HW1

2023-10-02

Quesiton 3

```
\# x \sim iid N(0,2)
# Set n's
n = c(10, 100, 1000)
# Set number of simulations
sims = 10000
# Create a bunch of sample means
xbars = sapply(1:sims, function(sims){
  \# Simulate x's for each value of n
  x = sapply(1:length(n), function(i){
    rnorm(n[i], 0, 2)
  })
  # Calculate sample means for each n
  xbar = sapply(1:length(n), function(i){
   1/n[i]*sum(x[[i]])
 })
})
# flip dimensions of xbars to make it more intuitive
xbars = t(xbars)
# Name columns for pretty output
colnames(xbars) = c("10", "100", "1000")
```

```
# Calculate means for each n
x_means = colMeans(xbars)

# Calculate variances for each n
x_vars = sapply(1:length(n), function(i){var(xbars[,i])})
names(x_vars) = c("10", "100", "1000")

# Make the output nice
(samp_mean_normal_df = data.frame(x_means, x_vars))
```

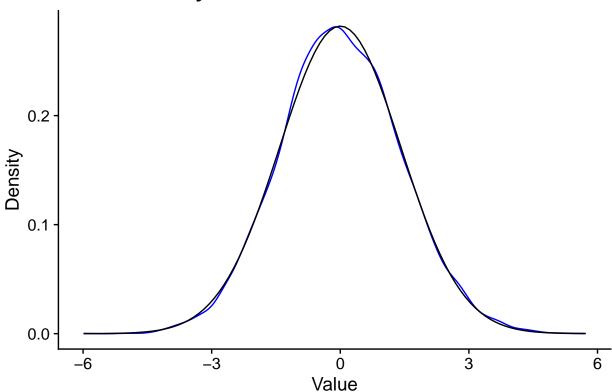
Part a)

```
##
              x means
                           x vars
## 10
       -0.0098341845 0.399823582
## 100
       0.0019700092 0.039568019
## 1000 -0.0002792192 0.003943637
# Calculate which simulations have absolute value greater than 0.1
xbars_large = abs(xbars) > 0.1
# Calculate the percentage for each n
large_percentage = sapply(1:length(n), function(i){
 percentage = sum(xbars_large[,i]) / nrow(xbars_large)
})
names(large_percentage) = c("10", "100", "1000")
large_percentage
Part b)
```

```
## 10 100 1000
## 0.8752 0.6203 0.1122
```

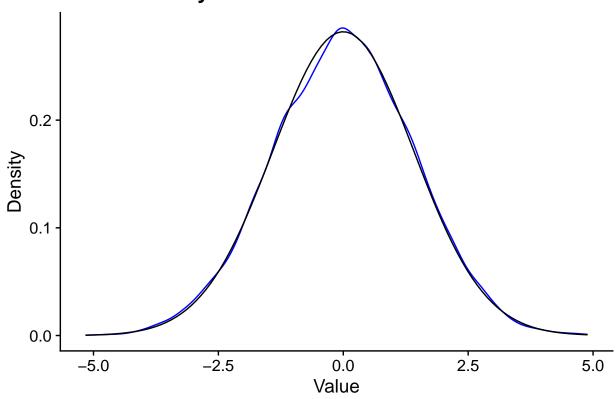
```
# Same as earlier part but instead of sample mean its dividing by sqrt(n)
xbars_sqrt = sapply(1:sims, function(sims){
  # Simulate x's for each value of n
  x = sapply(1:length(n), function(i){
   rnorm(n[i], 0, sqrt(2))
 })
  # Calculate sample means for each n
 xbar = sapply(1:length(n), function(i){
    1/sqrt(n[i])*sum(x[[i]])
 })
})
# flip dimensions of xbars to make it more intuitive
xbars_sqrt = t(xbars_sqrt)
# Name columns for pretty output
colnames(xbars_sqrt) = c("10", "100", "1000")
colnames(xbars) = c("10", "100", "1000")
# Rearrange data frame to make it work for graphing
xbars_graph = pivot_longer(as.data.frame(xbars_sqrt), cols = c("10", "100", "1000"))
# Graph kernal density
xbars_graph |> filter(name == "10") |>
  ggplot(aes(x = value)) +
 geom_density(color = "blue") +
```

```
stat_function(fun = dnorm, args = list(sd = sqrt(2))) +
xlab("Value") +
ylab("Density") +
labs(title = "Kernal Density for N = 10") +
cowplot::theme_cowplot()
```

