

Exam 1

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Question 1:

As predictive models get more and more complex the bias gets smaller and smaller. This is a good thing. Explain then why we don't simply always use the most complex, lowest bias model possible.

You don't always use the most complex, lowest bias model possible because of the bias-variance tradeoff. The more complex model you have, the lower the bias will be, but the higher the variance will be in exchange. Variance refers to how much \hat{f} would change if we estimated it using a different training set, and ideally you would not want your \hat{f} to change much between training sets. This is so that you have a generalizable model that applies to a variety of situations, and isn't trained specifically just to represent your training data. So, you want to minimize both the variance and the bias, not just a model with super lower variance and high bias or low bias and high variance.

Question 2:

Best subset regression checks all possible combinations of variables in the model. Why don't we just always use all subsets regression to choose our model?

First of all, best subsets regression suffers from computational limitations; the number of possible models that you would have to consider grows rapidly as p increases, which means that it quickly becomes computationally infeasible. Also, when p is large, that means you have a large search space, which means you have a higher chance of finding models with a high variance. So your models will look good on training data, but are not actually generalizable. A large search space can lead to overfitting and high covariance of coefficient estimates.

Furthermore, best subsets regression might not do very well in high dimensional datasets. You're going to have to use enormous computational power, and it's probably going to overfit the data.

Question 3:

K-fold cross validation is generally considered the best way to do cross-validation. Given an example of any scenario where you might not want to use k-fold cross validation and instead use a different method for cross validation.

When the size of the dataset is small, it's better to use LOOCV. So an example of this would be analyzing survey data from a small data science class of 15 or so people. You wouldn't have many observations, and so LOOCV would be preferable since it'll use more training samples than K-fold would.

Question 4:

Explain what leave-one-out cross validation (LOOCV) is and briefly explain one reason why we prefer k-fold cross validation where k is smaller than n .

With leave one out cross validation, you split the set of observations into two parts, but instead of having two

subsets which are of comparative size, only one observation is used for the validation set, and everything else is in the training set. You then fit a model on the training observations, and you make a prediction for the test error of the excluded observation. You then repeat that process by selecting a different observation point for the validation data, and you continue to train the model on your training set. The LOOCV estimate ends up being the average of the n test error estimates. LOOCV can be preferred for bias reduction. Each training set contains $n-1$ observations, which is pretty close to the number in the full data set. So, we might prefer it in situations where we want to reduce bias as much as possible.

Question 5

Stepwise Selection:

```
install.packages("leaps")
```

```
## Installing package into '/home/bultok/R/x86_64-pc-linux-gnu-library/4.0'
## (as 'lib' is unspecified)
```

```
library(leaps)
library(MASS)
load(url("https://github.com/gjm112/LoyolaTeaching/blob/main/data_exam1_2022.RData?raw=TRUE"))

lower <- lm(y ~ 1, data = data_exam1)
upper <- lm(y ~ ., data = data_exam1)
a <- lm(y ~ ., data = data_exam1)
```

First, I did a step AIC procedure to try to see if I could find a model with the lowest AIC. But since this data is high-dimensional, the AIC became negative infinity, since there are more features than observations; stepwise was producing very overfit functions. So we don't know if the coefficients included are actually relevant to predicting y , or if there are not problems with multi-collinearity.

I omitted the code for the stepwise procedure from the code chunk since it produced a ridiculous amount of output (obviously)

But, this is the code that I used after the above lines:

```
stewpise = stepAIC(lower, scope = list(lower = lower, upper = upper), direction = "both")
```

The model with the lowest AIC is:

```
y ~ X959 + X889 + X881 + X462 + X907 + X562 + X68 + X804 + X169 + X418 + X769 + X442 + X625 + X395
+ X181 + X280 + X315 + X140 + X389 + X65 + X153 + X771 + X26 + X634 + X597 + X428 + X410 + X164 +
X847 + X940 + X966 + X293 + X183 + X463 + X592 + X806 + X211 + X848 + X964 + X677 + X916 + X55 +
X48 + X777 + X110 + X535 + X837 + X397 + X102 + X54 + X532 + X674 + X887 + X30 + X491 + X262 +
X158 + X696 + X879 + X295 + X201 + X388 + X852 + X993 + X274 + X725 + X358 + X383 + X166 + X32 +
X118 + X464 + X97 + X391 + X666 + X273 + X987 + X61 + X834 + X765 + X117 + X139 + X768 + X733 +
X64 + X606 + X453 + X698 + X78 + X414 + X830 + X434 + X382 + X537 + X416 + X150 + X564 + X47 + X1
```

I decided to just do a forward stepwise procedure, since forward stepwise is more flexible and would hopefully give a more accurate model:

```
regfit.fwd = regsubsets(y ~., data = data_exam1, nvmax = 19, method = "forward")
```

```
## Warning in leaps.setup(x, y, wt = wt, nbest = nbest, nvmax = nvmax, force.in =  
## force.in, : 901 linear dependencies found
```

```
coef(regfit.fwd, 19)
```

```
## (Intercept)          X68          X140          X169          X181          X280  
##  5.8289399  1.0617298  0.9240698 -1.6552823  1.1484161 -0.8223481  
##          X315          X389          X395          X418          X442          X462  
## -0.8518664 -0.6025580  1.4072209  1.0267707 -1.5976523  1.1450267  
##          X562          X625          X769          X804          X881          X889  
##  1.3537767 -1.1168536  0.8748100 -1.3859445 -2.2934293 -2.0182441  
##          X907          X959  
## -1.3137278  1.3363291
```

