

Principal Component Analysis: Appendix

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March 14, 2016

Appendix 2: US County Median Household Income analysis: detailed results

```

disp('Analysis of US counties data: 15 features')
clear

load('counties_data.mat')
load('counties_result.mat')
X = counties_data;
% Y = median household income 2007 - 2013
y = counties_result(:,2);

% Variable selection
% Population measures:
% 1: pop 2014: 2
% 2: pop 2010 April: 3
% 3: pop 2010: 5
% 4: Veterans: 23
% 5: Housing Units: 25
% 6: Private non farm employment (abs): 33
% 7: Total number of firms: 36
% 8: Manufacturing shipments (value) 43
% 9: Retail sales (value) 45
% 10: Accommodation and food services sales (value) 47
% 11: Foreign born persons (%): 19
% 12: High school graduate or higher (%): 21
% 13: Bachelor's degree or higher (%): 22
% 14: Persons below the poverty level (%): 31
% 15: Women owned firms (%): 42

numsamples = (size(X,1));
finalX = zeros(numsamples,15);
index_data = [2, 3, 5, 23, 25, 33, 36, 43, 45, 47, 19, 21, 22, 31,
42];
for i=1:15
    finalX(:,i) = X(:,index_data(i));
end

% Scaling and mean normalizing features
fig = figure
fig.Position = [100 100 950 950]
plotStateData2(finalX,y,'Counties 2: raw data')

finalX = meanNormalize(finalX);
finalX = scaleX(finalX);
disp('Feature covariance matrix')
cov_features = (finalX' * finalX) * (1 / (numsamples - 1))
[meanx varx] = meanVarsX(finalX);

fig = figure
fig.Position = [100 100 950 950]
plotStateData3(finalX,y,'Counties 2: norm,sc')

% Generating training and test datasets

```

```

[ Xtest, XTRAIN, Xtr1, Xtr2, Xtr3, Xtr4, Xtr5, Xtr6, Xtr7, Xtr8, Xtr9,
  Xtr10, ...
      Xte1, Xte2, Xte3, Xte4, Xte5, Xte6, Xte7, Xte8, Xte9,
  Xte10, ...
  ytest, yTRAIN, ytr1, ytr2, ytr3, ytr4, ytr5, ytr6, ytr7, ytr8, ytr9,
  ytr10, ...
      yte1, yte2, yte3, yte4, yte5, yte6, yte7, yte8, yte9,
  yte10 ...
      ] = createTestTrainSets_v2( finalX, y );

disp('Training models on training data...')
disp('Unregularised linear regression')
[b_lr res_norm_lr] = linearLS(XTRAIN,yTRAIN)
lm = linearRegression(XTRAIN,yTRAIN)
% Calculating R squared
RSS_lr = res_norm_lr^2;
totalRes = yTRAIN - mean(yTRAIN);
TSS_lr = norm(totalRes)^2;
disp('R Squared')
rsq_lr = 1 - RSS_lr / TSS_lr

disp('Ridge regression')
[ k, RR_rsqa_train, RR_rsqa_test, rr_opt_bvec, rr_opt_rsqa, RR_all_bvec ]
= ridgeFindK( Xtr1, Xtr2, Xtr3, Xtr4, Xtr5, ...
              Xtr6, Xtr7, Xtr8, Xtr9,
  Xtr10, ...
              ytr1, ytr2, ytr3, ytr4,
  ytr5, ...
              ytr6, ytr7, ytr8, ytr9,
  ytr10, ...
              Xte1, Xte2, Xte3, Xte4,
  Xte5, ...
              Xte6, Xte7, Xte8, Xte9,
  Xte10, ...
              yte1, yte2, yte3, yte4,
  yte5, ...
              yte6, yte7, yte8, yte9,
  yte10);

disp('Principle component analysis')
[ PCA_rsqa, PCA_train_rsqa, pcr_opt_rsqa, pcr_opt_bvec, Xtest_PCA,
  XTRAIN_PCA...
  b_pcr1, b_pcr2, b_pcr3, b_pcr4, b_pcr5, b_pcr6, b_pcr7, b_pcr8,
  b_pcr9, b_pcr10] ...
    = pcrFindNumComponents(Xtest, XTRAIN, ytest, yTRAIN);

disp('b vector: linear regression')
disp(b_lr)
rsq_lr
disp('Ridge regression: test rsqa')
RR_rsqa_test
disp('Ridge regression: training rsqa')
RR_rsqa_train
disp('Ridge regression: b vector')

```

```
rr_opt_bvec
rr_opt_rsq
disp('Ridge regression: all b vectors')
RR_all_bvec
disp('Principal component regression: test rsq')
disp(PCA_rsq)
disp('Principal component regression: training rsq')
disp(PCA_train_rsq)
disp('Principal component regression: b vector')
disp(pcr_opt_bvec)
pcr_opt_rsq
disp('Principal component regression: all b vectors')
b_pcr1
b_pcr2
b_pcr3
b_pcr4
b_pcr5
b_pcr6
b_pcr7
b_pcr8
b_pcr9
b_pcr10

save USCountiesAnalysis2_Results.mat Xtest_PCA Xtest ytest b_lr pcr_opt_bvec rr
```

Analysis of US counties data: 15 features

fig =

Figure (17) with properties:

```
Number: 17
Name: ''
Color: [0.9400 0.9400 0.9400]
Position: [440 378 560 420]
Units: 'pixels'
```

Use GET to show all properties

fig =

Figure (17) with properties:

```
Number: 17
Name: ''
Color: [0.9400 0.9400 0.9400]
Position: [100 100 950 950]
Units: 'pixels'
```

Use GET to show all properties

Normalising X...
Scaling features....

