# STOR 320 Programming II

Lecture 13

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#### Introduction to Functions

- Most Important Programming Skill in R
- Functions in R
  - Take Inputs
  - Do Calculations
  - Produce Outputs
- Control Structures Such as "If-else" Statements and Loops are Used in Functions
- Advantages
  - Memorable Names
  - Code Updates Occur in 1 Place
  - Makes Code Accessible by All

#### **Build-in R Functions**

- Before Writing a Function, Always Search for a Function That Does What You Want
- To See What a Function Does: ?dplyr::lag

 To Understand How the Function Works, Algorithmically: dplyr::lag

#### **Build-in R Functions**

```
dplyr::lag
## function (x, n = 1L, default = NA, order by = NULL, ...)
       if (!is.null(order by)) {
##
           return(with order(order by, lag, x, n = n, default = default))
       if (inherits(x, "ts")) {
           bad args("x", "must be a vector, not a ts object, do you want `stats::lag()`?")
       if (length(n) != 1 || !is.numeric(n) || n < 0) {
           bad args("n", "must be a nonnegative integer scalar, ",
               "not {type of(n)} of length {length(n)}")
       if (n == 0)
           return(x)
       xlen <- length(x)</pre>
       n <- pmin(n, xlen)</pre>
       out <- c(rep(default, n), x[seq len(xlen - n)])</pre>
       attributes(out) <- attributes(x)</pre>
## <bytecode: 0x0000000123d4f48>
## <environment: namespace:dplyr>
```

## Creating R Functions

General Form:

```
NAME = function(INPUTS){
    ACTIONS
    return(OUTPUT)
}
```

- Functions are Objects in R
- To Call Function: NAME(INPUTS)
- Create an Object to Save an Output from a Function
   OUTPUT=NAME(INPUTS)

# Example

- Example: Lag Operator
  - Used for Vectors According to Time (i.e Time Series Data)
  - Suppose a Vector Contains Information at Time = t
  - A Lagged Vector Contains Information at Time = t-k
     where k = Lag
  - Suppose y<sub>t</sub> = Value of a Car at Time t. Then, y<sub>t-k</sub> = Value of a Car at Time t-k

## Example

- Example: Lag Operator
  - Vector of Values

```
V = c(35, 32, 30, 31, 27, 25)
```

Lagged Values for k=1

```
LV1 = c(NA, 35, 32, 30, 31, 27)
```

Lagged Values for k=2

```
LV2 = c(NA, NA, 35, 32, 30, 31)
```

- Want to Create a Function that:
  - Inputs Vector (x) and Lag (k)
  - Returns Lagged Vector

## Creating R Functions

- Example: Lag Operator
  - Attempt 1:

Attempt 2:

```
Uptown.Func2 = function(x,k){
    t=length(x)
    y1=x[1:(t-k)]
    y2=c(rep(NA,k),y1)
    return(y2)
}
```

Creating R Functions

Example: Lag Operator

```
Value=c(35, 32, 30, 31, 27, 25)
Uptown.Func1(x=Value)
  [1] NA 35 32 30 31 27
Uptown.Func2(x=Value, k=1)
   [1] NA 35 32 30 31 27
Uptown.Func1(x=Value, k=3)
   [1] NA NA NA 35 32 30
Uptown.Func2(x=Value, k=3)
       NA NA NA 35 32 30
```

## **Practicing Functions: 5 Summary**

- Computing Five Number Summary
  - Input Vector of Observations
  - Output Vector of Statistics

```
Summary.func = function(data){
    min=min(data)
    max=max(data)
    q1=quantile(data,0.25)
    q2=quantile(data,0.5)
    q3=quantile(data,0.75)
    y=c(min,q1,q2,q3,max)
    names(y)=c("Min","Q1","Q2","Q3","Max")
    return(y)
}
```

#### **Practicing Functions: T-Test**

- T-Test for Population Mean
  - Concept:
    - Null: Average # of Hours Spent Watching TV per Day is \_\_\_\_ in the USA
    - Alt: Average # of Hours Spent Watching TV per Day is not \_\_\_\_ in the USA
    - Does Data Provide Evidence that Alt is True