The best model from forward stepwise selection with 19 variables is

```
(Intercept) X68 X140 X169 X181 X280 X315 5.8289399 1.0617298 0.9240698 -1.6552823 1.1484161  
-0.8223481 -0.8518664 X389 X395 X418 X442 X462 X562 X625 -0.6025580 1.4072209 1.0267707  
-1.5976523 1.1450267 1.3537767 -1.1168536 X769 X804 X881 X889 X907 X959 0.8748100 -1.3859445  
-2.2934293 -2.0182441 -1.3137278 1.33632
```

```
library(glmnet)
```

```
## Loading required package: Matrix
```

```
## Loaded glmnet 4.1-4
```

```
x = model.matrix(y ~., data_exam1)[, -1]  
y = data_exam1$y
```

```
set.seed(1)  
train = sample(1:nrow(x), nrow(x) / 2)  
test = (-train)  
y.test = y[test]
```

Lasso Regression

```
set.seed(1)  
cv.out = cv.glmnet(x[train, ], y[train], alpha = 1)  
lam = cv.out$lambda.min  
lam
```

```
## [1] 1.315547
```

the model is $5.60620407 + X_{91} (0.13176002) + X_{94} (0.66938879) + X_{190} (-0.04044894) + X_{299} (-0.19255332) + X_{381} (-0.11070250) + X_{397} (0.64652487) + X_{528} (-0.032056650) + X_{783} (-0.008343261) + X_{889} (-0.538795882) + X_{959} (0.097651340)$

```
lasso.mod = glmnet(x[train, ], y[train], alpha = 1, lambda = 1.315547)
lass.coef = predict(lasso.mod, type = "coefficients", s=lam)
lass.coef[lass.coef[,1]!=0]
```

```
## <sparse>[ <logic> ] : .M.sub.i.logical() maybe inefficient
```

```
## [1] 5.606204066 0.131760018 0.669388785 -0.040448939 -0.192553318
## [6] -0.110702495 0.646524868 -0.032056650 -0.008343261 -0.538795882
## [11] 0.097651340
```

```
lass.coef
```

```
## 1001 x 1 sparse Matrix of class "dgCMatrix"
##                               s1
## (Intercept)  5.606204066
## X1          .
## X2          .
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## X885	.
## X886	.
## X887	.
## X888	.
## X889	-0.538795882
## X890	.
## X891	.
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## X958	.
## X959	0.097651340
## X960	.
## X961	.
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## X964	.
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## X992      .
## X993      .
## X994      .
## X995      .
## X996      .
## X997      .
## X998      .
## X999      .
## X1000     .
```

```
lasso.pred <- predict(lasso.mod, s = lam, newx = x[test,])
mean((lasso.pred - y.test)^2)
```

```
## [1] 21.8452
```

Ridge Regression

I did cross-validation to find the best lambda value, finding it as 1.315547.

```
set.seed(1)
cv.out1 = cv.glmnet(x[train, ], y[train], alpha = 0)
lam1 = cv.out$lambda.min
lam1
```



```
## [1] 1.315547
```

```
ridge.mod = glmnet(x, y, alpha = 0, lambda = 1.315547)  
dim(coef(ridge.mod))
```

```
## [1] 1001    1
```

```
ridge.mod$lambda[1]
```

```
## [1] 1.315547
```

```
coef(ridge.mod)[, 1]
```

##	(Intercept)	X1	X2	X3	X4
##	6.089391e+00	-4.560122e-02	2.548536e-03	-7.094993e-02	5.122055e-02
##	X5	X6	X7	X8	X9
##	-2.821592e-02	3.147503e-02	-1.308064e-02	-2.573540e-02	1.276530e-04
##	X10	X11	X12	X13	X14
##	-4.537072e-02	1.854451e-02	-3.597661e-02	-7.233811e-02	-1.037137e-03
##	X15	X16	X17	X18	X19
##	2.669354e-02	-2.084842e-02	-5.795093e-02	-7.152850e-02	-5.144660e-02
##	X20	X21	X22	X23	X24
##	-4.955320e-02	4.145706e-03	-3.598434e-03	2.701516e-04	4.805684e-02
##	X25	X26	X27	X28	X29
##	1.248141e-02	2.177975e-02	-4.695262e-02	3.606428e-02	-6.874100e-02
##	X30	X31	X32	X33	X34
##	-4.117462e-03	3.131608e-02	7.788934e-02	1.621698e-02	2.694962e-02
##	X35	X36	X37	X38	X39
##	2.473994e-02	-1.194042e-03	-2.367175e-02	3.963763e-02	5.513972e-02
##	X40	X41	X42	X43	X44
##	-3.579352e-02	6.343920e-02	3.325401e-02	-5.040209e-02	3.166122e-02
##	X45	X46	X47	X48	X49
##	-4.720694e-02	7.067574e-03	-7.817491e-02	-7.050879e-02	4.911324e-02
##	X50	X51	X52	X53	X54
##	1.844625e-02	-6.903010e-02	1.092693e-02	-5.772206e-02	1.341286e-02
##	X55	X56	X57	X58	X59
##	-1.036963e-02	-6.651975e-02	3.381721e-02	-8.748316e-02	1.819786e-02
##	X60	X61	X62	X63	X64
##	5.424168e-02	1.771153e-02	-6.267707e-02	-4.008766e-02	-1.928326e-03
##	X65	X66	X67	X68	X69
##	-4.482660e-02	-3.780057e-02	4.232880e-02	1.030632e-01	-5.868373e-02
##	X70	X71	X72	X73	X74
##	-6.219004e-03	-4.473851e-02	1.545461e-02	2.758604e-02	-4.116162e-02
##	X75	X76	X77	X78	X79
##	-3.716117e-02	7.219002e-03	-6.943233e-02	9.253062e-02	-4.025142e-02
##	X80	X81	X82	X83	X84
##	5.866104e-02	3.389455e-03	-3.455803e-03	1.608868e-02	3.014186e-02
##	X85	X86	X87	X88	X89
##	3.041363e-02	-9.413927e-02	-8.629784e-03	3.182163e-02	-3.445998e-02
##	X90	X91	X92	X93	X94
##	1.720817e-03	9.367182e-02	-6.379659e-02	5.235304e-02	1.141558e-01
##	X95	X96	X97	X98	X99
##	4.550485e-03	-8.285211e-02	2.501391e-02	5.762270e-03	-4.197631e-02
##	X100	X101	X102	X103	X104
##	-3.311424e-02	7.785670e-02	-1.563571e-02	-4.656809e-02	2.994331e-02
##	X105	X106	X107	X108	X109
##	8.396094e-02	-7.101964e-03	-1.903076e-03	3.175431e-02	-2.917621e-02
##	X110	X111	X112	X113	X114
##	-2.318703e-02	1.186889e-02	5.088147e-02	5.859377e-03	4.204426e-02
##	X115	X116	X117	X118	X119
##	-6.790972e-02	-1.945834e-02	-4.628877e-03	-5.212763e-02	-1.357669e-02
##	X120	X121	X122	X123	X124
##	-1.293385e-02	-1.674505e-02	2.276261e-02	6.327264e-02	-7.620998e-02
##	X125	X126	X127	X128	X129