Feature covariance matrix

cov_features =

Columns 1 through 7

1.0000	0.9996	0.9996	0.9268	0.9933	0.9559	0.9815
0.9996	1.0000	1.0000	0.9253	0.9932	0.9557	0.9818
0.9996	1.0000	1.0000	0.9253	0.9932	0.9557	0.9818
0.9268	0.9253	0.9253	1.0000	0.9421	0.8816	0.8738
0.9933	0.9932	0.9932	0.9421	1.0000	0.9634	0.9744
0.9559	0.9557	0.9557	0.8816	0.9634	1.0000	0.9585
0.9815	0.9818	0.9818	0.8738	0.9744	0.9585	1.0000
0.8293	0.8291	0.8291	0.7525	0.8220	0.8283	0.7930
0.9759	0.9751	0.9751	0.9265	0.9771	0.9730	0.9715
0.8227	0.8214	0.8214	0.7869	0.8366	0.8683	0.8290
0.4730	0.4690	0.4690	0.4352	0.4698	0.4430	0.4682
0.0624	0.0621	0.0621	0.1253	0.0781	0.0788	0.0658
0.3228	0.3200	0.3200	0.3678	0.3425	0.3545	0.3239
-0.0628	-0.0618	-0.0618	-0.1009	-0.0646	-0.0673	-0.0696
0.2171	0.2175	0.2175	0.2681	0.2297	0.2001	0.1979

Columns 8 through 14

0.8293	0.9759	0.8227	0.4730	0.0624	0.3228	-0.0628
0.8291	0.9751	0.8214	0.4690	0.0621	0.3200	-0.0618
0.8291	0.9751	0.8214	0.4690	0.0621	0.3200	-0.0618
0.7525	0.9265	0.7869	0.4352	0.1253	0.3678	-0.1009
0.8220	0.9771	0.8366	0.4698	0.0781	0.3425	-0.0646
0.8283	0.9730	0.8683	0.4430	0.0788	0.3545	-0.0673
0.7930	0.9715	0.8290	0.4682	0.0658	0.3239	-0.0696
1.0000	0.8161	0.6317	0.3159	0.0569	0.2430	-0.0577
0.8161	1.0000	0.8457	0.4627	0.0893	0.3581	-0.0849
0.6317	0.8457	1.0000	0.4013	0.0594	0.2905	-0.0498
0.3159	0.4627	0.4013	1.0000	-0.1636	0.3579	-0.0508
0.0569	0.0893	0.0594	-0.1636	1.0000	0.6023	-0.6598
0.2430	0.3581	0.2905	0.3579	0.6023	1.0000	-0.4222
-0.0577	-0.0849	-0.0498	-0.0508	-0.6598	-0.4222	1.0000
0.1763	0.2152	0.1679	0.1731	0.0692	0.2736	0.0016

Column 15

0.2171
0.2175
0.2175
0.2681
0.2297
0.2001
0.1979
0.1763
0.2152
0.1679
0.1731
0.0692

0.2736
0.0016
1.0000

fig =

Figure (18) with properties:

Number: 18
Name: ''
Color: [0.9400 0.9400 0.9400]
Position: [440 378 560 420]
Units: 'pixels'

Use GET to show all properties

fig =

Figure (18) with properties:

Number: 18
Name: ''
Color: [0.9400 0.9400 0.9400]
Position: [100 100 950 950]
Units: 'pixels'

Use GET to show all properties

*Training models on training data...
Unregularised linear regression*

b_lr =

1.0e+05 *

0.4586
0.0090
2.0790
-1.9885
0.0183
-0.1330
0.0154
-0.0128
-0.0002
0.0132
-0.0010
0.0213
-0.0021
0.0451
-0.0702
0.0100

res_norm_lr =

2.7947e+05

lm =

Linear regression model:

y ~ [Linear formula with 16 terms in 15 predictors]

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	45861	111.56	411.1	
0				
x1	901.8	4234.2	0.21298	
0.83136				
x2	2.079e+05	1.8455e+05	1.1265	
0.26006				
x3	-1.9885e+05	1.843e+05	-1.079	
0.2807				
x4	1830.8	512.32	3.5736	
0.00035874				
x5	-13298	1300.4	-10.226	
4.4842e-24				
x6	1543.7	640.75	2.4093	
0.016057				
x7	-1282.6	872.04	-1.4708	
0.14148				
x8	-24.74	224.66	-0.11012	
0.91232				
x9	1316.1	776.62	1.6947	
0.090262				
x10	-97.687	223.4	-0.43727	
0.66196				
x11	2130.9	157.4	13.538	
2.3832e-40				
x12	-208.26	204.16	-1.0201	
0.30779				
x13	4514	182.36	24.753	
3.7e-121				
x14	-7024.7	156.61	-44.854	
8.8932e-323				
x15	1001.9	120.17	8.3375	
1.2366e-16				

