

STA 445 Assignment 2

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Exercise 1

Write a function that calculates the density function of a Uniform continuous variable on the interval (a, b) . The function is defined as

$$f(x) = \begin{cases} \frac{1}{b-a} & \text{if } a \leq x \leq b \\ 0 & \text{otherwise} \end{cases}$$

We want to write a function `duniform(x, a, b)` that takes an arbitrary value of `x` and parameters `a` and `b` and return the appropriate height of the density function. For various values of `x`, `a`, and `b`, demonstrate that your function returns the correct density value. a) Write your function without regard for it working with vectors of data. Demonstrate that it works by calling the function with a three times, once where $x < a$, once where $a < x < b$, and finally once where $b < x$. b) Next we force our function to work correctly for a vector of `x` values. Modify your function in part (a) so that the core logic is inside a `for` statement and the loop moves through each element of `x` in succession. c) Install the R package `microbenchmark`. We will use this to discover the average duration your function takes. d) Instead of using a `for` loop, it might have been easier to use an `ifelse()` command. Rewrite your function to avoid the `for` loop and just use an `ifelse()` command. Verify that your function works correctly by producing a plot, and also run the `microbenchmark()`. Which version of your function was easier to write? Which ran faster?

#Exercise 1: Part a

```
duniform <- function(x,a,b){  
  if (a < x && x < b) {return(1/(b-a))}  
  else {return(0)}  
}  
duniform(0, 1, 2)
```

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

```
duniform(1, 0, 2)
```

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

```
duniform(3, 1, 2)
```

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

#Exercise 1: Part b

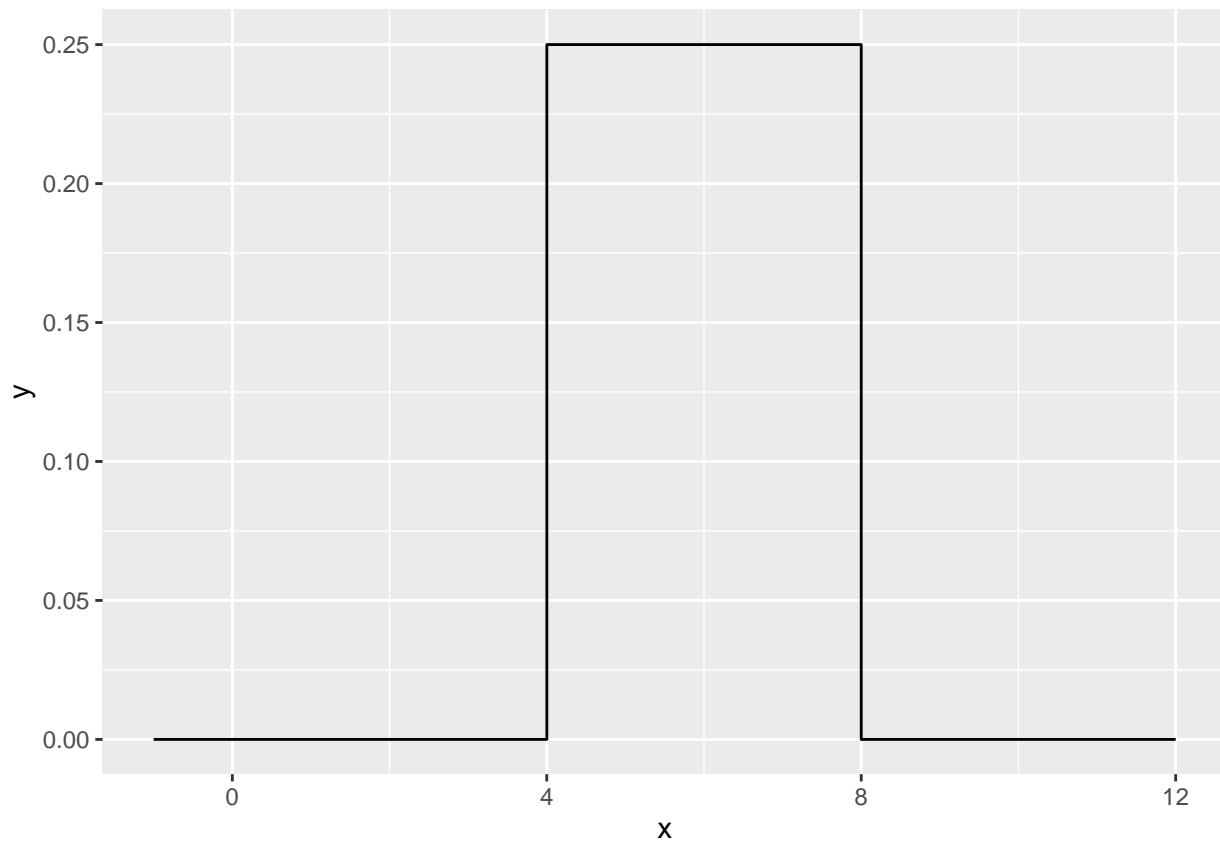
```
duniform <- function(x, a, b){  
  output <- NULL  
  for( i in 1:length(x) ){  
    if( x[i] > a && x[i] < b ){  
      output[i] = (1/(b-a))  
    }else{  
      output[i] = 0  
    }  
  }  
}
```

```

    }
  }
  return(output)
}

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