##	-3.097635e-03	8.243850e-03	-4.415282e-02	2.625118e-02	-3.851222e-02
##	X130	X131	X132	X133	X134
##	2.340332e-02	-5.892964e-02	9.570762e-03	1.719826e-02	-1.613259e-02
##	X135	X136	X137	X138	X139
##	-6.690973e-02	-5.757253e-02	1.015115e-02	-3.672560e-02	-6.817290e-03
##	X140	X141	X142	X143	X144
##	5.468010e-03	4.461670e-02	-1.779961e-02	-1.017862e-02	-7.660082e-02
##	X145	X146	X147	X148	X149
##	5.313811e-02	-2.115336e-02	-2.499626e-02	9.657225e-02	-2.665327e-02
##	X150	X151	X152	X153	X154
##	2.721252e-02	2.500314e-02	-5.050785e-02	-3.701475e-02	6.873843e-02
##	X155	X156	X157	X158	X159
##	9.979130e-03	2.914388e-02	-1.762852e-02	2.031218e-02	-4.802763e-02
##	X160	X161	X162	X163	X164
##	-4.892060e-02	6.181094e-03	-1.879357e-02	-3.805240e-03	-5.343638e-03
##	X165	X166	X167	X168	X169
##	9.051201e-02	4.925104e-02	-3.716513e-03	-2.969698e-02	-1.134576e-01
##	X170	X171	X172	X173	X174
##	-2.328530e-02	4.403674e-02	-1.525679e-02	-2.520025e-02	2.641348e-02
##	X175	X176	X177	X178	X179
##	4.068104e-02	7.038430e-02	1.022942e-01	-2.761564e-02	-8.300210e-03
##	X180	X181	X182	X183	X184
##	3.671236e-02	2.037429e-02	5.921763e-02	-1.616852e-02	-3.255575e-02
##	X185	X186	X187	X188	X189
##	-3.253981e-02	8.478193e-02	-8.667688e-02	6.552816e-03	-8.321407e-02
##	X190	X191	X192	X193	X194
##	-1.229873e-01	7.326831e-02	2.635266e-02	-2.378197e-02	-5.577467e-03
##	X195	X196	X197	X198	X199
##	4.627540e-03	1.488198e-01	-1.073895e-01	-9.652723e-02	-2.398745e-02
##	X200	X201	X202	X203	X204
##	6.261051e-02	-6.944357e-02	-8.534394e-02	-6.239260e-02	-1.097053e-03
##	X205	X206	X207	X208	X209
##	-1.362429e-02	7.511317e-02	2.608948e-02	1.056504e-02	-1.248420e-01
##	X210	X211	X212	X213	X214
##	1.694475e-02	-2.505236e-03	-5.146092e-02	-6.664386e-02	2.618997e-02
##	X215	X216	X217	X218	X219
##	-6.282810e-02	-5.141056e-02	-2.109464e-02	-3.290813e-02	4.643093e-02
##	X220	X221	X222	X223	X224
##	-9.306864e-03	6.850454e-03	-3.608217e-02	8.844661e-02	-4.624285e-02
##	X225	X226	X227	X228	X229
##	6.724184e-03	-1.900792e-02	2.596397e-02	5.163254e-02	4.436549e-02
##	X230	X231	X232	X233	X234
##	-5.260988e-02	2.244141e-02	8.109473e-02	-9.347921e-02	-6.427242e-02
##	X235	X236	X237	X238	X239
##	2.367690e-02	6.218407e-03	4.610679e-02	5.583538e-03	9.081019e-02
##	X240	X241	X242	X243	X244
##	-6.601586e-02	-2.304738e-02	-2.567316e-02	-3.523254e-03	6.464611e-02
##	X245	X246	X247	X248	X249
##	5.004898e-02	1.287796e-02	-2.498644e-02	2.435091e-02	6.063777e-02
##	X250	X251	X252	X253	X254
##	-4.997905e-02	-6.016275e-02	2.688482e-02	6.133215e-02	2.040690e-02