Number of observations: 2514, Error degrees of freedom: 2498

Root Mean Squared Error: 5.59e+03

R-squared: 0.783, Adjusted R-Squared 0.782
F-statistic vs. constant model: 600, p-value = 0
R Squared

rsq_lr =

0.7828

Ridge regression
Ridge regression

k_choices =

Columns 1 through 7

0.0010 0.0030 0.0100 0.0300 0.1000 0.3000 1.0000

Columns 8 through 10

3.0000 10.0000 30.0000

bnorms =

1.0e+04 *

Columns 1 through 7

8.9069 5.9995 5.0447 4.8979 5.1718 4.9221 4.8470

Columns 8 through 10

4.8225 4.7219 4.6878

Principle component analysis

Coefficient matrix

Columns 1 through 7

0.3190	-0.0474	-0.0497	-0.0199	-0.0291	0.0359	-0.1789
0.3188	-0.0484	-0.0515	-0.0245	-0.0305	0.0365	-0.1820
0.3188	-0.0484	-0.0515	-0.0245	-0.0305	0.0365	-0.1820
0.3022	0.0093	0.0020	-0.0678	-0.0025	-0.0614	-0.3415
0.3197	-0.0364	-0.0388	-0.0298	0.0008	-0.0041	-0.1594
0.3139	-0.0314	-0.0665	-0.0230	0.0551	-0.0478	0.2009
0.3144	-0.0433	-0.0612	0.0025	-0.0104	-0.0424	-0.1189
0.2704	-0.0476	-0.1214	-0.1423	-0.1173	0.6996	0.5371
0.3178	-0.0228	-0.0518	-0.0174	0.0139	-0.0350	-0.0511
0.2769	-0.0391	-0.0584	0.0133	0.1173	-0.6726	0.5765
0.1646	-0.0341	0.4715	0.7463	-0.1611	0.0626	-0.0428
0.0357	0.6320	-0.1850	-0.1830	0.2080	-0.0152	-0.2031
0.1298	0.4976	0.2940	0.1844	0.5773	0.1637	0.1573
-0.0344	-0.5720	0.1910	-0.1620	0.7046	0.0949	-0.1128
0.0832	0.0959	0.7631	-0.5665	-0.2595	-0.0721	0.0355

Columns 8 through 14

-0.0294	-0.1049	-0.2575	-0.0752	-0.2679	-0.1683	0.8216
-0.0304	-0.1173	-0.2644	-0.0963	-0.2678	-0.1713	-0.4023
-0.0304	-0.1173	-0.2644	-0.0964	-0.2682	-0.1717	-0.4033
-0.1065	0.7964	0.1209	0.0245	0.2738	-0.2184	-0.0075
-0.0326	0.0072	-0.0876	-0.3272	0.1565	0.8534	0.0044
-0.0385	-0.1978	0.6641	-0.5590	0.0195	-0.2351	0.0081
-0.0401	-0.4692	-0.0631	0.3301	0.7218	-0.1545	0.0002
0.1957	0.1843	-0.1046	0.0934	0.0778	0.0067	-0.0032
-0.0747	-0.0482	0.4860	0.6574	-0.3968	0.2397	-0.0139
0.1631	0.1564	-0.2513	0.0722	-0.0138	0.0061	-0.0035
0.3967	0.0289	0.0495	-0.0053	0.0041	-0.0059	-0.0040
0.6651	-0.0413	0.0515	-0.0039	-0.0094	-0.0082	0.0004
-0.4683	-0.0048	-0.1037	0.0130	-0.0029	-0.0009	-0.0017
0.3002	0.0004	0.0141	0.0179	0.0023	-0.0104	-0.0003
0.0460	-0.0673	0.0110	0.0009	-0.0067	-0.0006	0.0016

Column 15

-0.0005
0.7075
-0.7067
-0.0000
-0.0002
0.0001
-0.0002
-0.0000
0.0000
0.0000
-0.0000
0.0000
0.0000
0.0000
0.0000
0.0000

Singular values

Columns 1 through 7

173.8101	0	0	0	0	0	0
0	80.3406	0	0	0	0	0
0	0	57.5112	0	0	0	0
0	0	0	51.0844	0	0	0
0	0	0	0	38.5882	0	0
0	0	0	0	0	33.7811	0
0	0	0	0	0	0	26.4359
0	0	0	0	0	0	0
0	0	0	0	0	0	0
0	0	0	0	0	0	0
0	0	0	0	0	0	0
0	0	0	0	0	0	0
0	0	0	0	0	0	0
0	0	0	0	0	0	0
0	0	0	0	0	0	0
0	0	0	0	0	0	0

0	0	0	0	0	0	0
0	0	0	0	0	0	0
0	0	0	0	0	0	0
0	0	0	0	0	0	0
0	0	0	0	0	0	0

Columns 8 through 14

0	0	0	0	0	0	0
0	0	0	0	0	0	0
0	0	0	0	0	0	0
0	0	0	0	0	0	0
0	0	0	0	0	0	0
0	0	0	0	0	0	0
0	0	0	0	0	0	0
0	0	0	0	0	0	0
22.8197	0	0	0	0	0	0
0	20.7514	0	0	0	0	0
0	0	12.4585	0	0	0	0
0	0	0	8.0815	0	0	0
0	0	0	0	5.4425	0	0
0	0	0	0	0	4.0326	0
0	0	0	0	0	0	1.1972
0	0	0	0	0	0	0
0	0	0	0	0	0	0
0	0	0	0	0	0	0
0	0	0	0	0	0	0
0	0	0	0	0	0	0
0	0	0	0	0	0	0
0	0	0	0	0	0	0

Column 15

0
0
0
0
0
0
0
0
0
0
0
0
0
0
0
0
0.0226
0
0
0
0
0

Principal component regression: 12 components

b_pcr1 =

1.0e+04 *
4.5874
0.1212
0.6171
-0.1701
-0.3004
0.3264
-0.0374
-0.1393
-0.3840
-0.0574
0.0302
0.3126
0.4888

res_norm_pcr =

2.6932e+05

lm =

Linear regression model:

