P1.

(A)

Let $\max(2x_1 + 3x_2 - 5x_3, x_1, x_2, 2) \le t$. We can then form the following linear program:

$$\begin{array}{lllll} \min & t \\ \text{s.t.} & +2x_1 & +3x_2 & -5x_3 & \leq t \\ & & +2x_1 & -x_2 & +x_3 & \leq t \\ & & x_1, & x_2 & \leq t \end{array}$$

(B)

Let $t_1, t_2, t_3, t_4 \ge 0$.

Let $|x_1 + x_2| \le t_1$, $|x_2 - x_3| \le t_2$, $|x_3 - x_1| \le t_3$, $|x_1 + x_2 + x_3| \le t_4$.

You now have the linear problem

(C)

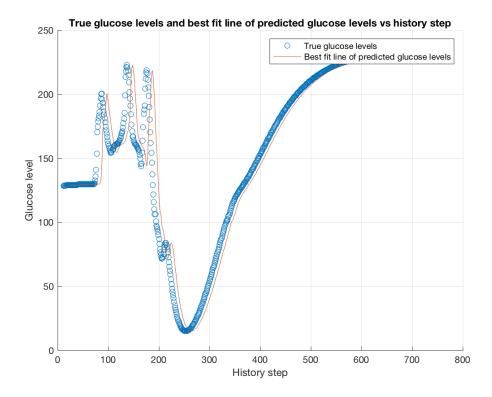
Let $\max(|x_1|, |x_2|, |x_3|, |x_1 + x_2|) \le t$.

You now have the linear problem

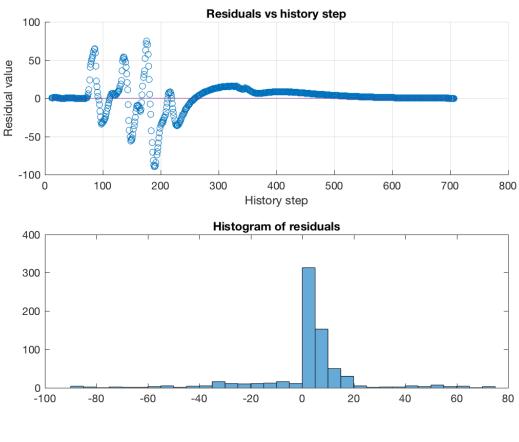
```
1 % Read in the csv file skipping the first row
2 M = csvread('insulinGlucose1.csv', 1);
   % Insulin input values
4
   u = M(:,1);
   % Glucose levels
  G = M(:,2);
9
   A = [];
10
   b = [];
11
   [m, \tilde{}] = size(G(12:end));
12
13
   for i = 1:m
        b = [b, G(i)];
14
        b = [b, -G(i)];
15
   end
16
17
18 b = b;
19
20 % Initialize the objective function which is the minimum of the sum
21 % of the absolute values of the residuals.
22 % Residuals have the form y - Ax.
23 f = ones(1,17);
25 \% Create the constraint matrix A which is the residuals
26 % of glucose levels 12 \rightarrow 707.
27 % If u = |x| then u = x and u = -x
28
   [m, \tilde{}] = size(G);
29
   for t = 11:(m-1)
       a = [G(t) \ G(t-1) \ G(t-2) \ G(t-3) \ G(t-4) \ G(t-5) \ G(t-6) \ G(t-7) \ G(t-8) \ G(t-9) \ G(t-10)];
30
        a = horzcat(a, [u(t) u(t-1) u(t-2) u(t-3) u(t-4) u(t-5)]);
32
       % Append the constraint ax.
33
       A = [A; a];
34
       \% Append the constraint -ax.
35
        a = times(a, -1);
       A = [A; a];
37
38
   end
39
   % Solve for the coefficents.
40
41 % x will have an array of the coefficient values when it is solved.
[x, fval] = linprog(f, A, b)
43
44 \% Recreate the matrix A without the additional absolute value constraints.
45 A = [];
46 for t = 11:(m-1)
        a = [G(t) \ G(t-1) \ G(t-2) \ G(t-3) \ G(t-4) \ G(t-5) \ G(t-6) \ G(t-7) \ G(t-8) \ G(t-9) \ G(t-10)];
47
        a = horzcat(a, [u(t) u(t-1) u(t-2) u(t-3) u(t-4) u(t-5)]);
48
       A \, = \, [\, A\, ; \, a\, ]\, ;
49
50
51
   % Create an array of the history steps.
52
53
   t = [];
   for i = 12:m
54
55
        t = [t, i];
56
57
   % Plot the true glucose levels at each history step t_i.
   figure (1);
59
   scatter (t, G(12: end));
61 hold on;
   % Plot the best fit line of the predicted glucose levels at each
63
   % history step t_i.
64
65~\% x stores the estimated coefficients of the best fit line.
66 \% |x| = sqrt(x^2)
67
   A = (A*x).^2;
68
  A = sqrt(A);
   plot(t, A);
69
   grid on;
   legend('True glucose levels', 'Best fit line of predicted glucose levels');
71
   title ('True glucose levels and best fit line of predicted glucose levels vs history step');
   xlabel('History step');
```

```
74 ylabel('Glucose level');
  saveas(1, 'predictions.png');
75
76
77 % Plot the residuals.
  figure(2);
78
  residuals = [];
  trueGlucoseValues = G(12:end);
80
   predictedGlucoseValues = A;
81
   [m, ~] = size (predicted Glucose Values);
82
   for i = 1:m
83
       residuals = [residuals, trueGlucoseValues(i) - predictedGlucoseValues(i)];
84
85
   end
86
   subplot(2,1,1);
87
   scatter(t, residuals);
88
89 hold on;
  plot(t, zeros(m));
90
   grid on;
  title ('Residuals vs history step');
93 xlabel('History step');
94 ylabel ('Residual value');
95
   subplot(2,1,2);
96
  histogram (residuals);
97
   title ('Histogram of residuals');
99 saveas(2, 'residuals.png');
```