##	X255	X256	X257	X258	X259
##	5.081023e-02	1.049660e-01	-2.118347e-02	7.433740e-03	3.648414e-02
##	X260	X261	X262	X263	X264
##	1.068442e-02	-6.176403e-03	2.992119e-02	7.482477e-03	-4.159417e-02
##	X265	X266	X267	X268	X269
##	4.499439e-02	1.052220e-01	-1.676508e-02	-2.210922e-02	-1.926806e-02
##	X270	X271	X272	X273	X274
##	2.232968e-02	-1.266976e-03	1.727994e-02	-1.366974e-02	1.610937e-02
##	X275	X276	X277	X278	X279
##	3.264451e-02	-1.810868e-02	1.251296e-01	4.271185e-02	5.163134e-02
##	X280	X281	X282	X283	X284
##	-7.847923e-03	-3.222281e-02	-4.128684e-02	-6.186629e-03	-5.114668e-02
##	X285	X286	X287	X288	X289
##	-3.864616e-02	-5.120954e-02	2.175819e-02	-4.153327e-02	1.162589e-02
##	X290	X291	X292	X293	X294
##	-6.395672e-02	5.165058e-02	1.101114e-02	6.126477e-02	5.852834e-02
##	X295	X296	X297	X298	X299
##	1.037729e-01	6.041836e-03	2.251173e-02	9.458879e-02	-1.121838e-02
##	X300	X301	X302	X303	X304
##	3.165441e-02	2.382751e-02	-3.358140e-03	5.982766e-02	-2.696259e-03
##	X305	X306	X307	X308	X309
##	8.757765e-02	-4.070649e-02	2.260372e-02	2.168582e-02	-6.430610e-02
##	X310	X311	X312	X313	X314
##	-3.240309e-02	-3.414193e-02	4.144218e-03	9.302344e-03	1.330254e-02
##	X315	X316	X317	X318	X319
##	1.875170e-02	8.753185e-02	2.802896e-03	1.416944e-01	2.086394e-04
##	X320	X321	X322	X323	X324
##	-1.887014e-02	8.055645e-02	-1.245736e-02	6.789471e-02	-2.056609e-02
##	X325	X326	X327	X328	X329
##	7.893402e-03	-2.393104e-02	5.862184e-02	2.696305e-02	-3.789898e-02
##	X330	X331	X332	X333	X334
##	-1.001195e-02	7.401437e-02	-3.348837e-02	-1.657466e-02	-6.817829e-02
##	X335	X336	X337	X338	X339
##	1.725469e-02	2.960667e-02	-6.246169e-02	-6.074140e-02	2.616682e-02
##	X340	X341	X342	X343	X344
##	-3.659178e-02	-6.441300e-02	-1.456280e-02	2.677478e-02	-2.400717e-02
##	X345	X346	X347	X348	X349
##	1.769825e-02	4.380807e-02	-2.142301e-03	-1.553330e-02	-4.518999e-02
##	X350	X351	X352	X353	X354
##	1.559659e-03	-7.641393e-02	4.251790e-02	-6.677130e-02	-8.156290e-03
##	X355	X356	X357	X358	X359
##	2.140940e-02	-6.366646e-02	-3.312095e-02	1.523818e-01	-5.080567e-02
##	X360	X361	X362	X363	X364
##	-4.268811e-02	3.776534e-02	1.240630e-01	-6.995003e-03	8.846592e-03
##	X365	X366	X367	X368	X369
##	-1.905698e-03	4.942342e-02	-4.857954e-02	-1.365569e-01	2.744980e-02
##	X370	X371	X372	X373	X374
##	-5.170340e-02	-1.392541e-02	-8.745321e-02	2.760037e-02	-6.549281e-02
##	X375	X376	X377	X378	X379
##	8.559980e-03	-2.144151e-02	4.146362e-02	5.387358e-02	-1.212799e-01
##	X380	X381	X382	X383	X384

##	-3.683541e-02	-3.588156e-02	5.531922e-02	-5.259996e-02	4.275551e-03
##	X385	X386	X387	X388	X389
##	9.249516e-02	3.846356e-02	-3.367228e-02	6.028533e-02	5.131490e-03
##	X390	X391	X392	X393	X394
##	-1.435622e-02	2.733393e-02	-3.476768e-03	-3.425775e-02	-5.870483e-02
##	X395	X396	X397	X398	X399
##	1.613948e-02	4.932019e-02	1.272001e-01	7.135422e-02	-2.159864e-02
##	X400	X401	X402	X403	X404
##	6.981039e-02	-3.858001e-02	-3.491469e-02	3.877886e-03	-3.945877e-02
##	X405	X406	X407	X408	X409
##	-1.501391e-02	-3.468984e-02	-1.563510e-02	-3.274572e-03	5.635707e-02
##	X410	X411	X412	X413	X414
##	1.767745e-02	2.755630e-02	-1.010898e-02	-3.569879e-02	-5.376168e-02
##	X415	X416	X417	X418	X419
##	1.699869e-02	2.530485e-02	2.367831e-02	3.718306e-02	1.232580e-04
##	X420	X421	X422	X423	X424
##	-2.761780e-02	1.406012e-03	-7.273022e-02	6.172031e-02	-4.852892e-02
##	X425	X426	X427	X428	X429
##	5.675246e-02	-2.607580e-02	1.906802e-02	-7.569992e-02	-2.551512e-02
##	X430	X431	X432	X433	X434
##	7.972895e-02	-5.744972e-02	-3.638531e-02	-3.213540e-03	-5.560584e-02
##	X435	X436	X437	X438	X439
##	1.038786e-02	1.683723e-02	-3.690930e-02	6.781736e-02	4.868535e-02
##	X440	X441	X442	X443	X444
##	-4.165488e-03	-1.110592e-03	-7.325837e-02	-3.261942e-03	4.723537e-02
##	X445	X446	X447	X448	X449
##	7.345883e-02	3.148358e-02	6.274631e-02	-2.219860e-02	2.207810e-02
##	X450	X451	X452	X453	X454
##	-7.287929e-02	-4.126214e-02	-4.494981e-02	8.058441e-03	5.774020e-02
##	X455	X456	X457	X458	X459
##	1.034363e-02	-2.337060e-02	3.813593e-02	1.447165e-02	2.279942e-02
##	X460	X461	X462	X463	X464
##	4.852359e-02	3.703035e-02	1.254287e-01	8.837801e-02	2.093773e-03
##	X465	X466	X467	X468	X469
##	-3.442027e-02	-3.647475e-02	-4.185995e-02	-1.974176e-02	-4.951807e-02
##	X470	X471	X472	X473	X474
##	5.376498e-03	1.690554e-02	4.602068e-05	-8.102650e-02	-2.411760e-02
##	X475	X476	X477	X478	X479
##	-7.723992e-02	-2.202253e-03	-1.770913e-02	6.271741e-02	-4.240298e-02
##	X480	X481	X482	X483	X484
##	-1.649026e-03	5.753745e-02	6.716665e-02	-1.032294e-01	2.742860e-02
##	X485	X486	X487	X488	X489
##	-3.855934e-02	-2.815885e-02	8.483368e-03	1.590991e-02	1.078503e-02
##	X490	X491	X492	X493	X494
##	-2.538859e-02	5.629614e-03	6.187885e-02	1.555117e-02	4.438323e-02
##	X495	X496	X497	X498	X499
##	-2.064228e-02	-2.331412e-02	8.261330e-02	-5.330616e-02	-7.229864e-02
##	X500	X501	X502	X503	X504
##	1.315395e-01	-6.377850e-02	-5.798858e-03	-2.198205e-02	-2.686903e-02
##	X505	X506	X507	X508	X509
##	-2.258608e-02	7.575239e-02	2.852933e-02	4.959634e-02	-6.738695e-03