$y \sim 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10 + x11 + x12$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	45874	119.41	384.18	0
x1	1212.5	35.609	34.05	2.5066e-205
x2	6170.9	83.158	74.206	0
x3	-1701.3	114.88	-14.809	2.0397e-47
x4	-3004.1	128.51	-23.376	2.2448e-108
x5	3263.9	172.9	18.877	6.7069e-74
x6	-374.49	179.24	-2.0894	0.036786
x7	-1393.4	234.61	-5.939	3.3135e-09
x8	-3840.1	292.66	-13.121	5.9185e-38
x9	-574	295.58	-1.9419	0.052268
x10	302.47	496.47	0.60925	0.54242
x11	3126.1	796.8	3.9233	8.9972e-05
x12	4887.7	1204.5	4.058	5.119e-05

Number of observations: 2263, Error degrees of freedom: 2250

Root Mean Squared Error: 5.68e+03

R-squared: 0.777, Adjusted R-Squared 0.776

F-statistic vs. constant model: 654, p-value = 0

R squared

rsq =

0.7606

Principal component regression: 11 components

b_pcr2 =

1.0e+04 *

4.5867
0.1205
0.6182
-0.1743
-0.3054
0.3233
-0.0422
-0.1497
-0.3698
-0.0441
0.0135
0.3107

res_norm_pcr =

2.7033e+05

lm =

Linear regression model:

$y \sim 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10 + x11$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	45867	119.85	382.68	0
x1	1204.8	35.298	34.132	3.8174e-206
x2	6182.2	82.721	74.735	0
x3	-1743.4	115.17	-15.138	2.2492e-49
x4	-3054.1	128.77	-23.717	3.504e-111
x5	3233.5	174.72	18.507	2.7134e-71
x6	-421.65	177.13	-2.3804	0.017375
x7	-1497.3	239.36	-6.2554	4.7321e-10
x8	-3698.5	291.41	-12.692	1.0678e-35
x9	-441.15	300.59	-1.4676	0.14235
x10	135.15	510.56	0.2647	0.79126
x11	3106.8	767.93	4.0457	5.3922e-05

Number of observations: 2263, Error degrees of freedom: 2251
 Root Mean Squared Error: 5.7e+03
 R-squared: 0.78, Adjusted R-Squared 0.779
 F-statistic vs. constant model: 725, p-value = 0
 R squared

rsq =

0.6958

Principal component regression: 10 components

b_pcr3 =

1.0e+04 *
 4.5899
 0.1197
 0.6168
 -0.1606
 -0.2990
 0.3227
 -0.0365
 -0.1524
 -0.3637
 -0.0393
 0.0201

res_norm_pcr =

2.7304e+05

lm =

Linear regression model:

$y \sim 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	45899	120.97	379.41	0
x1	1197.2	35.495	33.728	3.2161e-202
x2	6168.2	84.552	72.952	0
x3	-1605.9	117.46	-13.673	6.0848e-41
x4	-2990.3	131.81	-22.686	8.7556e-103
x5	3227	176.89	18.243	1.8636e-69
x6	-364.66	178.92	-2.0381	0.041653
x7	-1524.4	233.69	-6.5229	8.4886e-11
x8	-3637	300.45	-12.105	1.0033e-32

x9	-392.57	292.52	-1.3421	0.17971
x10	200.91	495.27	0.40566	0.68503

Number of observations: 2263, Error degrees of freedom: 2252
 Root Mean Squared Error: 5.75e+03
 R-squared: 0.767, Adjusted R-Squared 0.766
 F-statistic vs. constant model: 743, p-value = 0
 R squared

rsq =

0.8104

Principal component regression: 9 components

b_pcr4 =

1.0e+04 *
 4.5854
 0.1209
 0.6178
 -0.1768
 -0.3062
 0.3121
 -0.0400
 -0.1404
 -0.3807
 -0.0472

res_norm_pcr =

2.7281e+05

lm =

Linear regression model:

$y \sim 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	45854	120.84	379.45	0
x1	1209.3	35.886	33.699	6.0378e-202
x2	6178.2	83.327	74.144	0
x3	-1767.8	116.27	-15.205	8.88e-50
x4	-3061.5	130.24	-23.507	1.8366e-109
x5	3121.4	175.52	17.784	2.5959e-66
x6	-400.07	179.16	-2.233	0.025648

x7	-1403.6	238.53	-5.8842	4.5986e-09
x8	-3806.9	293.02	-12.992	2.8724e-37
x9	-472.22	295.85	-1.5962	0.11059

Number of observations: 2263, Error degrees of freedom: 2253
 Root Mean Squared Error: 5.75e+03
 R-squared: 0.774, Adjusted R-Squared 0.773
 F-statistic vs. constant model: 859, p-value = 0
 R squared

rsq =

0.7481

Principal component regression: 8 components

b_pcr5 =

1.0e+04 *
 4.5858
 0.1192
 0.6169
 -0.1636
 -0.2992
 0.3245
 -0.0307
 -0.1473
 -0.3849

res_norm_pcr =

2.7086e+05

lm =

Linear regression model:

$y \sim 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	45858	119.98	382.2	0
x1	1191.7	37.116	32.108	1.3444e-186
x2	6169.2	83.604	73.79	0
x3	-1635.8	115.26	-14.192	7.5255e-44
x4	-2992	132.48	-22.585	5.4975e-102
x5	3245.5	175.88	18.452	6.4351e-71
x6	-306.9	195.52	-1.5696	0.11664

x7	-1472.9	257.8	-5.7132	1.2552e-08
x8	-3848.9	299.41	-12.855	1.4955e-36

Number of observations: 2263, Error degrees of freedom: 2254
 Root Mean Squared Error: 5.71e+03
 R-squared: 0.771, Adjusted R-Squared 0.77
 F-statistic vs. constant model: 950, p-value = 0
 R squared

rsq =

0.7770

Principal component regression: 7 components

b_pcr6 =

1.0e+04 *

4.5847
 0.1203
 0.6202
 -0.1739
 -0.2940
 0.3123
 -0.0363
 -0.1300

res_norm_pcr =

2.7977e+05

lm =

Linear regression model:

