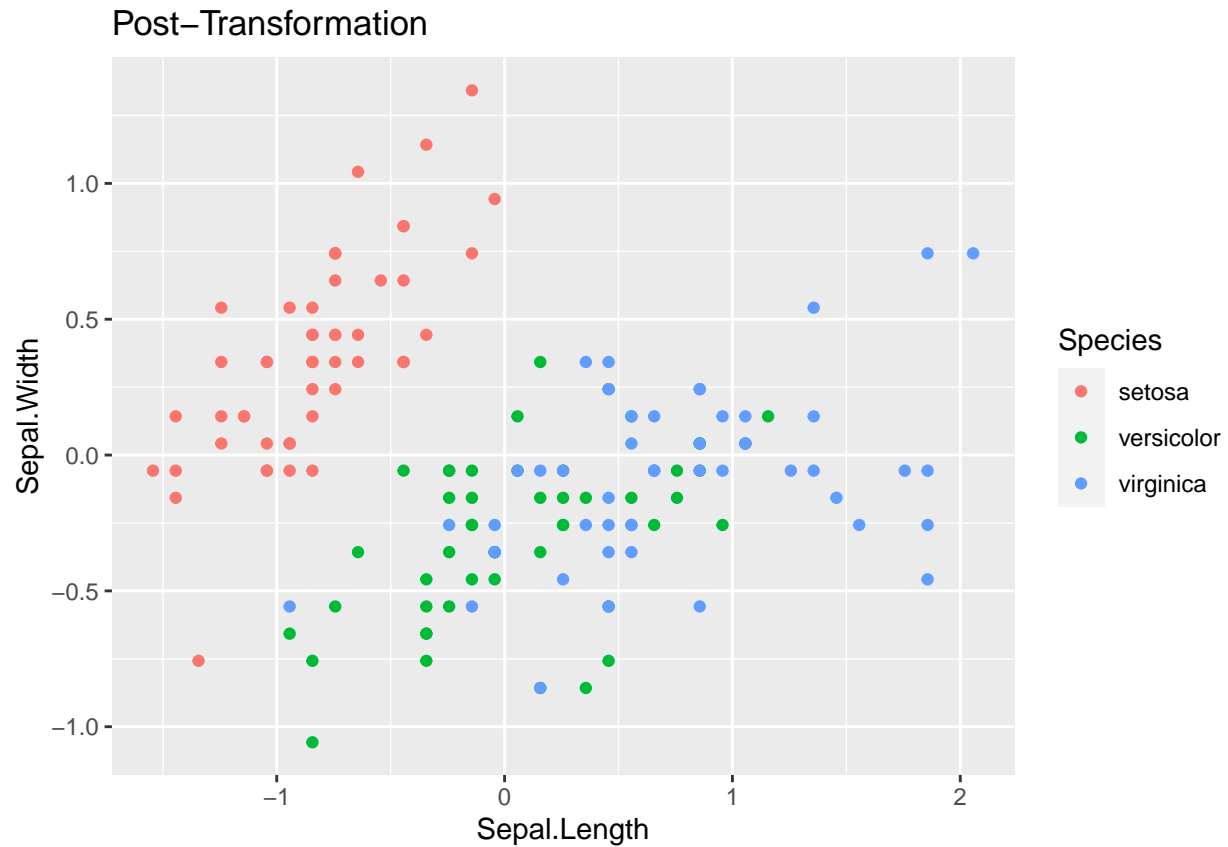
```
#This is working properly as I set default values in duniform in exercise 1.
# 1/(b-a) = 1/(1-0) = 0. As x = 0 and a <= x <= b = 0 <= 0 <= 1.
```

Exercise 3 - Function to produce a standardized vector of numeric variables.

```
standardize <- function(x){  
  return(x-mean(x))/(sd(x))  
}  
  
data( 'iris' )  
# Graph the pre-transformed data.  
ggplot(iris, aes(x=Sepal.Length, y=Sepal.Width, color=Species)) +  
  geom_point() +  
  labs(title='Pre-Transformation')
```



```
# Standardize the numeric columns  
iris.z <- iris %>% mutate_if( is.numeric, standardize )  
  
# Graph the post-transformed data.  
ggplot(iris.z, aes(x=Sepal.Length, y=Sepal.Width, color=Species)) +  
  geom_point() +  
  labs(title='Post-Transformation')
```



Exercise 4 - FizzBuzz function in R

```
fizzbuzz = function(n){  
  output = NULL  
  for(i in 1:n){  
    if((i %% 3 == 0) && (i %% 5 == 0)){  
      output[i] = "FizzBuzz"  
    }  
    else if(i %% 3 == 0){  
      output[i] = "Fizz"  
    }  
  
    else if(i %% 5 == 0){  
      output[i] = "Buzz"  
    }  
  
    else{  
      output[i] = i  
    }  
  }  
  return(output)  
}  
  
fizzbuzz(20)
```

```
## [1] "1"      "2"      "Fizz"    "4"      "Buzz"    "Fizz"
## [7] "7"      "8"      "Fizz"    "Buzz"    "11"      "Fizz"
## [13] "13"     "14"     "FizzBuzz" "16"     "17"      "Fizz"
## [19] "19"     "Buzz"
```

Exercise 5 - Debugging a bootstrap CI function

```
## Calculate bootstrap CI for an lm object
##
## @param model
## @param N
boot.lm <- function(model, N=1000){
  #browser()
  data <- model$model # Extract the original data
  formula <- model$terms # and model formula used
  # Start the output data frame with the full sample statistic
  output <- broom::tidy(model) %>%
    select(term, estimate) %>%
    spread(term, estimate)

  for( i in 1:N ){
    data <- data %>% sample_frac( replace=TRUE )
    model.boot <- lm( formula, data=data)
    coefs <- broom::tidy(model.boot) %>%
      select(term, estimate) %>%
      spread(term, estimate)
  }
  output <- output %>% rbind( coefs )
  return(output)
}

# Run the function on a model
m <- lm( Volume ~ Girth, data=trees )
boot.dist <- boot.lm(m)

# If boot.lm() works, then the following produces a nice graph
boot.dist %>% gather('term', 'estimate') %>%
  ggplot( aes(x=estimate) ) +
  geom_histogram() +
  facet_grid(.~term, scales='free')
```

