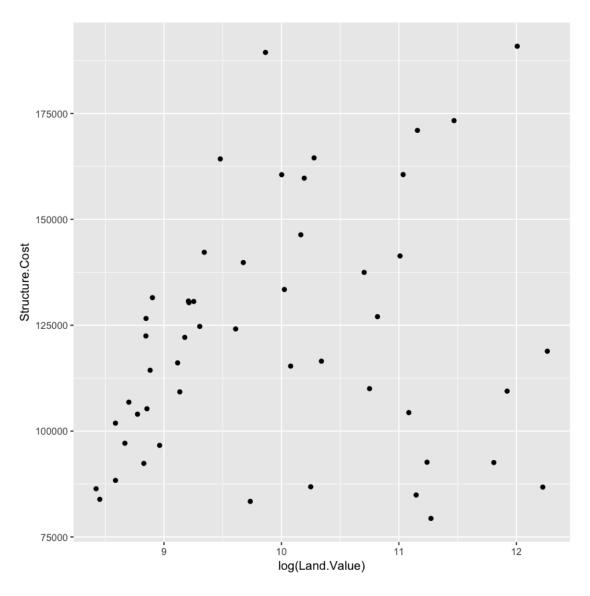
lab3q1

March 20, 2017

```
In [1]: ## Question 1
                         library(ggplot2)
                         housing <- read.csv("landdata-states.csv")</pre>
                         hp2001Q1 <- subset (housing, Date == 2001.25)
                          ggplot (hp2001Q1,
                                                 aes(y = Structure.Cost, x = log(Land.Value))) +
                                       geom_point()
                          ## Linear model
                         hp2001Q1$pred.SC <- predict(lm(Structure.Cost ~ log(Land.Value), data = hp2
                         p1 \leftarrow ggplot(hp2001Q1, aes(x = log(Land.Value), y = Structure.Cost))
                         p1 + geom_point(aes(color = Home.Value)) +
                                      geom_line(aes(y = pred.SC))
                          ## Kmeans
                         km = kmeans(hp2001Q1[,4:6],3)
                         km
                          clusters = as.factor(km$cluster)
                         ggplot(hp2001Q1, aes(x = Home.Value, y = Land.Value, color = clusters)) + geometric + ge
                          ## knn
                         data(iris)
                         library(class)
                         iris$Species = as.character(iris$Species)
                          iris$Species[iris$Species == "setosa"] = 1
                          iris$Species[iris$Species == "versicolor"] = 2
                          iris$Species[iris$Species == "virginica"] = 3
                          iris$Species = as.factor(iris$Species)
                         training <- sample(1:nrow(iris), 0.7*nrow(iris), replace = F)</pre>
```

```
train.iris <- iris[training,]
test.iris <- iris[-training,]
cl = train.iris$Species</pre>
```

knn.model = knn(train.iris,test.iris,cl,k = 3)



K-means clustering with 3 clusters of sizes 6, 26, 19

Cluster means:

```
Home.Value Structure.Cost Land.Value
1 288639.2 128640.2 159998.67
2 120017.2 109528.7 10488.35
3 179902.1 137320.8 42581.11
```

Clustering vector:

143 226 380 532 685 907 991 1144 1329 1482 1635 1788 1941 2094 2247 2400 2 3 1 1 3 3 2553 2706 2859 3012 3165 3318 3471 3624 3777 3930 4083 4236 4389 4542 4695 4848 5001 5154 5307 5460 5613 5766 5919 6072 6225 6378 6531 6684 6837 6990 7143 7296 7449 7602 7765

Within cluster sum of squares by cluster:
[1] 30063376761 15906513733 37833925012
(between_SS / total_SS = 76.0 %)

Available components:

[1] "cluster" "centers" "totss" "withinss" "tot.withinss"

[6] "betweenss" "size" "iter" "ifault"