$y \sim 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	45847	123.88	370.08	0
x1	1202.9	36.786	32.699	2.721e-192
x2	6201.5	85.981	72.126	0
x3	-1739.3	118.63	-14.662	1.4614e-46
x4	-2940.2	133.39	-22.041	1.1586e-97
x5	3122.6	180.06	17.342	2.423e-63
x6	-363.39	183.08	-1.9849	0.047276
x7	-1300.2	236.9	-5.4883	4.5122e-08

Number of observations: 2263, Error degrees of freedom: 2255
 Root Mean Squared Error: 5.89e+03
 R-squared: 0.76, Adjusted R-Squared 0.759
 F-statistic vs. constant model: 1.02e+03, p-value = 0
 R squared

rsq =

0.7188

Principal component regression: 6 components

b_pcr7 =

1.0e+04 *
 4.5757
 0.1186
 0.6160
 -0.1637
 -0.2914
 0.3299
 -0.0492

res_norm_pcr =

2.7882e+05

lm =

Linear regression model:

$y \sim 1 + x1 + x2 + x3 + x4 + x5 + x6$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	45757	123.42	370.75	0
x1	1185.9	37.072	31.99	1.7217e-185
x2	6160.4	86.117	71.535	0
x3	-1636.9	118.82	-13.776	1.6289e-41
x4	-2913.6	132.28	-22.026	1.5086e-97
x5	3298.6	181.09	18.215	2.8219e-69
x6	-492.45	200.53	-2.4557	0.014134

Number of observations: 2263, Error degrees of freedom: 2256
 Root Mean Squared Error: 5.87e+03
 R-squared: 0.753, Adjusted R-Squared 0.752
 F-statistic vs. constant model: 1.14e+03, p-value = 0

R squared

rsq =

0.7481

Principal component regression: 5 components

b_pcr8 =

1.0e+04 *

4.5860
0.1190
0.6143
-0.1710
-0.2960
0.3027

res_norm_pcr =

2.8438e+05

lm =

Linear regression model:

$y \sim 1 + x1 + x2 + x3 + x4 + x5$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	45860	125.86	364.37	0
x1	1190.1	36.642	32.479	3.3541e-190
x2	6142.8	86.752	70.808	0
x3	-1710.2	120.57	-14.184	8.2896e-44
x4	-2960	134.48	-22.011	1.9853e-97
x5	3027.2	181.28	16.699	4.0753e-59

Number of observations: 2263, Error degrees of freedom: 2257

Root Mean Squared Error: 5.99e+03

R-squared: 0.751, Adjusted R-Squared 0.751

F-statistic vs. constant model: 1.36e+03, p-value = 0

R squared

rsq =

0.7577

Principal component regression: 4 components

b_pcr9 =

1.0e+04 *
4.5817
0.1221
0.6152
-0.1682
-0.2948

res_norm_pcr =

3.0249e+05

lm =

Linear regression model:

$y \sim 1 + x1 + x2 + x3 + x4$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	45817	133.86	342.29	0
x1	1221	41.467	29.444	1.5124e-161
x2	6152.1	93.133	66.057	0
x3	-1681.9	129.81	-12.956	4.3989e-37
x4	-2948.3	143.3	-20.575	2.3611e-86

Number of observations: 2263, Error degrees of freedom: 2258

Root Mean Squared Error: 6.37e+03

R-squared: 0.719, Adjusted R-Squared 0.719

F-statistic vs. constant model: 1.45e+03, p-value = 0

R squared

rsq =

0.7164

Principal component regression: 3 components

b_pcr10 =

1.0e+04 *
4.5812
0.1239
0.6186
-0.1738

res_norm_pcr =

3.3220e+05

lm =

Linear regression model:

$y \sim 1 + x1 + x2 + x3$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	45812	146.96	311.74	0
x1	1239.4	52.38	23.662	9.1615e-111
x2	6185.6	103	60.056	0
x3	-1738.4	144.98	-11.991	3.688e-32

Number of observations: 2263, Error degrees of freedom: 2259

Root Mean Squared Error: 6.99e+03

R-squared: 0.667, Adjusted R-Squared 0.667

F-statistic vs. constant model: 1.51e+03, p-value = 0

R squared

rsq =

0.6435

numcomps =

12 11 10 9 8 7 6 5 4 3

rsqAll =

0.7606	0.7772
0.6958	0.7800
0.8104	0.7673
0.7481	0.7742
0.7770	0.7712
0.7188	0.7597
0.7481	0.7528
0.7577	0.7514
0.7164	0.7192
0.6435	0.6672

b vector: linear regression

1.0e+05 *

0.4586
0.0090
2.0790
-1.9885
0.0183
-0.1330
0.0154
-0.0128
-0.0002
0.0132
-0.0010
0.0213
-0.0021
0.0451
-0.0702
0.0100

rsq_lr =

0.7828

Ridge regression: test rsq

RR_rsqa_test =

Columns 1 through 7

0.7924 0.7936 0.8109 0.7468 0.6489 0.7623 0.7335

Columns 8 through 10

0.7950 0.7811 0.7544

Ridge regression: training rsq

RR_rsqa_train =

Columns 1 through 7

0.7817 0.7803 0.7776 0.7861 0.7892 0.7843 0.7872

Columns 8 through 10

0.7798 0.7798 0.7797

Ridge regression: b vector

rr_opt_bvec =

1.0e+04 *

4.5826

0.2622
0.3188
0.3143
0.1315
-1.0756
-0.0044
-0.1610
0.0172
0.2015
-0.0210
0.2073
-0.0202
0.4502
-0.6987
0.1070