```



```

#Exercise 1: Part C
#Installing the R package
g <- microbenchmark::microbenchmark( duniform( seq(-4,12,by=.0001), 4, 8), times=100)
g

```

```

## Unit: milliseconds
##           expr      min       lq      mean     median
##  duniform(seq(-4, 12, by = 1e-04), 4, 8) 40.0166 44.6277 46.86997 46.31375
##           uq      max neval
## 47.1124 102.6368   100

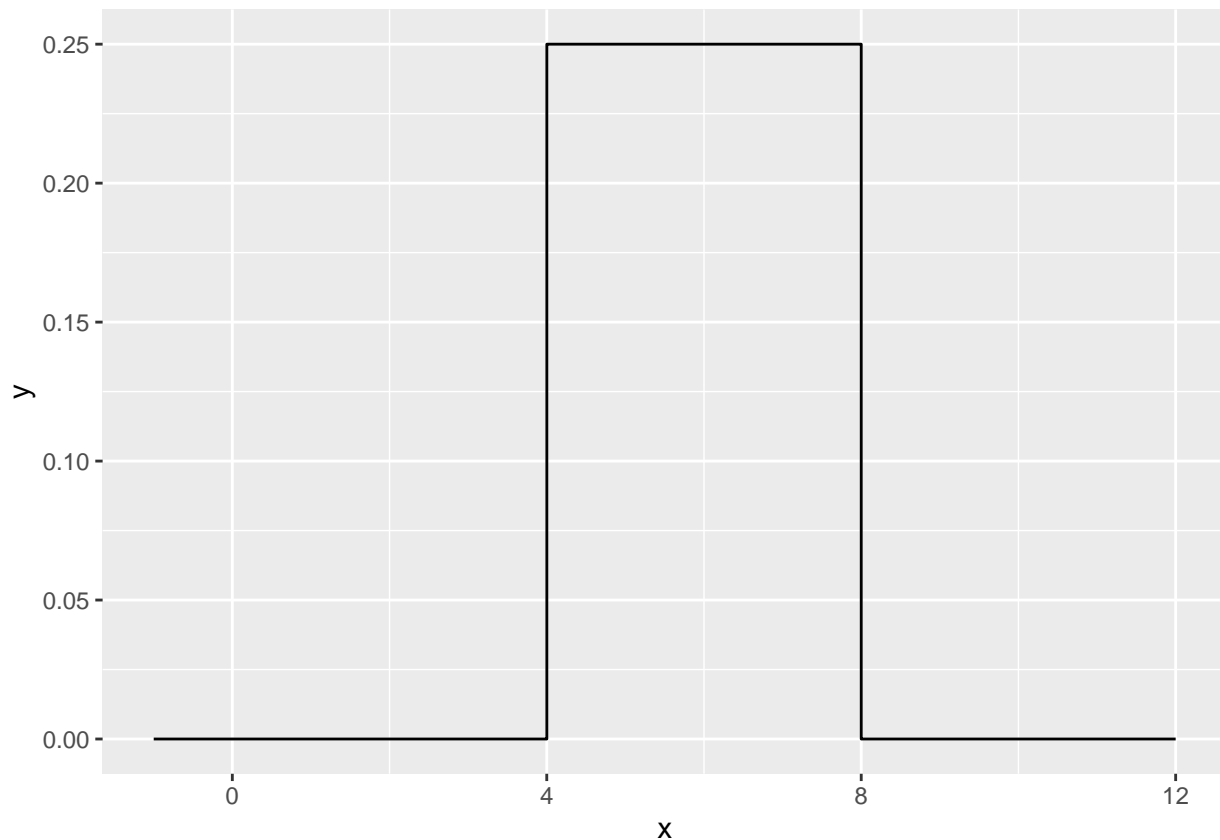
```

```

#Exercise 1: Part D
duniform <- function(x, a, b){
  density <- ifelse( x > a & x < b,
    (1/(b-a)), 0)
  return(density)
}

