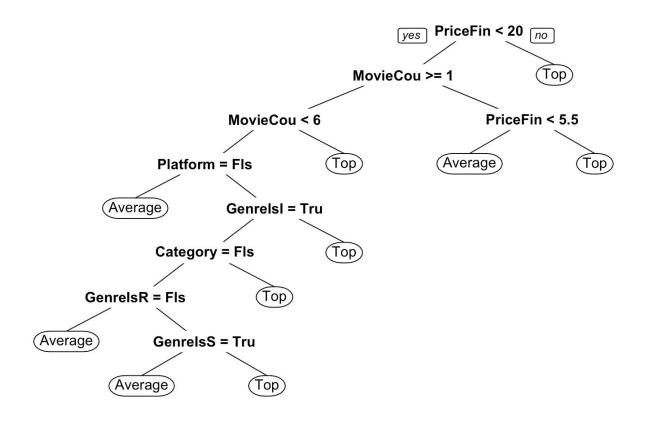
Part 1

A.



B.

PriceFinal was the most important/predictive variable in my model.

> tree\$variable.importance		
PriceFinal	MovieCount	GenreIsIndie
10.6025912	7.5222707	5.3305541
PlatformMac	MacReqsHaveMin	GenreIsStrategy
3.3753650	3.2411175	3.2088889
LinuxReqsHaveMin	${\tt CategoryIncludeLevelEditor}$	PlatformLinux
2.4661463	2.1725338	2.0904249
GenreIsRPG	ScreenshotCount	MacReqsHaveRec
1.7780791	1.5134488	1.0548016
IsFree	PublisherCount	GenreIsFreeToPlay
0.8323040	0.7586616	0.7356454
AchievementCount	CategorySinglePlayer	GenreIsSports
0.6768997	0.5939116	0.4330653
DeveloperCount	CategoryCoop	
0.3794024	0.3169524	

C.

> printcp(tree)

Classification tree:

rpart(formula = score ~ ., data = training, method = "class")

Variables actually used in tree construction:

- [1] CategoryIncludeLevelEditor GenreIsIndie
- [4] GenreIsStrategy

MovieCount

GenreIsRPG PlatformMac

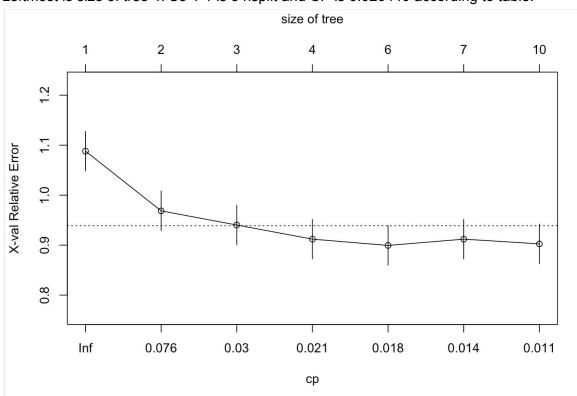
[7] PriceFinal

Root node error: 318/636 = 0.5

n = 636

	CP	nsplit	rel	error	xerror	xstd
1	0.141509	0	1.	.00000	1.08805	0.039499
2	0.040881	1	0.	.85849	0.96855	0.039633
3	0.022013	2	0	.81761	0.94025	0.039582
4	0.020440	3	0.	. 79560	0.91195	0.039499
5	0.015723	5	0.	. 75472	0.89937	0.039451
6	0.012579	6	0.	.73899	0.91195	0.039499
7	0.010000	9	0.	.70126	0.90252	0.039464
>						

Leftmost is size of tree 4. So 4-1 is 3 nsplit and CP is 0.020440 according to table.



> confusionMatrix(tree.pred, testing\$score)

Confusion Matrix and Statistics

Reference

Prediction Average Top

Average 95 63 Top 61 93

Accuracy : 0.6026

95% CI: (0.5459, 0.6573)

No Information Rate: 0.5

P-Value [Acc > NIR] : 0.0001731

Kappa : 0.2051

Mcnemar's Test P-Value: 0.9284440

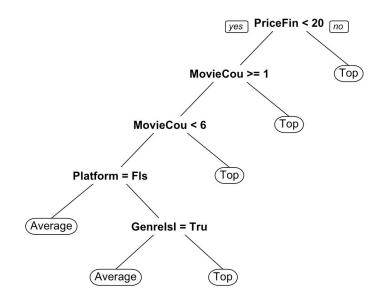
Sensitivity: 0.6090 Specificity: 0.5962 Pos Pred Value: 0.6013 Neg Pred Value: 0.6039 Prevalence: 0.5000