## **Practicing Functions: T-Test**

- T-Test for Population Mean
  - Process:
    - Specify α (Type 1 Error)
    - Compute Test Statistic

$$t_{s} = \frac{\bar{x} - \mu_{Guess}}{s / \sqrt{n}}$$

- Find P-value
- If P-value < α, Reject Null</li>

## **Building Functions**

- T-Test for Population Mean
  - Inputs
    - Vector of Observations (ob)
    - Null Hypothesis (h0)
    - Alpha (a)

- Output List
  - Test Statistic
  - P-value
  - Decision:
    - Reject
    - Fail to Reject
  - Plot Data and Null Guess

# **Building Functions**

- T-Test for Population Mean
  - Function in R

```
ttest = function(ob,h0,a){
 n=length(ob)
 ts=(mean(ob,na.rm=T)-h0)/(sd(ob,na.rm=T)/sqrt(n))
 pval=2*pt(-abs(ts),df=n-1)
 conclusion = if(pval<a){</pre>
           "Reject Null Hypothesis"
          } else{
           "Fail to Reject Null Hypothesis"
 plot=ggplot() +
  geom bar(aes(x=ob),fill="lightskyblue1") +
  theme_minimal() + geom_vline(xintercept=h0)
 return(list(ts=ts,pval=pval,
     conclusion=conclusion,plot=plot))
```

#### Results

- T-Test for Population Mean
  - Guess 4 Hours

ttest(ob=forcats::gss cat\$tvhours,h0=4,a=0.05)

```
## $ts

## [1] -57.74276

##

## $pval

## [1] 0

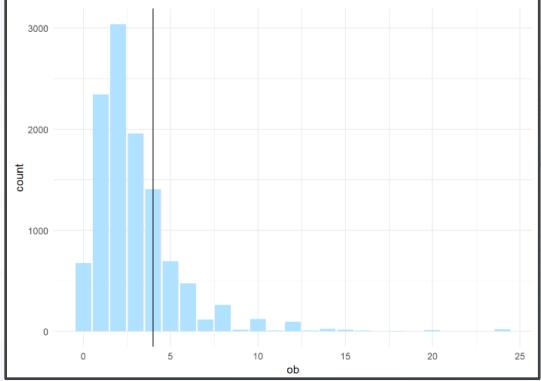
##

## $conclusion

## [1] "Reject Null Hypothesis"

##

## $plot
```



#### Results

- T-Test for Population Mean
  - Guess 3 Hours

```
\texttt{ttest(ob=forcats::gss\_cat\$tvhours,h0=3,a=0.05)}
## $ts
## [1] -1.089392
## $pval
## [1] 0.2759934
## $conclusion
## [1] "Fail to Reject Null Hypothesis"
## $plot
 3000
 2000
 1000
```

## **Practicing Functions: CLT**

- Central Limit Theorem
  - Let X be a Random Variable
  - $\bar{X} \sim N\left(\mu_X, \frac{\sigma_X}{\sqrt{n}}\right)$  where n = sample size
  - One of the Biggest Results in Statistics
  - Foundational in Introductory Statistics Classes

## **Practicing Functions: CLT**

- Central Limit Theorem
  - Inputs
    - n=sample size
    - S=number of simulations
    - D=distribution={1,2}

- Output List
  - Theoretical Mean
  - Theoretical Standard Error

$$SE(\bar{X}) = \frac{\sigma_X}{\sqrt{n}}$$

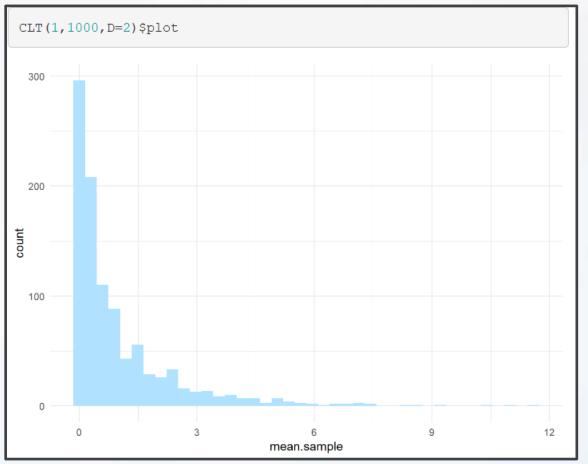
- Simulated Mean
- Simulated Standard Error
- Figure: Histogram of  $\bar{X}$