rr_opt_rsqr =

0.7950

Ridge regression: all b vectors

RR_all_bvec =

1.0e+04 *

Columns 1 through 7

4.5911	4.5891	4.5852	4.5830	4.5894	4.5888	4.5841
0.0075	0.2818	0.1538	0.0369	-0.3314	-0.0780	0.0832
5.7169	2.7447	1.2900	0.6970	0.9632	0.5243	0.3861
-4.7995	-2.0013	-0.2785	0.1972	0.7696	0.4727	0.3728
0.1959	0.2425	0.1677	0.1768	0.1842	0.2006	0.1808
-1.2990	-1.5563	-1.3345	-1.2724	-1.7932	-1.3436	-1.1597
0.1776	0.1702	-0.0010	0.1460	0.1689	0.1670	0.1257
-0.0976	-0.1052	-0.2279	-0.1275	0.0237	-0.0959	-0.0979
-0.0003	0.0139	0.0072	0.0029	0.0248	0.0144	0.0123
0.1227	0.1926	0.2313	0.1467	0.0464	0.1387	0.1000
-0.0312	-0.0008	-0.0244	-0.0124	0.0065	-0.0053	-0.0055
0.2142	0.2106	0.2080	0.2235	0.1900	0.2091	0.2061
-0.0216	-0.0214	-0.0199	-0.0226	-0.0425	-0.0208	-0.0299
0.4464	0.4564	0.4504	0.4544	0.4675	0.4478	0.4610
-0.7056	-0.7021	-0.7000	-0.6953	-0.7128	-0.7136	-0.7112
0.0988	0.1078	0.0913	0.0988	0.0965	0.1011	0.0962

Columns 8 through 10

4.5826	4.5794	4.5886
0.2622	0.1498	0.0830
0.3188	0.1428	0.0829
0.3143	0.1412	0.0826
0.1315	0.0939	0.0660
-1.0756	-0.6999	-0.3881

-0.0044	0.0074	-0.0020
-0.1610	-0.0212	-0.0629
0.0172	0.0155	0.0273
0.2015	0.1760	0.1129
-0.0210	-0.0104	-0.0101
0.2073	0.2253	0.2122
-0.0202	-0.0125	-0.0138
0.4502	0.4391	0.4399
-0.6987	-0.6937	-0.6916
0.1070	0.0881	0.0994

Principal component regression: test rsq
Columns 1 through 7

0.7606	0.6958	0.8104	0.7481	0.7770	0.7188	0.7481
--------	--------	--------	--------	--------	--------	--------

Columns 8 through 10

0.7577	0.7164	0.6435
--------	--------	--------

Principal component regression: training rsq
Columns 1 through 7

0.7772	0.7800	0.7673	0.7742	0.7712	0.7597	0.7528
--------	--------	--------	--------	--------	--------	--------

Columns 8 through 10

0.7514	0.7192	0.6672
--------	--------	--------

Principal component regression: b vector
1.0e+04 *

4.5860
0.1190
0.6143
-0.1710
-0.2960
0.3027

pcr_opt_rsqa =

0.7577

Principal component regression: all b vectors

b_pcr1 =

1.0e+04 *
4.5874
0.1212
0.6171
-0.1701

-0.3004
0.3264
-0.0374
-0.1393
-0.3840
-0.0574
0.0302
0.3126
0.4888

b_pcr2 =

1.0e+04 *

4.5867
0.1205
0.6182
-0.1743
-0.3054
0.3233
-0.0422
-0.1497
-0.3698
-0.0441
0.0135
0.3107

b_pcr3 =

1.0e+04 *

4.5899
0.1197
0.6168
-0.1606
-0.2990
0.3227
-0.0365
-0.1524
-0.3637
-0.0393
0.0201