##	X510	X511	X512	X513	X514
##	-4.781362e-02	-7.015072e-02	2.899582e-02	7.041381e-02	7.508980e-03
##	X515	X516	X517	X518	X519
##	-4.201484e-02	5.291336e-02	3.729307e-02	-5.783726e-02	-5.693410e-02
##	X520	X521	X522	X523	X524
##	8.497320e-02	-3.628182e-02	6.298211e-02	-3.668400e-02	-3.350182e-02
##	X525	X526	X527	X528	X529
##	-2.293385e-02	6.051010e-02	-3.101182e-02	-3.300478e-02	2.821832e-02
##	X530	X531	X532	X533	X534
##	2.209218e-02	-2.838413e-02	5.135393e-02	1.976659e-02	6.583822e-02
##	X535	X536	X537	X538	X539
##	-1.663232e-02	3.401808e-02	-1.597987e-02	7.305027e-02	-2.004515e-02
##	X540	X541	X542	X543	X544
##	-3.039414e-02	9.764798e-03	3.622685e-02	-2.410946e-02	4.865095e-02
##	X545	X546	X547	X548	X549
##	3.119953e-02	-1.686832e-02	1.075649e-01	-1.610299e-02	-1.236257e-01
##	X550	X551	X552	X553	X554
##	6.693173e-02	9.583006e-03	1.270519e-02	6.120516e-02	-1.881596e-02
##	X555	X556	X557	X558	X559
##	-4.684424e-02	4.403463e-02	1.000881e-01	-2.150870e-02	-3.374429e-02
##	X560	X561	X562	X563	X564
##	-3.189695e-02	-2.355717e-03	1.300824e-01	8.550405e-02	2.619064e-02
##	X565	X566	X567	X568	X569
##	-6.328449e-03	5.161452e-02	-9.933880e-03	-5.164999e-04	-6.470112e-03
##	X570	X571	X572	X573	X574
##	-5.864460e-02	-2.240050e-02	7.508363e-02	3.889876e-03	4.102525e-03
##	X575	X576	X577	X578	X579
##	2.373619e-02	1.294633e-01	-7.556885e-03	2.778113e-02	1.123347e-01
##	X580	X581	X582	X583	X584
##	-1.026992e-01	3.407560e-02	9.645412e-03	1.916149e-02	-4.485410e-02
##	X585	X586	X587	X588	X589
##	4.920429e-03	-3.752126e-02	2.649285e-02	4.983561e-02	2.800426e-02
##	X590	X591	X592	X593	X594
##	-1.428237e-02	-3.844820e-02	-7.316392e-03	4.240627e-03	-8.911309e-03
##	X595	X596	X597	X598	X599
##	3.973491e-02	1.178701e-01	-7.720871e-03	4.038794e-02	6.728378e-02
##	X600	X601	X602	X603	X604
##	-2.897725e-02	1.477915e-01	-3.191085e-02	-5.176530e-04	1.060767e-02
##	X605	X606	X607	X608	X609
##	1.305718e-02	3.899152e-02	-1.074894e-02	1.367337e-02	-3.648356e-02
##	X610	X611	X612	X613	X614
##	7.073940e-02	5.260232e-02	-3.243580e-02	2.977031e-03	-8.878343e-02
##	X615	X616	X617	X618	X619
##	4.826371e-02	-3.176779e-03	-4.204940e-02	-2.093391e-02	7.172025e-02
##	X620	X621	X622	X623	X624
##	-1.175854e-02	1.365247e-01	-4.190125e-02	-5.648104e-03	-5.535918e-02
##	X625	X626	X627	X628	X629
##	-7.030709e-02	4.632193e-02	7.673381e-02	-1.221974e-02	5.057213e-02
##	X630	X631	X632	X633	X634
##	7.518268e-02	6.957426e-02	-7.019493e-02	1.619386e-02	-2.168759e-02
##	X635	X636	X637	X638	X639