```

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



```
h <- microbenchmark::microbenchmark( duniform( seq(-4,12,by=.0001), 4, 8), times=100)
h
```

Unit: milliseconds	expr	min	lq	mean	median	uq
##	duniform(seq(-4, 12, by = 1e-04), 4, 8)	3.9607	4.36205	6.285891	5.8741	7.22265
##	max neval					
##	16.8904 100					

Exercise 2

I very often want to provide default values to a parameter that I pass to a function. For example, it is so common for me to use the `pnorm()` and `qnorm()` functions on the standard normal, that R will automatically use `mean=0` and `sd=1` parameters unless you tell R otherwise. To get that behavior, we just set the default parameter values in the definition. When the function is called, the user specified value is used, but if none is specified, the defaults are used. Look at the help page for the functions `dunif()`, and notice that there are a number of default parameters. For your `duniform()` function provide default values of 0 and 1 for `a` and `b`. Demonstrate that your function is appropriately using the given default values.

```
#Exercise 2
#Setting a=0 and b=1 so that they are the default values if there is no input for a and b
duniform <- function(x, a=0, b=1){
```

```

    density <- ifelse( x > a & x < b,
      (1/(b-a)), 0)
    return(density)
  }
duniform(.5)

## [1] 1

duniform(2)

## [1] 0

```

Exercise 3

A common data processing step is to *standardize* numeric variables by subtracting the mean and dividing by the standard deviation. Mathematically, the standardized value is defined as

