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:

(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

```
options = optimoptions('linprog', 'Algorithm', 'dual-simplex');
     % Read in the csv file skipping the first row M = csvread('insulinGlucosel.csv', 1);
     % Insulin input values
     u = M(:,1);
% Glucose levels
10
     G = M(:,2);
11
\frac{13}{14}
     b = [];
       \begin{array}{lll} [m,\tilde{\ }] &=& \mathtt{size} \, (G(\,1\,2\,:\, \mathtt{end}\,)\,)\,; \\ & \hspace{-0.5em} \text{for} & \hspace{-0.5em} i \,=\, 1\,:m \end{array} 
16
        b = [b, G(i)];
18
19
            b = [b, -G(i)];
20
21
     b = b';
     \% Initialize the objective function which is the minimum of the sum \% of the absolute values of the residuals.  

% Residuals have the form y - Ax.
\frac{23}{24}
26
27
     f = ones(1,17);
     \% Create the constraint matrix A which is the residuals
     % Create the constraint matrix A which is the residuals
% of glucose levels 12 -> 707.
% If u = |x| then u = x and u = -x
[m,~] = size(G);
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)];
a = horzcat(a, [u(t) u(t-1) u(t-2) u(t-3) u(t-4) u(t-5)]);
% Append the constraint ax
29
30
32
33
34
\frac{35}{36}
           % Append the constraint ax. A = [A; a];
37
           % Append the constraint -ax.

a = times(a, -1);
\frac{38}{39}
40
            A = [A; a];
     end
41
    \% Solve for the coefficents. \% x will have an array of the coefficient values when it is solved.
43
44
     x = linprog(f, A, b);
46
     \% Recreate the matrix A without the additional absolute value constraints.
47
     A = [];
for t = 11:(m-1)
49
         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)];
50
51
             a = horz(at)(a, [u(t) u(t-1) u(t-2) u(t-3) u(t-4) u(t-5)]);
52
             A = [A; a];
     end
54
     \% Create an array of the history steps.
\frac{55}{56}
      t = [];
     for i = 12:m
57
\frac{58}{59}
            t = [t, i];
     end
60
61
     % Plot the true glucose levels at each history step t_i.
     figure(1);
     scatter(t,G(12:end));
hold on;
\frac{63}{64}
65
     \% Plot the best fit line of the predicted glucose levels at each
66
     % history step t_i. % x stores the estimated coefficients of the best fit line. % |x| = \sup_{x \in \mathbb{R}^n} (x^2)
68
69
70
71
      A = (A*x).^2;
     A = (A*x).^2;

A = sqrt(A);

plot(t, A);

grid on;

legend ('True glucose levels', 'Best fit line of predicted glucose levels');

title ('True glucose levels and best fit line of predicted glucose levels vs history step');

xlabel ('History step');

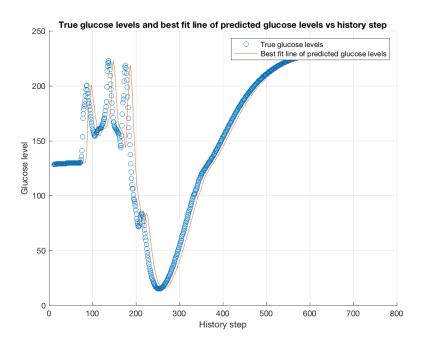
ylabel ('Glucose level');

saveas(1. 'predictions.png');
74
\begin{array}{c} 77 \\ 78 \end{array}
      saveas(1, 'predictions.png');
79
    % Plot the residuals.
80
     figure(2);
      residuals = [];
82
83
      trueGlucoseValues = G(12:end);
      predicted GlucoseValues = A;

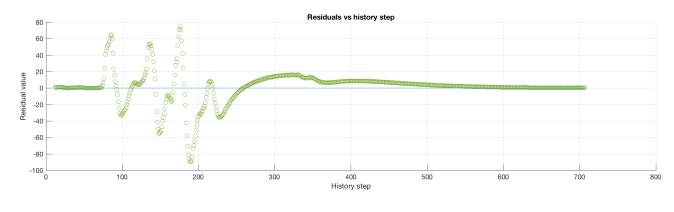
[m, "] = size(predicted GlucoseValues);

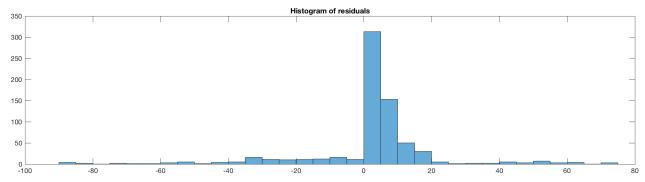
for i = 1:m
86
            residuals = [residuals , trueGlucoseValues(i) - predictedGlucoseValues(i)];
88
89
90
     subplot (2,1,1);
      scatter(t, residuals);
91
     hold on;
plot(t, zeros(m));
93
      grid on;
     title('Residuals vs history step');
xlabel('History step');
ylabel('Residual value');
96
97
     subplot(2,1,2);
99
```

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. $\{x_1, x_2, x_3, w_1, w_2, w_3\}$.

2. $\{x_1, x_2, x_5, w_3, w_5, w_6\}$.

