STAT243 Problem Set 5

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options(digits=22)

- 1. (a) The accuracy we should have is 16 places.
- (b) Indeed, the accuracy of sum is 16 places.

```
#Create the vector
x <- c(1, rep(1e-16, 10000))
sum(x)
## [1] 1.0000000000000999644811
```

(c) When using the sum in Python, it loses the accuracy after the decimal point.

```
import numpy as np
import decimal as dm

# Create Vector
vec = np.array([1e-16]*(10001))
vec[0] = 1

v0 = dm.Decimal(sum(vec))
print(v0)
## 1
```

(d) For the loop in a reversed way, it gives us the correct accuracy while for the original order, it lose the accuracy.

```
#1 is the first term
value <- 0

for(i in 1:length(x)){
  value <- value + x[i]
}

#Ouput
value
## [1] 1</pre>
```

```
#1 is the last term
value <- 0

for(i in length(x):1){
  value <- value + x[i]
}

#Ouput
value
## [1] 1.00000000001000088901</pre>
```

The same thing happens in Python.

```
import numpy as np
import decimal as dm

# Create Vector
vec = np.array([1e-16]*(10001))
vec[0] = 1

# 1 is the first term
value = 0
index = np.arange(10000)

for i in index:
value = value + vec[i]

v1 = dm.Decimal(value)
print(v1)
## 1
```

```
import numpy as np
import decimal as dm

# Create Vector
vec = np.array([1e-16]*(10001))
vec[0] = 1

# 1 is the last term
value = 0
index = np.arange(10000)

for i in reversed(index):
value = value + vec[i]

v2 = dm.Decimal(value)
print(v2)

## 1.00000000000000998668559774159803055226802825927734375
```

(e) The result of **sum** indicate that it address the problem smartly when there are some extremely small or large number. From previous result, it shows that the **sum** definitely does not use

the **for** loop easily from the beginning to end. From the help page, it does not show too many details on it. We can use the **show_c_source** function in package **pryr** to look the source code of sum(**show_c_source**(.Primitive(sum(x)))).

2. I did three operation, for-loop sum, inverse matrix, subset, to compare the speed between integer and double.

First, I created a all 1 vector, \mathbf{x} , and use for loop to get the sum of them. The speed for integer is quite similar to double.

```
#Change the output digits back to 7
options(digits=7)

#for loop
x <- rep(1, 10000000)

#Change data type
intx <- as.integer(x)
#Examine type
typeof(intx)

## [1] "integer"
ins <- 0</pre>
```

```
#Examine type
typeof(x)
## [1] "double"
s <- 0
library(rbenchmark)
#Comparison
benchmark(
  integer=for(i in 1:length(intx)){
    ins <- ins + intx[i]</pre>
  double=for(i in 1:length(x)){
    s \leftarrow s + x[i]
  },
  replications=5
)
##
        test replications elapsed relative user.self sys.self user.child
## 2 double
                         5 31.231
                                       1.000
                                                 30.776
                                                            0.344
                                                                            0
## 1 integer
                         5 33.560
                                       1.075
                                                 33.480
                                                            0.048
                                                                            0
##
     sys.child
## 2
## 1
```

Second, the inverse of matrix, I created a big matrix and use **solve** to get its inverse. It turns that the speed for integer is faster than double.

```
# Inverse of matrix
# Integer
```

```
#Change data type
m <- apply(intm, 1, as.numeric)</pre>
#Examine type
typeof(m)
## [1] "double"
#Comparison
benchmark(
 integer=solve(intm),
 double=solve(m),
 replications=5
)
        test replications elapsed relative user.self sys.self user.child
## 2 double
                      5 1.719 1.064 3.204 0.212
                       5 1.615
## 1 integer
                                    1.000
                                              3.056
                                                       0.152
                                                                     0
## sys.child
## 2
## 1
```

Finally, the subset, I created a vector \mathbf{x} from 1 to 100000 and use **sample** to make a random index vector to subset \mathbf{x} . The result is that integer performs faster than doouble.

```
# Subsetting
# Integer
intx <- 1:1000000

#Examine type
typeof(intx)

## [1] "integer"</pre>
```

```
#Change data type
x <- as.numeric(x)

#Examine type
typeof(x)

## [1] "double"

#Comparison
benchmark(
integer=intx[sample(length(intx))],</pre>
```