b_pcr4 =

1.0e+04 *

4.5854
0.1209
0.6178
-0.1768

-0.3062
0.3121
-0.0400
-0.1404
-0.3807
-0.0472

b_pcr5 =

1.0e+04 *

4.5858
0.1192
0.6169
-0.1636
-0.2992
0.3245
-0.0307
-0.1473
-0.3849

b_pcr6 =

1.0e+04 *

4.5847
0.1203
0.6202
-0.1739
-0.2940
0.3123
-0.0363
-0.1300

b_pcr7 =

1.0e+04 *

4.5757
0.1186
0.6160
-0.1637
-0.2914
0.3299
-0.0492

b_pcr8 =

1.0e+04 *

4.5860
0.1190
0.6143
-0.1710
-0.2960
0.3027

b_pcr9 =

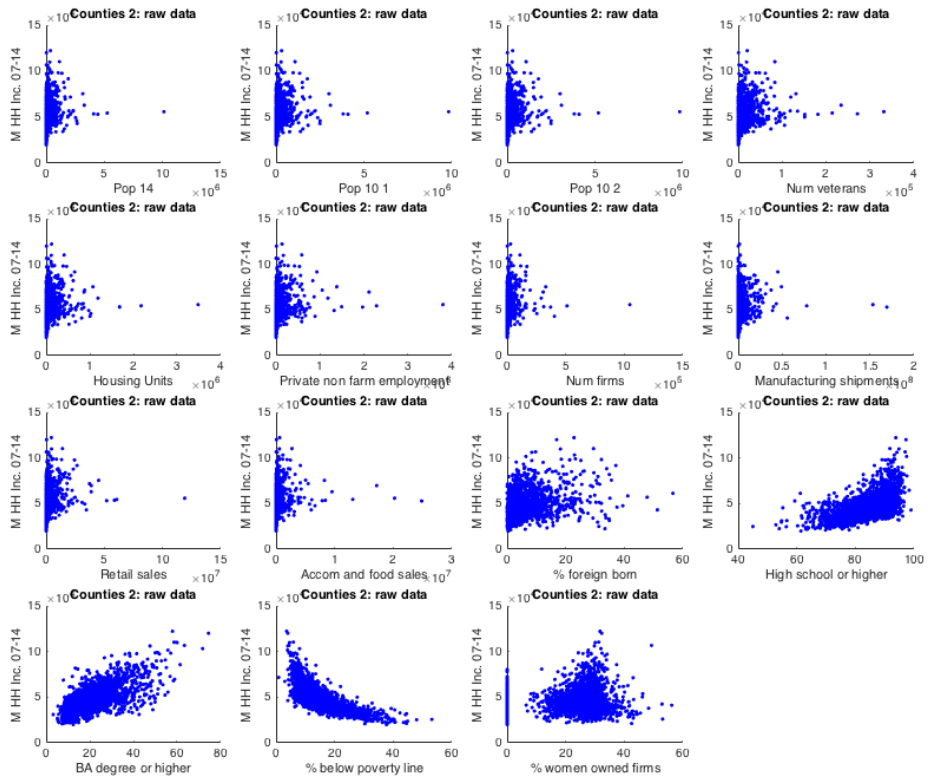
1.0e+04 *

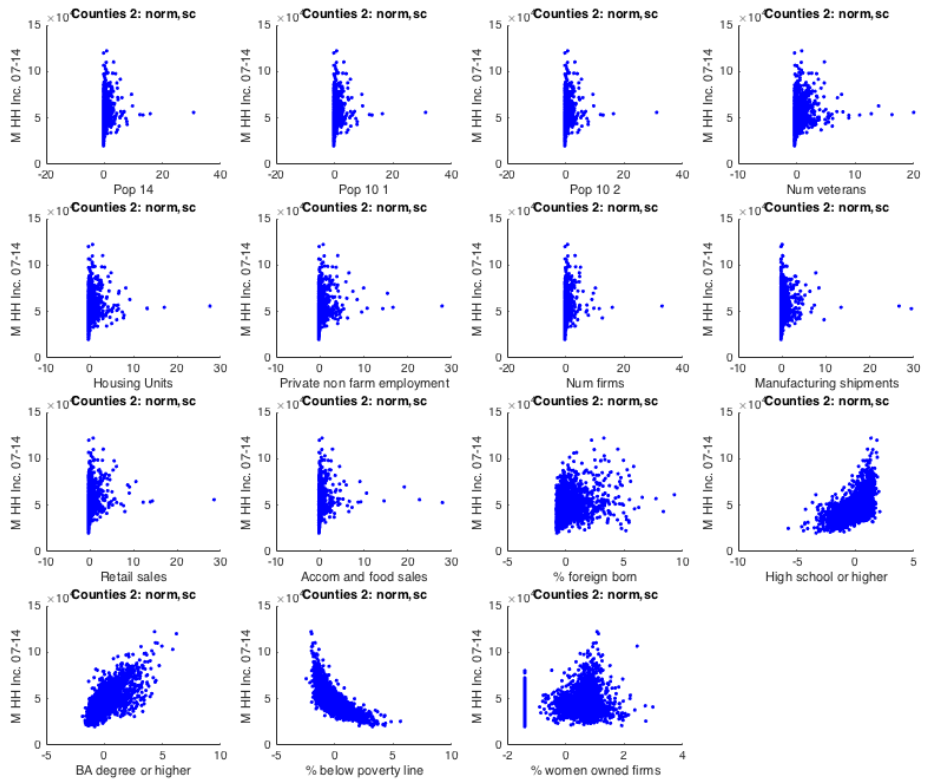
4.5817
0.1221
0.6152
-0.1682
-0.2948

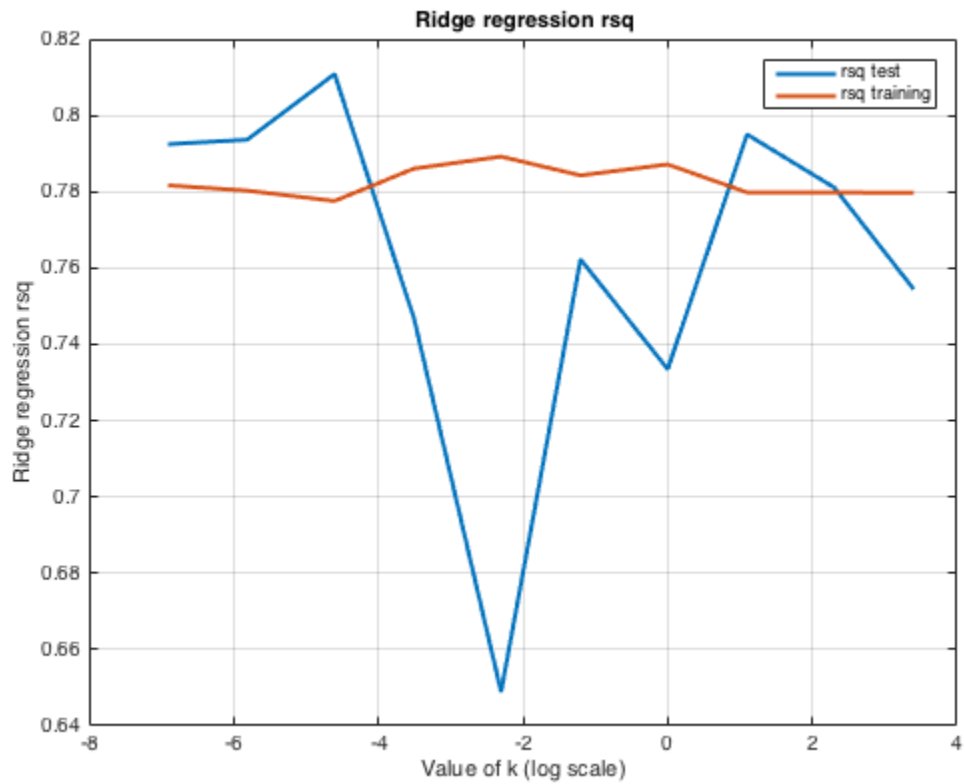
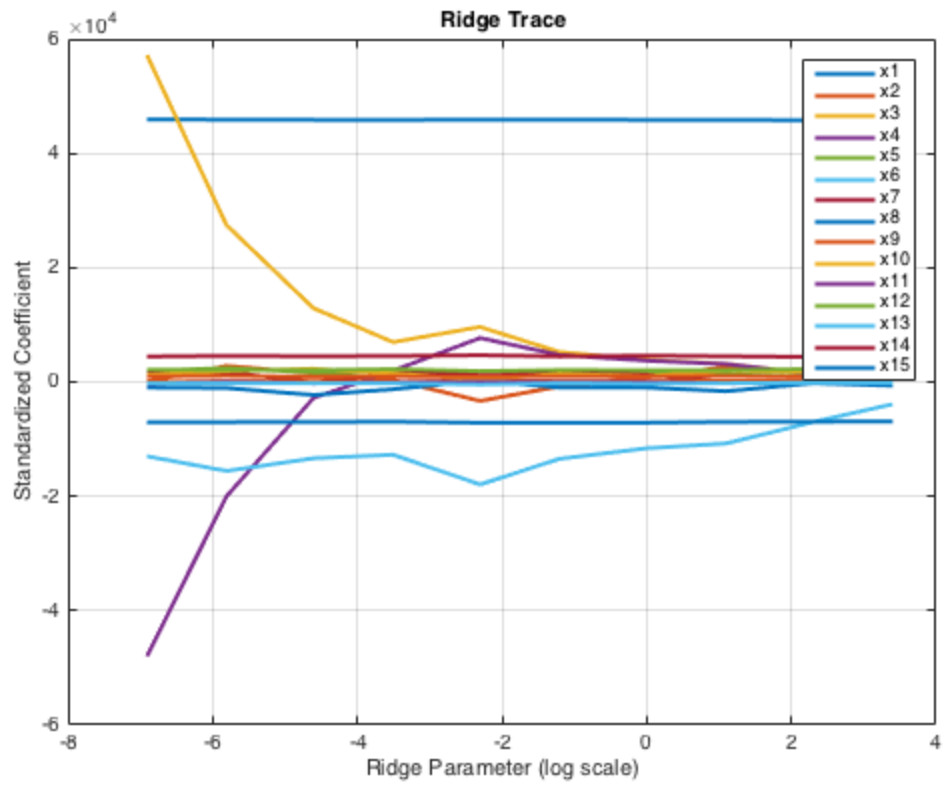
b_pcr10 =

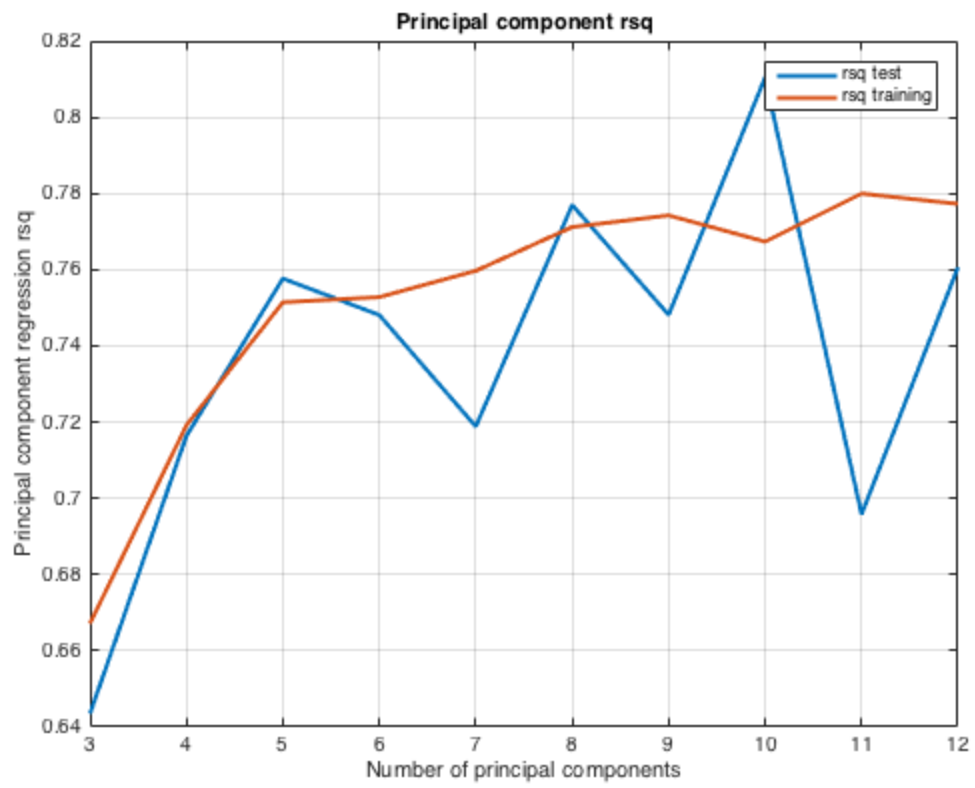
1.0e+04 *

4.5812
0.1239
0.6186
-0.1738









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