```
%% HW3 Q1
load fiberpaper.dat
Y = fiberpaper(:, 1 : 4);
X = fiberpaper(:, [7 5 6]);
alpha = 0.01;
X1 = X(1, :)'; X4 = X(4, :)';
X = X(setdiff(1:size(X,1),1),:); Y = Y(setdiff(1:size(Y,1),1),:);
u1 = lrt env(X 1, Y 1, alpha);
ModelOutput1 = env(X 1, Y 1, u1);
plenv = predict env(ModelOutput1, X1, 'prediction');
plpenv = predict env2(X 1, Y 1, X1, 'prediction');
[plenv.value plenv.SE plpenv.value plpenv.SE plenv.SE./plpenv.SE]
X = X(setdiff(1:size(X,1),4),:); Y = Y(setdiff(1:size(Y,1),4),:);
u4 = lrt env(X 4, Y 4, alpha);
ModelOutput4 = env(X 4, Y 4, u4);
p4env = predict env(ModelOutput4, X4, 'prediction');
p4penv = predict env2(X 4, Y 4, X4, 'prediction');
[p4env.value p4env.SE p4penv.value p4penv.SE p4env.SE./p4penv.SE]
% OUTPUT
응
% u =
응
응
       2
응
용
% ans =
응
     21.0006
              2.6388
                        21.0352
                                   2.6254
                                             1.0051
                                            1.0282
용
     7.0854
              0.7162 7.0385
                                 0.6966
     5.3011
              1.2500
                        5.2725
                                  1.2356
                                            1.0116
응
용
      0.8613
               0.5833
                         0.8633
                                   0.5711
                                             1.0212
응
용
% u =
용
용
       2
용
용
% ans =
응
응
    21.8770
               2.5521
                        21.8708
                                   2.5608
                                             0.9966
응
     7.3094
              0.6952
                         7.3202
                                   0.6792
                                             1.0236
용
     5.7082
              1.2082
                         5.7123
                                  1.2017
                                             1.0054
응
     1.0508
               0.5639
                         1.0571
                                   0.5584
                                             1.0099
```

Problem 3.4.

When *S* is the parameter of interest:

- 1. u is selected to be 1, 1 and 0 by AIC, BIC and LRT at $\alpha = 0.01$, respectively. Hence we build our partial envelope model with u = 1.
- 2. The OLS and partial envelope estimator of the coefficient vector are close to each other (in the first output matrix, cols 1 and 2).
- 3. Only one of the envelope coefficient/SE ratios is > 2 (column 4), hence significant at 95% level.
- 4. Partial envelope gives gains in SE for all components of the coefficient matrix, although they are at max about 2-fold (col 5).
- 5. In the $\hat{\Gamma}$ matrix (which is a vector here, col 6), all the elements are small in absolute value compared to the 4th one. This means that **variation in Ozone levels is a big part of the material variation in solar radiation**, and variations in other variables are mostly immaterial.

```
%% HW3 Q2
load Ozone.txt
Y = Ozone(:, 3:7); X = Ozone(:, 1:2);
ols = fit OLS(X,Y); olsSE = sqrt(diaq(ols.SigmaOLS,0)/ols.n);
alpha = 0.01;
%% parameter of interest is solar radiation
XS.X1 = X(:,2); XS.X2 = X(:,1);
[modelselectaic(XS, Y, 'penv') modelselectbic(XS, Y, 'penv')
modelselectlrt(XS, Y, alpha, 'penv')] %% OUTPUT: 1 1 0
uS = 1;
penvS = penv(XS, Y, uS);
[ols.betaOLS(:,2) penvS.beta1 penvS.asySE ...
sqrt(ols.n)*penvS.betal./penvS.asySE penvS.ratio penvS.Gamma]
% coefficient estiamtes by OLS and coeff estimates, asymptotic SE, ratio
statistic, SE
% ratio and Gamma for partial envelope
     0.0117 0.0097 0.0384
                                1.6379
                                           1.8033 -0.0988
    -0.0064 -0.0033
                       0.0236
                                -0.9146
                                           2.5548 0.0340
응
    0.0205 0.0153 0.1041 0.9540
응
                                          1.8561 -0.1562
응
     0.0952 0.0963 0.2953 2.1135 1.0067 -0.9817
     0.0027 0.0031 0.0167 1.1986 2.3644 -0.0315
```

When W is the parameter of interest:

- 1. Here also u is selected to be 1, 1 and 0 by AIC, BIC and LRT at $\alpha = 0.01$, respectively. Hence we take u = 1 for our penv model.
- 2. The OLS and partial estimator of the coefficient vector are very different, especially for NO₂ and Ozone levels (in the first output matrix, cols 1 and 2).
- 3. All the envelope coefficient/SE ratios are > 2 (col 4) in absolute value.
- 4. Partial envelope gives gains in SE over OLS for all components of the coefficient matrix, and the gains are massive for, yet again, NO₂ and Ozone levels (col 5).
- 5. In the $\hat{\Gamma}$ matrix (which is a vector here, col 6), elements corresponding to these two variables are smaller in absolute value than other elements.
- 6. Compared to OLS the much smaller coefficients and standard errors, and also the corresponding elements in $\hat{\Gamma}$ for NO₂ and Ozone means that according to the partial envelope analysis, variations in these two variables are mostly immaterial in wind speed variations.

```
%% parameter of interest is wind speed
XW.X1 = X(:,1); XW.X2 = X(:,2);
[modelselectaic(XW, Y, 'penv') modelselectbic(XW, Y, 'penv')
modelselectlrt(XW, Y, alpha, 'penv')] %% OUTPUT: 1 1 0
uW = 1;
penvW = penv(XW, Y, uW);
[ols.betaOLS(:,1) penvW.beta1 penvW.asySE ...
sqrt(ols.n)*penvW.beta1./penvW.asySE penvW.ratio penvW.Gamma]
% coefficient estiamtes by OLS and coeff estimates, asymptotic SE, ratio
statistic, SE
% ratio and Gamma for partial envelope
    -0.1382 0.0710 0.1674 2.7481
                                        4.5328
                                                0.4504
    -0.1925 -0.0749 0.1687 -2.8771
                                        3.9233 -0.4754
용
   -0.2113 -0.0166 0.0410 -2.6197 51.6941 -0.1051
용
   -0.7868 -0.0106 0.0256 -2.6786 127.1239 -0.0673
```