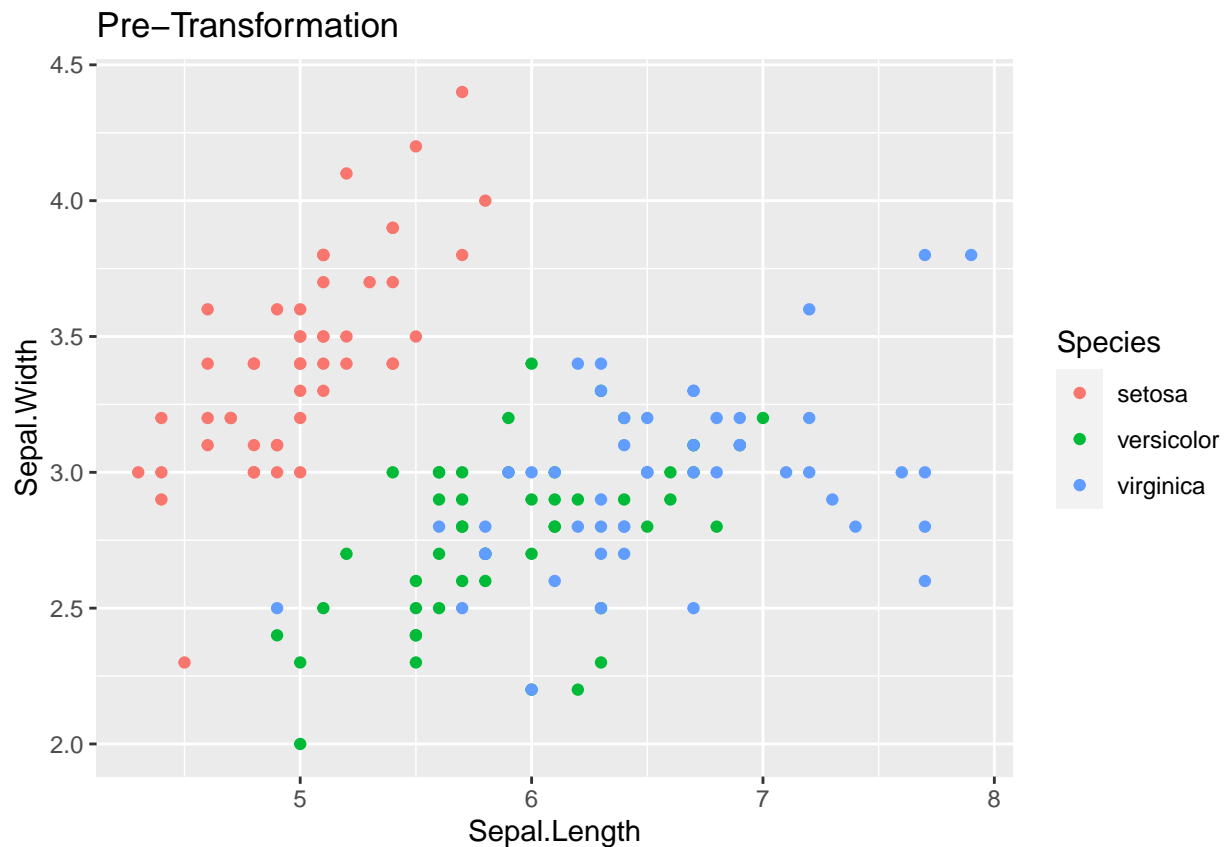
$$z = \frac{x - \bar{x}}{s}$$

where \bar{x} is the mean and s is the standard deviation. Create a function that takes an input vector of numerical values and produces an output vector of the standardized values. We will then apply this function to each numeric column in a data frame using the `dplyr::across()` or the `dplyr::mutate_if()` commands. *This is often done in model algorithms that rely on numerical optimization methods to find a solution. By keeping the scales of different predictor covariates the same, the numerical optimization routines generally work better.*

```

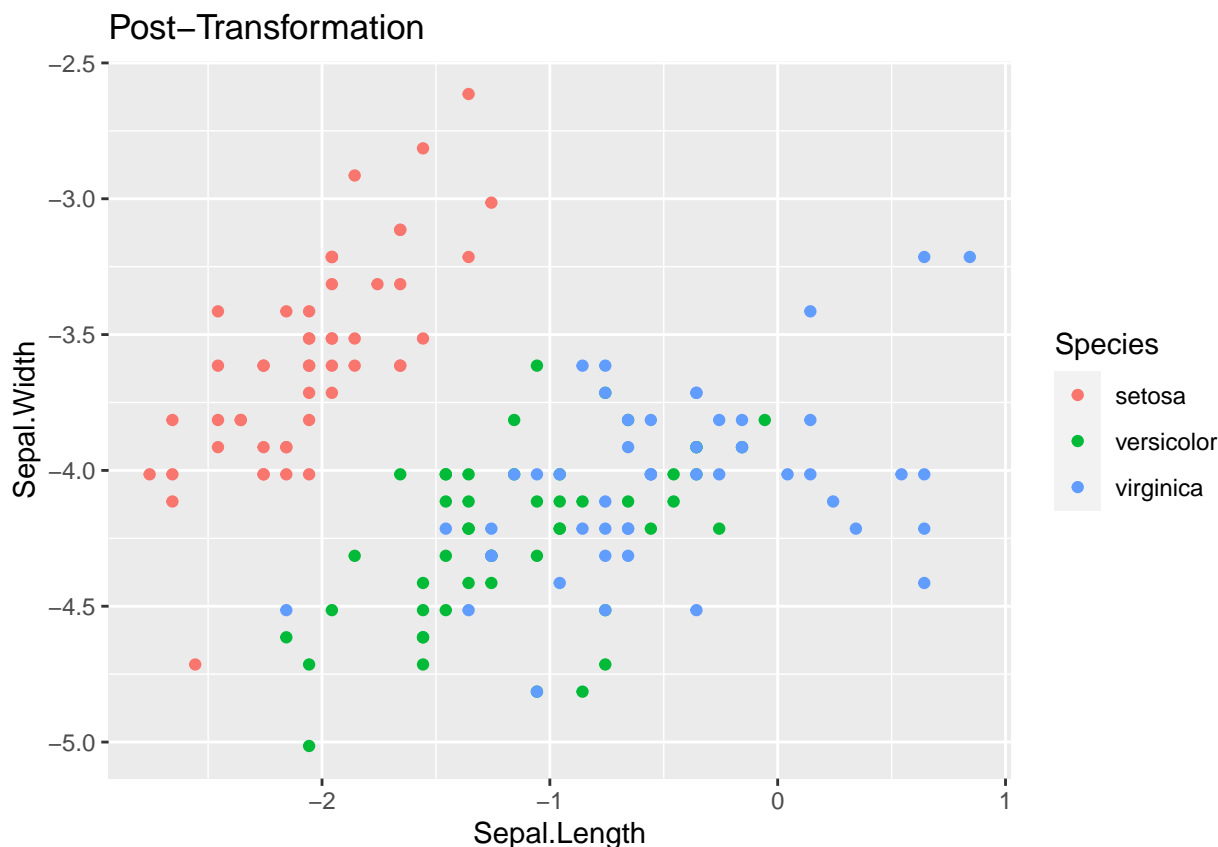
#Exercise 3
#I copied code from the book and created a new function.
standardize <- function(x){
  (x-mean(x)/sd(x))
}
data( 'iris' )
# Graph the pre-transformed data.
ggplot(iris, aes(x=Sepal.Length, y=Sepal.Width, color=Species)) +
  geom_point() +
  labs(title='Pre-Transformation')