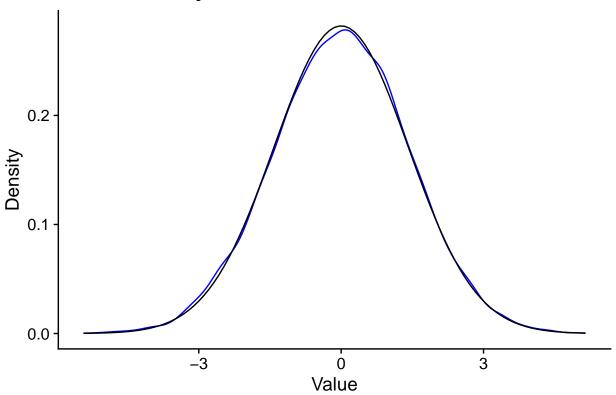


Part c)

```
xbars_graph |> filter(name == "100") |>
ggplot(aes(x = value)) +
geom_density(color = "blue") +
stat_function(fun = dnorm, args = list(sd = sqrt(2))) +
xlab("Value") +
ylab("Density") +
labs(title = "Kernal Density for N = 100") +
cowplot::theme_cowplot()
```



```
xbars_graph |> filter(name == "1000") |>
ggplot(aes(x = value)) +
geom_density(color = "blue") +
stat_function(fun = dnorm, args = list(sd = sqrt(2))) +
xlab("Value") +
ylab("Density") +
labs(title = "Kernal Density for N = 1000") +
cowplot::theme_cowplot()
```



```
# Calculate ks.test
sapply(1:length(n), function(i){
   ks.test(xbars_sqrt[,i], "pnorm", 0, sqrt(2))
})
```

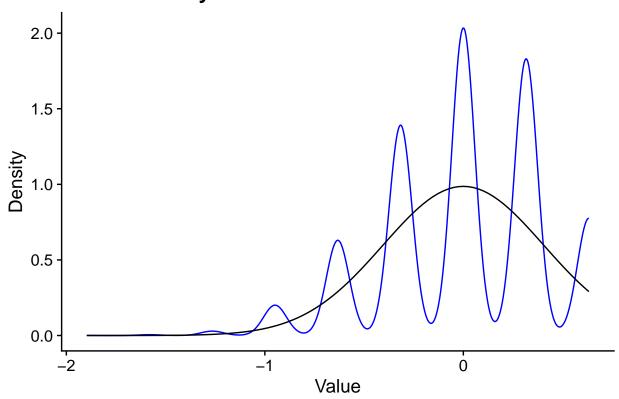
```
[,1]
##
               0.008916184
## statistic
## p.value
               0.4044093
## alternative "two-sided"
## method
               "Asymptotic one-sample Kolmogorov-Smirnov test"
## data.name
               "xbars_sqrt[, i]"
## data
               list,2
               FALSE
## exact
##
               [,2]
               0.008979004
## statistic
## p.value
               0.3956392
## alternative "two-sided"
               "Asymptotic one-sample Kolmogorov-Smirnov test"
## method
## data.name
               "xbars_sqrt[, i]"
## data
               list,2
## exact
               FALSE
##
               [,3]
               0.006980379
## statistic
## p.value
               0.714499
## alternative "two-sided"
               "Asymptotic one-sample Kolmogorov-Smirnov test"
## method
```

```
xbars_binom = sapply(1:sims, function(sims){
  # Simulate x's for each value of n
  x = sapply(1:length(n), function(i){
   rbinom(n[i], 1, 0.8)
  })
  # Calculate sample means for each n
  xbar = sapply(1:length(n), function(i){
    1/n[i]*sum(x[[i]])
  })
})
# flip dimensions of xbars to make it more intuitive
xbars_binom = t(xbars_binom)
# Name columns for pretty output
colnames(xbars_binom) = c("10", "100", "1000")
# Calculate means for each n
x_means_binom = colMeans(xbars_binom)
# Calculate variances for each n
x vars binom = sapply(1:length(n), function(i){var(xbars binom[,i])})
names(x_vars_binom) = c("10", "100", "1000")
# Make the output nice
(samp_mean_binom_df = data.frame(x_means_binom, x_vars_binom))
```