Detection Rate : 0.3045 Detection Prevalence : 0.5064

Balanced Accuracy: 0.6026

'Positive' Class : Average

E. Generate pruned tree using optimal complexity parameter of **0.020440** from question 1C.



> confusionMatrix(tree.pruned, testing\$score)

Confusion Matrix and Statistics

Reference

Prediction Average Top Average 92 63 Top 64 93

Accuracy : 0.5929

95% CI: (0.5362, 0.648)

No Information Rate: 0.5

P-Value [Acc > NIR] : 0.0006078

Kappa : 0.1859

Mcnemar's Test P-Value : 1.0000000

Sensitivity: 0.5897 Specificity: 0.5962 Pos Pred Value: 0.5935 Neg Pred Value: 0.5924 Prevalence: 0.5000

Detection Rate : 0.2949 Detection Prevalence : 0.4968

Balanced Accuracy: 0.5929

'Positive' Class : Average

The unpruned tree has a depth of 7, while the pruned tree has a depth of 4. Because we are making the pruned tree more flexible by making it simpler, the pruned tree is slightly less accurate. The pruned tree had an accuracy score of 0.5929 compared to the unpruned tree 0.6026.

A.

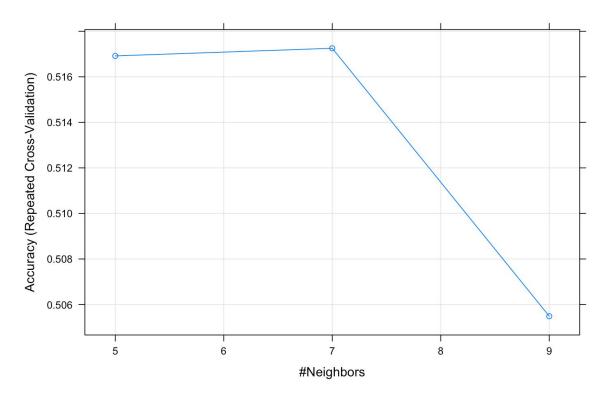
> varImp(knn1)

ROC curve variable importance

only 20 most important variables shown (out of 58)

	Importance
PriceFinal	100.00
GenreIsIndie.False	84.15
GenreIsIndie.True	84.15
PlatformMac.False	61.41
PlatformMac.True	61.41
DeveloperCount	56.59
MacReqsHaveMin.True	47.76
MacReqsHaveMin.False	47.76
PublisherCount	47.41
CategoryMultiplayer.True	43.21
CategoryMultiplayer.False	43.21
PlatformLinux.False	36.39
PlatformLinux.True	36.39
GenreIsAdventure.True	31.84
GenreIsCasual.False	31.84
GenreIsCasual.True	31.84
GenreIsAdventure.False	31.84
MacReqsHaveRec.True	27.29
MacReqsHaveRec.False	27.29
${\tt CategoryIncludeLevelEditor.True}$	22.74
>	

B.



Build a model with the maximum training accuracy would take 7 neighbors as seen in the graph.

> confusionMatrix(pred, testing\score)

Confusion Matrix and Statistics

Reference

Prediction Average Top

Average 82 58 Top 60 84

Accuracy : 0.5845

95% CI: (0.5248, 0.6424)

No Information Rate: 0.5

P-Value [Acc > NIR] : 0.002599

Kappa: 0.169

Mcnemar's Test P-Value: 0.926652

Sensitivity: 0.5775

Specificity: 0.5915

Pos Pred Value : 0.5857

Neg Pred Value : 0.5833

Prevalence: 0.5000

Detection Rate : 0.2887

Detection Prevalence : 0.4930

Balanced Accuracy: 0.5845

'Positive' Class : Average

D.

I would want to optimize my model for sensitivity because it would be worse to miss games that were top games, then it would be to make a bad guess and, in this case, write about a game that does not become a top hit. As discussed in the lecture, sensitivity is used when it costs more to miss something than to guess incorrectly. Whereas, specificity is concerned with the classifier correctly identifying when something is not the target class.