Coefficient	Value
a_0	8.704148513061227e-14
a_1	-3.375077994860476e-14
a_2	-1.474376176702208e-13
a_3	2.131628207280301e-13
a_4	-1.243449787580175e-13
a_5	-1.056932319443149e-13
a_6	1.376676550535194e-13
a_7	-3.979039320256561e-13
a_8	3.179678742526448e-13
a_9	6.430411758628907e-13
a_{10}	0.9999999999309
b_0	1.989519660128281e-13
b_1	-2.060573933704291e-13
b_2	1.847411112976261e-13
b_3	-8.526512829121202e-14
b_4	-1.207922650792170e-13
b_5	7.815970093361102e-14



(a) Linear Regression Predictions



(b) Residuals

P3.

(A)

1.

2.

3.

(B)

1.

- 2. x_6 will be the entering variable because it has the largest coefficient.
- 3. The equation for the entering variable column a_i is $a_i = -B^{-1}Ne_i$.
- 4. The leaving variable with be w_6 because it constrains the entering variable the most.
- 5. The basic variables will be $x_3, x_4, x_5, w_1, w_2, x_6$.

(C)

(D) No solution.

(E) No solution.

```
2 \quad A = \begin{bmatrix} +1 & -1 & 0 & 0 & 0 & -1 & +1 & 0 & 0 & 0 & 0 \\ \end{bmatrix}; \dots
           +1 \ 0 \ 0 \ -1 \ -1 \ 0 \ 0 \ +1 \ 0 \ 0 \ 0 \ 0; \dots
3
           0\ 0\ -1\ 0\ 0\ -1\ 0\ 0\ +1\ 0\ 0\ 0;\dots
4
           0 \ -1 \ -1 \ +1 \ 0 \ +1 \ 0 \ 0 \ 0 \ +1 \ 0 \ 0; \dots
           +1 \ 0 \ +1 \ 0 \ +1 \ +1 \ 0 \ 0 \ 0 \ +1 \ 0; \dots
          -1 \ 0 \ 0 \ -1 \ +1 \ -1 \ 0 \ 0 \ 0 \ 0 \ +1;
7
    b = [3 -1 -2 +4 +6 -2];
9
10
    c = \begin{bmatrix} -2 & -3 & -1 & -1 & 0 & +1 & 0 & 0 & 0 & 0 & 0 \end{bmatrix};
11
12
13 % P3 a1
14 	ext{ xB} = [1 	ext{ 2} 	ext{ 3} 	ext{ 7} 	ext{ 8} 	ext{ 9}];
15 \text{ xN} = \begin{bmatrix} 4 & 5 & 6 & 10 & 11 & 12 \end{bmatrix};
16 B = A(:, xB);
17 \ N \, = \, A\,(\,:\,,\ xN\,)\;;
    Binv = inv(B);
    \% Dictionary constant value column.
19
20 Binv*b;
21 % Dictionary coefficient value rows.
-Binv*N;
23
cB = c(:, xB);
cN = c(:, xN);
26 % Objective function constant value.
    cB*(Binv*b);
    % Object function coefficient values.
29 cN - cB*(Binv*N);
31 % P3 a2
32 \text{ xB} = [5 \ 6 \ 7 \ 8 \ 9 \ 10];
33
    xN = [1 \ 2 \ 3 \ 4 \ 11 \ 12];
34
35
   B = A(:, xB);
36
    N = A(:, xN);
37
   Binv = inv(B);
39 % Dictionary constant value column.
40 Binv*b;
41 \% Dictionary coefficient value rows.
42
    -\operatorname{Binv}*N;
43
dA cB = c(:, xB);
45 \text{ cN} = c(:, xN);