Part d)

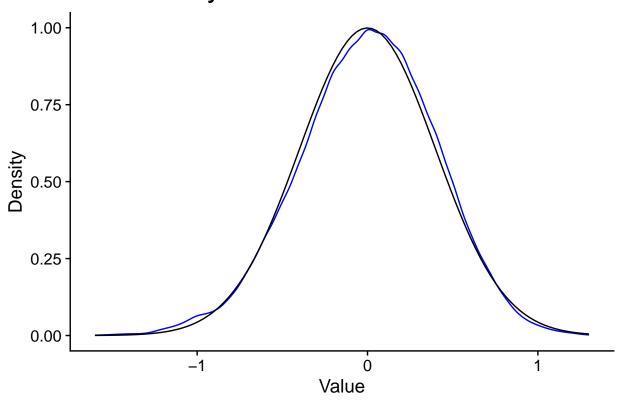
```
## x_means_binom x_vars_binom
## 10 0.799660 0.0161474991
## 100 0.799436 0.0016083227
## 1000 0.799884 0.0001626534
```

```
# Same as earlier part but instead of sample mean its dividing by sqrt(n)
xbars_sqrt_binom = sapply(1:sims, function(sims){
  \# Simulate x's for each value of n
  x = sapply(1:length(n), function(i){
    rbinom(n[i], 1, 0.8)
  })
  \# Calculate sample means for each n
  xbar = sapply(1:length(n), function(i){
    sqrt(n[i])*(1/n[i]*sum(x[[i]]) - 0.8)
  })
})
# flip dimensions of xbars to make it more intuitive
xbars_sqrt_binom = t(xbars_sqrt_binom)
# Name columns for pretty output
colnames(xbars_sqrt_binom) = c("10", "100", "1000")
# Get empirical variance for comparison distribution
x_vars_sqrt_binom = sapply(1:length(n), function(i){var(xbars_sqrt_binom[,i])})
# Rearrange data frame to make it work for graphing
xbars_sqrt_binom_graph = pivot_longer(as.data.frame(xbars_sqrt_binom), cols = c("10", "1000", "1000"))
# Graph kernal density
xbars_sqrt_binom_graph |> filter(name == "10") |>
  ggplot(aes(x = value)) +
  geom_density(color = 'blue') +
  stat function(fun = dnorm, args = list(sd = sqrt(x vars sqrt binom[1]))) +
  xlab("Value") +
  ylab("Density") +
  labs(title = "Kernal Density for N = 10") +
  cowplot::theme cowplot()
```