##	3.392340e-02	4.140015e-02	-1.254206e-02	2.524458e-02	-3.706777e-02
##	X640	X641	X642	X643	X644
##	3.390174e-03	-1.523154e-03	-3.459769e-02	-5.865275e-02	8.267947e-02
##	X645	X646	X647	X648	X649
##	-6.467757e-02	4.758451e-03	4.822972e-03	7.105190e-02	-2.280398e-02
##	X650	X651	X652	X653	X654
##	6.723283e-03	5.277623e-02	1.495820e-02	-1.138255e-02	4.115144e-02
##	X655	X656	X657	X658	X659
##	4.761195e-02	-4.358562e-02	3.832315e-03	3.060249e-02	1.363510e-03
##	X660	X661	X662	X663	X664
##	5.804950e-02	-3.333648e-04	-1.614620e-02	2.896445e-02	2.069340e-04
##	X665	X666	X667	X668	X669
##	5.702779e-02	1.057120e-02	2.071905e-02	-9.718523e-03	-7.811483e-04
##	X670	X671	X672	X673	X674
##	-8.850736e-02	-9.006361e-03	-3.301345e-02	-2.982653e-02	-1.944257e-02
##	X675	X676	X677	X678	X679
##	1.206554e-02	-5.255131e-02	5.657415e-02	-1.584078e-02	8.740665e-03
##	X680	X681	X682	X683	X684
##	-4.450857e-02	1.467700e-02	-9.985254e-03	4.273509e-02	-1.789419e-02
##	X685	X686	X687	X688	X689
##	-1.832097e-02	-1.848179e-02	3.331465e-02	5.533442e-03	9.735808e-03
##	X690	X691	X692	X693	X694
##	-1.746752e-03	1.461180e-02	-2.249471e-02	4.014098e-02	-8.717310e-02
##	X695	X696	X697	X698	X699
##	-4.716088e-02	6.981189e-02	3.851831e-03	-9.500737e-03	-5.110246e-02
##	X700	X701	X702	X703	X704
##	-1.433139e-02	-3.919722e-02	-2.995089e-02	5.727983e-02	2.251607e-02
##	X705	X706	X707	X708	X709
##	-4.498633e-02	-1.183621e-02	2.155812e-02	6.828155e-02	-5.528859e-03
##	X710	X711	X712	X713	X714
##	-1.632253e-02	2.031787e-02	-9.833363e-02	4.971139e-02	-3.461545e-02
##	X715	X716	X717	X718	X719
##	-2.265060e-02	3.526368e-02	6.387049e-02	-2.882719e-02	1.132460e-02
##	X720	X721	X722	X723	X724
##	2.169412e-02	-6.926310e-02	6.582650e-02	2.914810e-02	-3.876096e-02
##	X725	X726	X727	X728	X729
##	-4.394038e-02	-4.923672e-03	2.039611e-02	4.270144e-02	-5.673195e-03
##	X730	X731	X732	X733	X734
##	-3.673780e-02	5.415696e-03	5.372396e-02	9.420792e-03	1.398979e-02
##	X735	X736	X737	X738	X739
##	1.613242e-02	-6.673675e-03	9.473852e-03	-2.362723e-02	-6.911000e-03
##	X740	X741	X742	X743	X744
##	-3.282014e-02	-1.854860e-02	-1.544703e-03	4.314837e-02	1.321840e-02
##	X745	X746	X747	X748	X749
##	-6.900528e-04	2.181383e-02	-1.075672e-02	-1.353829e-02	1.535456e-02
##	X750	X751	X752	X753	X754
##	-7.135722e-02	-5.381154e-02	-5.964982e-03	-6.992877e-03	1.352382e-02
##	X755	X756	X757	X758	X759
##	7.012827e-02	-2.224347e-02	1.712620e-02	-5.714053e-03	-1.910433e-02
##	X760	X761	X762	X763	X764
##	5.479634e-02	3.034145e-02	-2.868591e-02	-3.680679e-02	-1.961633e-02