3. $\{x_1, x_2, x_6, w_4, w_5, w_6\}$.

(B) Perform one step of the revised simplex method for the dictionary with the basis:

$$\{x_3, x_4, x_5, w_1, w_2, w_6\}$$

1. Compute the constant column and objective rows of this dictionary.

2. Choose the entering variable with the largest coefficient.

 x_6 will be the entering variable because it has the largest coefficient.

3. Set up the equations to construct the column for the entering variable.

The equation for the entering variable column a_i is $a_i = -B^{-1}Ne_i$.

4. Choose the leaving variable.

The leaving variable with be w_6 because it constrains the entering variable the most.

5. Write down the basic variables in the next dictionary.

The basic variables will be $x_3, x_4, x_5, w_1, w_2, x_6$.

(C)

Problem is unbounded.

(D)

No solution.

 (\mathbf{E})

No solution.

```
options = optimoptions('linprog', 'Algorithm', 'dual-simplex');
                  A = [+1 \ -1 \ 0 \ 0 \ 0 \ -1 \ +1 \ 0 \ 0 \ 0 \ 0 \ ] \dots
                                          b = [3 -1 -2 +4 +6 -2];
 10
                  c = \begin{bmatrix} -2 & -3 & -1 & -1 & 0 & +1 & 0 & 0 & 0 & 0 & 0 \end{bmatrix};
               % P3 a1
                 \begin{array}{l} \text{70 Fo al} \\ \text{xB} = [1 \ 2 \ 3 \ 7 \ 8 \ 9]; \\ \text{xN} = [4 \ 5 \ 6 \ 10 \ 11 \ 12]; \\ \text{B} = \text{A}(:, \ \text{xB}); \\ \text{N} = \text{A}(:, \ \text{xN}); \\ \text{Binv} = \text{inv}(\text{B}); \\ \end{array} 
20
21
22
23
24
25
                   % Dictionary constant value column.
                  Binv*b;
% Dictionary coefficient value rows.
                  -\mathrm{Binv}*\mathrm{N};
                cB = c(:, xB);
cN = c(:, xN);
% Objective function constant value.
                 CB*(Binv*b);
% Object function coefficient values.
cN - cB*(Binv*N);
29
30
31
32
33
34
35
36
37
38
39
              % P3 a2

xB = [1 \ 2 \ 5 \ 9 \ 11 \ 12];

xN = [3 \ 4 \ 6 \ 7 \ 8 \ 10];
                \begin{split} B &= A(:, xB); \\ N &= A(:, xN); \\ Binv &= inv(B); \\ \% \ Dictionary \ constant \ value \ column. \end{split}
41
42
43
44
45
46
47
48
49
50
51
                % Dictionary coefficient value rows.
-Binv*N;
               cB = c(:, xB);
cN = c(:, xN);
% Objective function constant value.
cB*(Binv*b);
% Object function coefficient values.
cN - cB*(Binv*N);
                 xB = \begin{bmatrix} 1 & 2 & 6 & 10 & 11 & 12 \end{bmatrix};

xN = \begin{bmatrix} 3 & 4 & 5 & 7 & 8 & 9 \end{bmatrix};
52
53
54
                \begin{split} B &= A(:, xB); \\ N &= A(:, xN); \\ Binv &= inv(B); \\ \% & Dictionary constant value column. \\ Bruch &= A(xB) + A(xB)
                  Binv*b;
                 % Dictionary coefficient value rows.
-Binv*N;
                 cN = c(:, xN);

% Objective function constant value.
                 cB*(Binv*b);
% Object function coefficient values.
                  cN - cB*(Binv*N);
70 % p3 b
71 xB = [3 4 5 7 8 12];
```

```
72  xN = [1 2 6 9 10 11];
73
74  B = A(:, xB);
75  N = A(:, xN);
76  Binv = inv(B);
77  % Dictionary constant
78  Binv*b;
79  % Dictionary coefficie
80  -Binv*N;
81
82  cB = c(:, xB);
83  cN = c(:, xN);
84  % Objective function coefficie
85  cB*(Binv*b);
86  % Object function coefficie
87  cN - cB*(Binv*N);
88
98  Choose the third validation of the coefficient coe
                                   B = A(:, xB);
N = A(:, xN);
Binv = inv(B);
% Dictionary constant value column.
                                          Binv*b;
% Dictionary coefficient value rows.
                                       \begin{array}{lll} cB = c\left(:, & xB\right); \\ cN = c\left(:, & xN\right); \\ \% & \text{Objective function constant value.} \\ cB*(Binv*b); \\ \% & \text{Object function coefficient values.} \\ cN - cB*(Binv*N); \end{array}
   88 % Choose the third variable as the entering variable.
90 % Get the column of the entering variable.
91 a_i = -Binv*N(:, 3);
92
                                       \% Determine the leaving variable row index by finding the value that \% constraints the entering variable the most.
        93
94
        95
                                              -(Binv*b)./a_i;
        \frac{96}{97}
                                          % Get the leaving variable row.
                                          a_{-j} = -(Binv*N);

a_{-j} = a_{-j}(6,:);
       98
99
   100
x = 100 \text{ m} x = 100 \text{ m
```