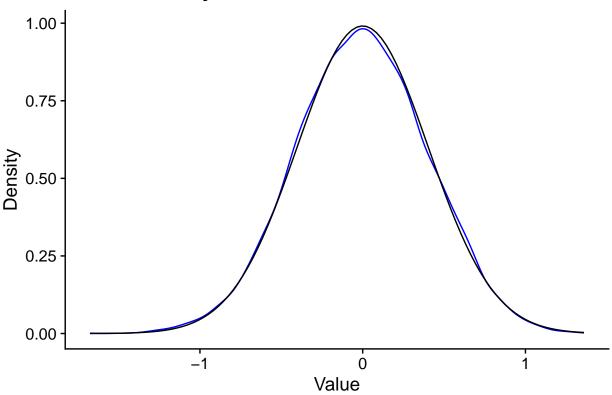


Part e)

```
xbars_sqrt_binom_graph |> filter(name == "100") |>
ggplot(aes(x = value)) +
geom_density(color = 'blue') +
stat_function(fun = dnorm, args = list(sd = sqrt(x_vars_sqrt_binom[2]))) +
xlab("Value") +
ylab("Density") +
labs(title = "Kernal Density for N = 100") +
cowplot::theme_cowplot()
```



```
xbars_sqrt_binom_graph |> filter(name == "1000") |>
ggplot(aes(x = value)) +
geom_density(color = 'blue') +
stat_function(fun = dnorm, args = list(sd = sqrt(x_vars_sqrt_binom[3]))) +
xlab("Value") +
ylab("Density") +
labs(title = "Kernal Density for N = 1000") +
cowplot::theme_cowplot()
```



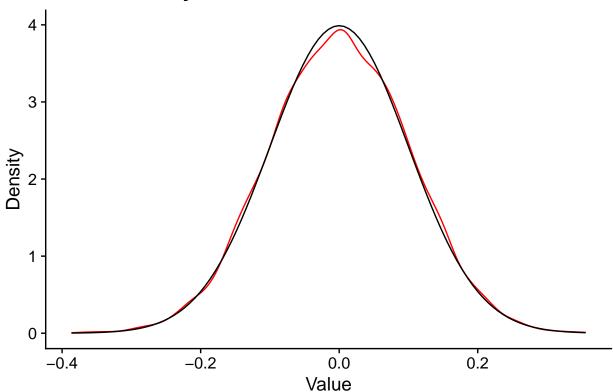
```
# Calculate ks.test
sapply(1:length(n), function(i){
  ks.test(xbars_sqrt_binom[,i], "pnorm", 0, sqrt(0.16))
## Warning in ks.test.default(xbars_sqrt_binom[, i], "pnorm", 0, sqrt(0.16)): ties
## should not be present for the Kolmogorov-Smirnov test
## Warning in ks.test.default(xbars_sqrt_binom[, i], "pnorm", 0, sqrt(0.16)): ties
## should not be present for the Kolmogorov-Smirnov test
## Warning in ks.test.default(xbars_sqrt_binom[, i], "pnorm", 0, sqrt(0.16)): ties
## should not be present for the Kolmogorov-Smirnov test
##
               [,1]
## statistic
               0.1728
               0
## p.value
## alternative "two-sided"
               "Asymptotic one-sample Kolmogorov-Smirnov test"
## method
## data.name
               "xbars_sqrt_binom[, i]"
## data
               list,2
## exact
               FALSE
               [,2]
##
## statistic
               0.064
## p.value
               0
```

```
## alternative "two-sided"
## method
             "Asymptotic one-sample Kolmogorov-Smirnov test"
## data.name "xbars_sqrt_binom[, i]"
## data
             list,2
## exact
              FALSE
##
              [,3]
## statistic 0.02337185
              3.60097e-05
## p.value
## alternative "two-sided"
## method
           "Asymptotic one-sample Kolmogorov-Smirnov test"
## data.name "xbars_sqrt_binom[, i]"
              list,2
## data
              FALSE
## exact
Question 5)
# Set n and sims
N = 100
sims = 10000
# Simulate data a bunch of times
beta = sapply(1:sims, function(i){
  # Simulate data
  x = matrix(c(rep(1,N), rnorm(N, 0, 2)), ncol = 2)
  e = rnorm(N, 0, 1)
  # Calculate y
  y = 2*x[,2] + e
  # Calculate OLS estimator
  beta = solve(t(x)%*%x)%*%t(x)%*%y
})
# Calculate mean and variance
(means = sapply(1:2, function(i){mean(beta[i,])}))
a)
## [1] 0.0002779961 1.9996014094
(vars = sapply(1:2, function(i){var(beta[i,])}))
## [1] 0.010137806 0.002628666
```

beta_graph = as.data.frame(t(beta))

```
# Graph densities
beta_graph |> ggplot(aes(x = V1)) +
  geom_density(color = 'red') +
  stat_function(fun = dnorm, args = list(sd = 0.1)) +
  xlab("Value") +
  ylab("Density") +
  labs(title = "Kernal Density for Beta 0") +
  cowplot::theme_cowplot()
```