##	X765	X766	X767	X768	X769
##	3.576993e-03	-1.151127e-01	-1.876158e-02	3.380367e-03	5.893856e-02
##	X770	X771	X772	X773	X774
##	4.502666e-02	1.104148e-02	4.154110e-02	6.829509e-02	-3.357247e-02
##	X775	X776	X777	X778	X779
##	1.912706e-03	2.647066e-02	1.454531e-02	4.464222e-02	1.287837e-02
##	X780	X781	X782	X783	X784
##	3.494737e-02	1.265006e-02	-1.195310e-02	-5.407062e-02	-8.677286e-03
##	X785	X786	X787	X788	X789
##	-1.368783e-02	5.774024e-02	-2.943051e-02	-2.535654e-02	5.419085e-02
##	X790	X791	X792	X793	X794
##	-3.801427e-02	-4.584316e-02	-1.983026e-02	6.355585e-02	-6.106563e-02
##	X795	X796	X797	X798	X799
##	1.223160e-02	-1.534376e-03	-5.936168e-02	-1.843167e-02	-3.254771e-02
##	X800	X801	X802	X803	X804
##	9.116473e-03	-5.856663e-02	-4.267010e-03	2.513142e-02	-8.324084e-02
##	X805	X806	X807	X808	X809
##	1.441487e-02	8.789497e-04	1.650018e-02	8.915709e-02	1.263102e-02
##	X810	X811	X812	X813	X814
##	8.659973e-03	1.670368e-02	-4.776582e-02	-3.672885e-03	4.179946e-02
##	X815	X816	X817	X818	X819
##	1.069638e-03	1.642991e-02	2.379681e-02	3.779971e-02	-1.518080e-02
##	X820	X821	X822	X823	X824
##	-6.703634e-02	-5.768249e-03	-2.958030e-02	1.004523e-02	3.139700e-02
##	X825	X826	X827	X828	X829
##	4.698676e-02	5.807272e-02	4.772272e-02	-5.795298e-02	-2.142131e-02
##	X830	X831	X832	X833	X834
##	-2.176816e-02	-7.083455e-03	1.441852e-02	-1.488912e-02	-5.662829e-02
##	X835	X836	X837	X838	X839
##	-5.055883e-02	-2.817138e-02	2.439850e-02	-1.562503e-02	-4.520425e-02
##	X840	X841	X842	X843	X844
##	4.573818e-03	-3.132320e-03	-1.111703e-02	-4.089783e-02	3.554434e-02
##	X845	X846	X847	X848	X849
##	-4.695234e-02	2.212680e-02	1.620711e-02	-6.565560e-03	-5.749106e-02
##	X850	X851	X852	X853	X854
##	1.974847e-02	2.160327e-02	-1.627522e-02	-4.191646e-02	-7.023234e-02
##	X855	X856	X857	X858	X859
##	-1.300940e-02	-3.918302e-02	-5.579642e-02	3.508213e-02	3.097069e-02
##	X860	X861	X862	X863	X864
##	6.012384e-02	5.007037e-02	7.283432e-03	-3.991054e-03	-3.432225e-02
##	X865	X866	X867	X868	X869
##	-3.484013e-02	6.608920e-02	-7.884873e-02	-2.151452e-02	-5.454756e-03
##	X870	X871	X872	X873	X874
##	7.904581e-03	-1.041044e-02	-5.298208e-03	-1.884633e-02	1.296602e-02
##	X875	X876	X877	X878	X879
##	-6.866586e-02	1.763339e-02	-4.771665e-02	6.554724e-03	1.971925e-03
##	X880	X881	X882	X883	X884
##	5.034486e-02	-1.507179e-01	4.118716e-02	1.219052e-03	2.390702e-02
##	X885	X886	X887	X888	X889
##	-1.552242e-02	-3.218282e-02	8.738379e-02	5.080105e-02	-1.379932e-01
##	X890	X891	X892	X893	X894


```
## -4.219577e-02  8.351119e-02 -8.301334e-03  8.116669e-03  2.700451e-02
##           X895           X896           X897           X898           X899
##  6.486273e-02  3.574095e-02 -2.190381e-02  4.071088e-02  4.258393e-02
##           X900           X901           X902           X903           X904
## -6.186159e-02  1.200646e-01  1.567217e-02 -1.803865e-02  7.448683e-03
##           X905           X906           X907           X908           X909
##  8.357300e-02 -3.833535e-03 -1.570614e-01  6.623944e-02 -3.123559e-02
##           X910           X911           X912           X913           X914
##  7.753600e-02 -3.260707e-02 -1.050832e-02 -4.012580e-03 -6.948405e-03
##           X915           X916           X917           X918           X919
## -1.181122e-02  6.981476e-02  3.375249e-02 -3.382473e-03  6.522869e-02
##           X920           X921           X922           X923           X924
##  7.362210e-02  9.807710e-02  1.120671e-02  5.166859e-02 -2.208998e-02
##           X925           X926           X927           X928           X929
## -3.749099e-02 -9.995865e-03 -9.152830e-02 -4.861111e-02  3.709477e-02
##           X930           X931           X932           X933           X934
##  1.571583e-02 -3.921777e-02 -8.479789e-02 -3.780521e-03  7.861921e-02
##           X935           X936           X937           X938           X939
## -7.908303e-02 -3.035190e-03 -4.537824e-02 -5.436357e-02 -1.689915e-02
##           X940           X941           X942           X943           X944
##  3.468607e-02  4.148208e-03 -2.772144e-02 -4.366137e-02  5.162481e-02
##           X945           X946           X947           X948           X949
## -8.089292e-03 -1.514661e-03 -1.128415e-01 -3.859648e-02 -5.118973e-02
##           X950           X951           X952           X953           X954
## -3.726109e-02 -2.584366e-02  2.156953e-02 -2.219766e-02 -8.327907e-02
##           X955           X956           X957           X958           X959
## -5.164756e-03 -2.752255e-02  2.486825e-03 -8.100441e-02  1.312424e-01
##           X960           X961           X962           X963           X964
##  1.526268e-02  8.228648e-02 -9.675698e-03 -4.472605e-02 -2.468861e-02
##           X965           X966           X967           X968           X969
## -4.857841e-02  4.152457e-02  5.698161e-02  3.017360e-02  4.876449e-05
##           X970           X971           X972           X973           X974
## -4.636669e-02  1.559010e-02 -7.534974e-02 -4.779904e-03 -3.585161e-03
##           X975           X976           X977           X978           X979
##  4.657299e-02  3.682302e-02  1.307552e-02  4.155421e-02  1.868196e-02
##           X980           X981           X982           X983           X984
## -2.197971e-03  6.951517e-03 -3.537370e-02 -2.563264e-02 -3.283644e-03
##           X985           X986           X987           X988           X989
## -2.859790e-03 -3.856493e-02  8.267037e-02 -5.240554e-03 -3.067957e-02
##           X990           X991           X992           X993           X994
## -3.260888e-02 -1.421226e-02  8.845498e-03  5.306815e-02 -9.883008e-02
##           X995           X996           X997           X998           X999
##  8.922903e-02  4.347686e-03  2.628362e-02  9.635920e-03 -1.881914e-02
##           X1000
## -3.952854e-02
```

```
sqrt(sum(coef(ridge.mod)[, 1]^2))
```

```
## [1] 6.268897
```

```
ridge.pred <- predict(ridge.mod, s = lam, newx = x[test,])
mean((ridge.pred - y.test)^2)
```

```
## [1] 0.02245913
```

Which is the best model?

Since it has the least complexity, the lasso regression model would be best. Even though the test MSE of the lasso model is higher, being at 21.8542, and the ridge model having one of 0.02245913, I suspect that the ridge model likely only has such a small test MSE because it is still overfitting the data (it has 1,000 features after all!) So, we should go with the lasso model, so we may not have to worry about as much of the potential covariance issues that can occur in high-dimensional datasets; it will hopefully be more generalizable to testing data.