    % Objective function constant value.
46
47
    cB*(Binv*b);
    % Object function coefficient values.
49
    cN - cB*(Binv*N);
50
   % P3 a3
51
    xB = [1 \ 2 \ 6 \ 10 \ 11 \ 12];
52
53 \text{ xN} = [3 \ 4 \ 5 \ 7 \ 8 \ 9];
54
55 B = A(:, xB);
56
    N \,=\, A\,(\,:\,,\ xN\,)\;;
    Binv = inv(B);
    % Dictionary constant value column.
60~\% Dictionary coefficient value rows.
    -Binv*N;
61
62
63 cB = c(:, xB);
   cN = c(:, xN);
65 % Objective function constant value.
    cB*(Binv*b);
    % Object function coefficient values.
67
    cN - cB*(Binv*N);
68
69
70 % p3 b
71 xB = [3 \ 4 \ 5 \ 7 \ 8 \ 12];
72 \text{ xN} = \begin{bmatrix} 1 & 2 & 6 & 9 & 10 & 11 \end{bmatrix};
73
74 B = A(:, xB);
75 N = A(:, xN);
```

```
76 Binv = inv(B);
77 % Dictionary constant value column.
    Binv*b;
78
79 % Dictionary coefficient value rows.
80 - Binv*N;
81
cB = c(:, xB);
    cN = c(:, xN);
83
84~\% Objective function constant value.
85 cB*(Binv*b);
 86 % Object function coefficient values.
cN - cB*(Binv*N);
89 % Choose the third variable as the entering variable.
90 % Get the column of the entering variable.
91 a_i = -Binv*N(:, 3);
92
93 % Determine the leaving variable row index by finding the value that
94~\% constraints the entering variable the most.
    -(Binv*b)./a_i;
96
97 % Get the leaving variable row.
    a_{-j} = -(Binv*N);
98
    a_{-j} = a_{-j} (6,:);
99
100
101 % Р3 с
    f = [-2 \ -3 \ -1 \ -1 \ 0 \ 1];
102
    Y = \begin{bmatrix} 1 & -1 & 0 & 0 & 0 & -1; 1 & 0 & 0 & -1 & -1 & 0; 0 & 0 & -1 & 0 & 0 & -1; 0 & -1 & -1 & 1 & 0 & 1; 1 & 0 & 1 & 0; 1; -1 & 0 & 0 & -1 & 1 & -1 \end{bmatrix};
104 \quad x = \begin{bmatrix} 3 & -1 & -2 & 4 & 6 & -2 \end{bmatrix};
    options = optimoptions('linprog', 'Algorithm', 'dual-simplex');
[x,fval] = linprog(-f, Y, x, [], [], zeros(size(f)), [], options)
106
107
    xB = [5 \ 6 \ 7 \ 9 \ 11 \ 12];
108
   xN = \begin{bmatrix} 1 & 2 & 3 & 4 & 8 & 10 \end{bmatrix};
109
110
111 B = A(:, xB);
112 N = A(:, xN);
113 Binv = inv(B);
114 % Dictionary constant value column.
115 Binv*b
116 \% Dictionary coefficient value rows.
117
    -\operatorname{Binv}*N
118
119 	ext{ cB} = c(:, xB);
120 \text{ cN} = c(:, xN);
121 % Objective function constant value.
122 cB*(Binv*b)
123 % Object function coefficient values.
```

124 cN - cB*(Binv*N)