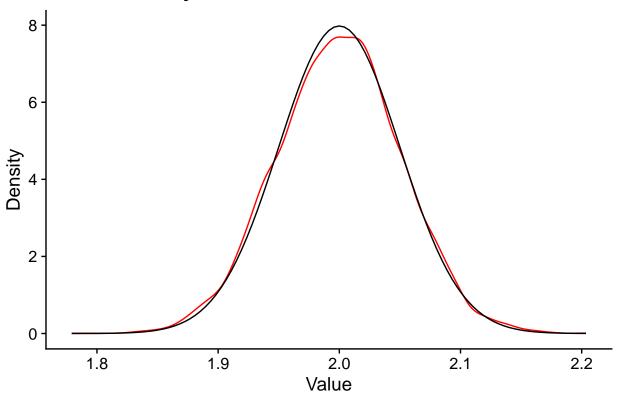
Kernal Density for Beta 0



b)

```
beta_graph |> ggplot(aes(x = V2)) +
  geom_density(color = 'red') +
  stat_function(fun = dnorm, args = list(mean = 2, sd = 0.05)) +
  xlab("Value") +
  ylab("Density") +
  labs(title = "Kernal Density for Beta 1") +
  cowplot::theme_cowplot()
```

Kernal Density for Beta 1



```
# Run ks test for both betas
ks.test(beta_graph$V1, 'pnorm', 0, 0.1)
##
    Asymptotic one-sample Kolmogorov-Smirnov test
##
##
## data: beta_graph$V1
## D = 0.0084663, p-value = 0.4704
## alternative hypothesis: two-sided
ks.test(beta_graph$V2, 'pnorm', 2, 0.05)
##
   Asymptotic one-sample Kolmogorov-Smirnov test
##
##
## data: beta_graph$V2
## D = 0.01326, p-value = 0.05939
## alternative hypothesis: two-sided
```