Question 6

```
data(mtcars)
library(boot)
cor(mtcars$mpg, mtcars$wt, method = "spearman")
```

```
## [1] -0.886422
```

```
spearman.func = function(data, i) {
  cor(mtcars$mpg[i], mtcars$wt[i], method = "spearman")
}

spearman.func (mtcars, 1:10)
```

```
## [1] -0.7339484
```

```
boot_strappin = boot(mtcars, statistic = spearman.func, R = 1000 )
boot_strappin
```

```
##
## ORDINARY NONPARAMETRIC BOOTSTRAP
##
##
## Call:
## boot(data = mtcars, statistic = spearman.func, R = 1000)
##
##
## Bootstrap Statistics :
##      original      bias      std. error
## t1*  -0.886422  0.01711517  0.05326715
```

The bias is 0.01375834 and the standard error is 0.05087421. ### Question 7

```
library(class)
data("mtcars")
set.seed(1)
splitt = sort(sample(nrow(mtcars), nrow(mtcars)*.7))
train_mtcars <-mtcars[splitt,]
test_mtcars <-mtcars[-splitt,]
knn.mod = knn(train = train_mtcars, test = test_mtcars, cl=train_mtcars$gear, k = 1)
knn.mod
```

```
## [1] 3 3 4 3 3 3 3 4 4 5
## Levels: 3 4 5
```

```
table(knn.mod, test_mtcars$gear)
```

```
##
## knn.mod 3 4 5
##      3 5 1 0
##      4 0 2 1
##      5 0 0 1
```

LOOCV to find the best k value:

```
nrow(train_mtcars)
```

```
## [1] 22
```

```
output = NULL
for (i in 1:nrow(train_mtcars)){
  valid = train_mtcars[i,]
  training2 = train_mtcars[-i,]
  knn1 = knn(train = valid, test = training2, cl = valid$gear, k = i)
  output[[i]] = knn1
}
```

```
## Warning in knn(train = valid, test = training2, cl = valid$gear, k = i): k = 2
## exceeds number 1 of patterns
```

```
## Warning in knn(train = valid, test = training2, cl = valid$gear, k = i): k = 3
## exceeds number 1 of patterns
```

```
## Warning in knn(train = valid, test = training2, cl = valid$gear, k = i): k = 4
## exceeds number 1 of patterns
```

```
## Warning in knn(train = valid, test = training2, cl = valid$gear, k = i): k = 5  
## exceeds number 1 of patterns
```

```
## Warning in knn(train = valid, test = training2, cl = valid$gear, k = i): k = 6  
## exceeds number 1 of patterns
```

```
## Warning in knn(train = valid, test = training2, cl = valid$gear, k = i): k = 7  
## exceeds number 1 of patterns
```

```
## Warning in knn(train = valid, test = training2, cl = valid$gear, k = i): k = 8  
## exceeds number 1 of patterns
```

```
## Warning in knn(train = valid, test = training2, cl = valid$gear, k = i): k = 9  
## exceeds number 1 of patterns
```

```
## Warning in knn(train = valid, test = training2, cl = valid$gear, k = i): k = 10  
## exceeds number 1 of patterns
```

```
## Warning in knn(train = valid, test = training2, cl = valid$gear, k = i): k = 11  
## exceeds number 1 of patterns
```

```
## Warning in knn(train = valid, test = training2, cl = valid$gear, k = i): k = 12  
## exceeds number 1 of patterns
```

```
## Warning in knn(train = valid, test = training2, cl = valid$gear, k = i): k = 13  
## exceeds number 1 of patterns
```

```
## Warning in knn(train = valid, test = training2, cl = valid$gear, k = i): k = 14  
## exceeds number 1 of patterns
```

```
## Warning in knn(train = valid, test = training2, cl = valid$gear, k = i): k = 15  
## exceeds number 1 of patterns
```

```
## Warning in knn(train = valid, test = training2, cl = valid$gear, k = i): k = 16  
## exceeds number 1 of patterns
```

```
## Warning in knn(train = valid, test = training2, cl = valid$gear, k = i): k = 17  
## exceeds number 1 of patterns
```

```
## Warning in knn(train = valid, test = training2, cl = valid$gear, k = i): k = 18  
## exceeds number 1 of patterns
```

```
## Warning in knn(train = valid, test = training2, cl = valid$gear, k = i): k = 19  
## exceeds number 1 of patterns
```

```
## Warning in knn(train = valid, test = training2, cl = valid$gear, k = i): k = 20  
## exceeds number 1 of patterns
```

```
## Warning in knn(train = valid, test = training2, cl = valid$gear, k = i): k = 21  
## exceeds number 1 of patterns
```

```
## Warning in knn(train = valid, test = training2, cl = valid$gear, k = i): k = 22  
## exceeds number 1 of patterns
```

output

```
## [[1]]
## [1] 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4
## Levels: 4
##
## [[2]]
## [1] 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4
## Levels: 4
##
## [[3]]
## [1] 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3
## Levels: 3
##
## [[4]]
## [1] 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3
## Levels: 3
##
## [[5]]
## [1] 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3
## Levels: 3
##
## [[6]]
## [1] 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4
## Levels: 4
##
## [[7]]
## [1] 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4
## Levels: 4
##
## [[8]]
## [1] 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4
## Levels: 4
##
## [[9]]
## [1] 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3
## Levels: 3
##
## [[10]]
## [1] 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3
## Levels: 3
##
## [[11]]
## [1] 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3
## Levels: 3
##
## [[12]]
## [1] 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3
## Levels: 3
##
## [[13]]
## [1] 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4
## Levels: 4
```

```
##
## [[14]]
## [1] 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4
## Levels: 4
##
## [[15]]
## [1] 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4
## Levels: 4
##
## [[16]]
## [1] 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3
## Levels: 3
##
## [[17]]
## [1] 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3
## Levels: 3
##
## [[18]]
## [1] 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3
## Levels: 3
##
## [[19]]
## [1] 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5
## Levels: 5
##
## [[20]]
## [1] 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5
## Levels: 5
##
## [[21]]
## [1] 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5
## Levels: 5
##
## [[22]]
## [1] 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4
## Levels: 4
```