```



```
# Standardize all of the numeric columns
# across() selects columns and applies a function to them
# there column select requires a dplyr column select command such
# as starts_with(), contains(), or where(). The where() command
# allows us to use some logical function on the column to decide
# if the function should be applied or not.
iris.z <- iris %>% mutate( across(where(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

In this example, we'll write a function that will output a vector of the first n terms in the child's game *Fizz Buzz*. The goal is to count as high as you can, but for any number evenly divisible by 3, substitute "Fizz" and any number evenly divisible by 5, substitute "Buzz", and if it is divisible by both, substitute "Fizz Buzz". So the sequence will look like 1, 2, Fizz, 4, Buzz, Fizz, 7, 8, Fizz, ... *Hint: The `paste()` function will squish strings together, the remainder operator is `%%` where it is used as `9 %% 3 = 0`. This problem was inspired by a wonderful YouTube video that describes how to write an appropriate loop to do this in JavaScript, but it should be easy enough to interpret what to do in R. I encourage you to try to write your function first before watching the video.*

```
fizzybuzz <- function(x){
  output <- NULL
  for (i in 1:length(x)) {
    if (x[i] %% 3 == 0 & x[i] %% 5 != 0){
      output[i]='Fizz'}
    if (x[i] %% 5 == 0 & x[i] %% 3 != 0){
      output[i]='Buzz'}
    if(x[i] %% 3 ==0 & x[i] %% 5 ==0){
      output[i]='FizzBuzz'}
    if (x[i] %% 3 !=0 & x[i] %% 5 !=0) {output[i]= x[i]}
  }
  return(output)
}

fizzybuzz(1:30)
```

```
## [1] "1"      "2"      "Fizz"   "4"      "Buzz"   "Fizz"
## [7] "7"      "8"      "Fizz"   "Buzz"   "11"     "Fizz"
## [13] "13"     "14"     "FizzBuzz" "16"     "17"     "Fizz"
## [19] "19"     "Buzz"   "Fizz"   "22"     "23"     "Fizz"
## [25] "Buzz"   "26"     "Fizz"   "28"     "29"     "FizzBuzz"
```

Exercise 5

The `dplyr::fill()` function takes a table column that has missing values and fills them with the most recent non-missing value. For this problem, we will create our own function to do the same.

```
## Fill in missing values in a vector with the previous value.
##
## @param x An input vector with missing values
## @result The input vector with NA values filled in.
##
##
myFill <- function(x){
  for (i in 1:length(x)) {
    if (is.na(x[i])){
      x[i] = x[i-1]}
    #else{x[i]=x[i]}
  }
  return(x)
}
test.vector <- c('A',NA,NA, 'B','C', NA,NA,NA)
myFill(test.vector)
```

```
## [1] "A" "A" "A" "B" "C" "C" "C" "C"
```