# Writing Functions

```
CLT = function(n,S,D)
if(D==1){
  initial=rnorm(1000000)
 } else if(D==2){
  initial=rgamma(1000000)
t.mean=mean(initial)
t.se=sd(initial)/sqrt(n)
mean.sample=rep(NA,S)
for(k in 1:S){
  if(D==1){
   sample=rnorm(n)
  } else if(D==2){
   sample=rgamma(n)
  mean.sample[k]=mean(sample)
s.mean=mean(mean.sample)
s.se=sd(mean.sample)
 plot=ggplot()+
  geom_histogram(aes(x=mean.sample),
  fill=skyblue1)+theme minimal()
 OUT=list(theory.mean=t.mean,
      theory.se=t.se,
      sim.mean=s.mean,
      sim.se=s.se,
      plot=plot)
return(OUT)
```

#### Results

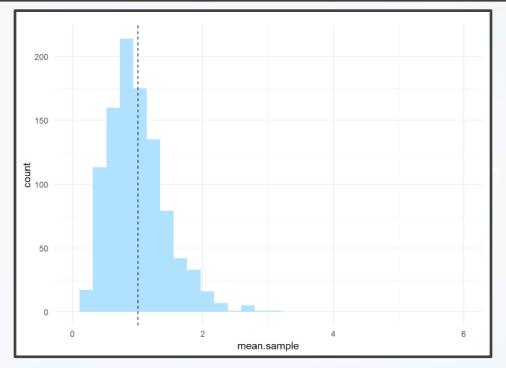
- Central Limit Theorem
  - Plot of Gamma Population



#### Results: n=10

- Central Limit Theorem
  - Sampling Distribution of  $\bar{X}$  when n=10

```
OUT=CLT(10,1000,D=2)
OUT[[5]]+scale_x_continuous(limits=c(0,6))+
   geom_vline(xintercept=OUT$theory.mean,linetype="dashed")
```



\$theory.mean Γ17 1.001844

\$theory.se [1] 0.4472895

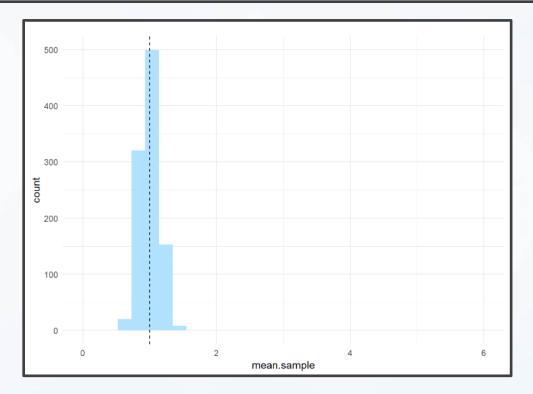
\$sim.mean
[1] 1.031698

\$sim.se [1] 0.4547647

#### Results: n=100

- Central Limit Theorem
  - Sampling Distribution of  $\bar{X}$  when n=100

```
OUT=CLT(100,1000,D=2)
OUT[[5]]+scale_x_continuous(limits=c(0,6))+
geom_vline(xintercept=OUT$theory.mean,linetype="dashed")
```



\$theory.mean [1] 0.999974

\$theory.se
[1] 0.141634

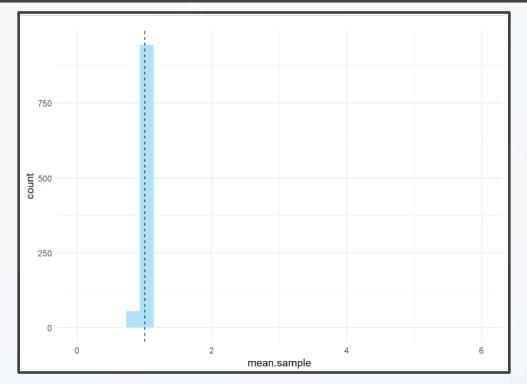
\$sim.mean
[1] 1.001891

\$sim.se [1] 0.1438765

#### Results: n=1000

- Central Limit Theorem
  - Sampling Distribution of  $\bar{X}$  when n=1000

```
OUT=CLT(1000,1000,D=2)
OUT[[5]]+scale_x_continuous(limits=c(0,6))+
   geom_vline(xintercept=OUT$theory.mean,linetype="dashed")
```



\$theory.mean

[1] 0.9992336

\$theory.se [1] 0.04454787

\$sim.mean [1] 0.9979233

\$sim.se [1] 0.04499497