set vector of n's n = c(10, 100, 1000)

```
# Calculate betas for each n
beta_ns = sapply(1:sims, function(i){
  sapply(1:3, function(i){
    # Simulate data
    x = matrix(c(rep(1,n[i]), rnorm(n[i], 0, 2)), ncol = 2)
    e = rnorm(n[i], 0, 1)
    # Calculate y
    y = 2*x[,2] + e
    # Calculate OLS estimator
    beta = solve(t(x)%*%x)%*%t(x)%*%y
    # Centered distribution of betas
    sqrt(n[i])*(beta - c(0,2))
  })
})
# KS test for n = 10 for beta 0 then beta1
ks.test(beta_ns[1,], 'pnorm', 0, 1)
c)
##
   Asymptotic one-sample Kolmogorov-Smirnov test
##
##
## data: beta ns[1, ]
## D = 0.020516, p-value = 0.0004417
## alternative hypothesis: two-sided
ks.test(beta_ns[2,], 'pnorm', 0, 0.5)
##
  Asymptotic one-sample Kolmogorov-Smirnov test
##
## data: beta_ns[2, ]
## D = 0.032813, p-value = 8.895e-10
## alternative hypothesis: two-sided
# n = 100
ks.test(beta_ns[3,], 'pnorm', 0, 1)
##
  Asymptotic one-sample Kolmogorov-Smirnov test
## data: beta_ns[3, ]
## D = 0.0060459, p-value = 0.8581
## alternative hypothesis: two-sided
```

```
ks.test(beta_ns[4,], 'pnorm', 0, 0.5)
##
##
   Asymptotic one-sample Kolmogorov-Smirnov test
##
## data: beta_ns[4, ]
## D = 0.0077811, p-value = 0.5801
## alternative hypothesis: two-sided
# n = 1000
ks.test(beta_ns[5,], 'pnorm', 0, 1)
##
   Asymptotic one-sample Kolmogorov-Smirnov test
## data: beta_ns[5, ]
## D = 0.010874, p-value = 0.1878
## alternative hypothesis: two-sided
ks.test(beta_ns[6,], 'pnorm', 0, 0.5)
##
##
  Asymptotic one-sample Kolmogorov-Smirnov test
## data: beta_ns[6, ]
## D = 0.0093916, p-value = 0.341
## alternative hypothesis: two-sided
# same as part a, but we're now making the errors ~ U(0,2)
# Set n and sims
N = 100
sims = 10000
# Simulate data a bunch of times
beta = sapply(1:sims, function(i){
  # Simulate data
  x = matrix(c(rep(1,N), rnorm(N, 0, 2)), ncol = 2)
  e = runif(N, 0, 2)
  # Calculate y
  y = 2*x[,2] + e
  # Calculate OLS estimator
  beta = solve(t(x)%*%x)%*%t(x)%*%y
})
# Calculate mean and variance
(means = sapply(1:2, function(i){mean(beta[i,])}))
```

```
## [1] 1.000933 1.999613
(vars = sapply(1:2, function(i){var(beta[i,])}))
## [1] 0.0034218555 0.0008504284
# Same as part c, but we're changing the erros again. We could have written a function which takes the
# set vector of n's
n = c(10, 100, 1000)
# Calculate betas for each n
beta_ns = sapply(1:1, function(q){
  sapply(1:3, function(i){
   # Simulate data
    x = matrix(c(rep(1,n[i]), rnorm(n[i], 0, 2)), ncol = 2)
    e = runif(n[i], 0, 2)
    # Calculate y
    y = 2*x[,2] + e
    # Calculate OLS estimator
    beta = solve(t(x)%*%x)%*%t(x)%*%y
    # Centered distribution of betas
    sqrt(n[i])*(beta - c(1,2))
  })
})
# KS test for n = 10 for beta 0 then beta1
ks.test(beta_ns[1,], 'pnorm', 0, 1/3)
e)
##
## Exact one-sample Kolmogorov-Smirnov test
## data: beta_ns[1, ]
## D = 0.98104, p-value = 0.03793
## alternative hypothesis: two-sided
ks.test(beta_ns[2,], 'pnorm', 0, I(1/3+1/4))
## Exact one-sample Kolmogorov-Smirnov test
## data: beta_ns[2, ]
## D = 0.69851, p-value = 0.603
## alternative hypothesis: two-sided
```

d)

```
# n = 100
ks.test(beta_ns[3,], 'pnorm', 0, 1/3)
##
   Exact one-sample Kolmogorov-Smirnov test
## data: beta_ns[3, ]
## D = 0.91858, p-value = 0.1628
## alternative hypothesis: two-sided
ks.test(beta_ns[4,], 'pnorm', 0, I(1/3+1/4))
##
## Exact one-sample Kolmogorov-Smirnov test
## data: beta_ns[4, ]
## D = 0.57283, p-value = 0.8543
## alternative hypothesis: two-sided
# n = 1000
ks.test(beta_ns[5,], 'pnorm', 0, 1/3)
##
## Exact one-sample Kolmogorov-Smirnov test
##
## data: beta_ns[5, ]
## D = 0.83894, p-value = 0.3221
## alternative hypothesis: two-sided
ks.test(beta_ns[6,], 'pnorm', 0, I(1/3+1/4))
##
## Exact one-sample Kolmogorov-Smirnov test
##
## data: beta_ns[6, ]
## D = 0.5194, p-value = 0.9612
## alternative hypothesis: two-sided
# Now add correlation between x and epsilon
# Set n and sims
N = 100
sims = 10000
# Simulate data a bunch of times
beta = sapply(1:sims, function(i){
# Simulate data
```

```
x_e = MASS::mvrnorm(n = 100, c(0,0), matrix(c(4,.4,.4,1), nrow = 2))

# Calculate y
y = 2*x_e[,1] + x_e[,2]

# Calculate OLS estimator
x = matrix(c(rep(1,N), x_e[,1]), ncol = 2)
beta = solve(t(x)%*%x)%*%t(x)%*%y

})

colMeans(t(beta))
```

[1] -0.001494128 2.099817750

e)