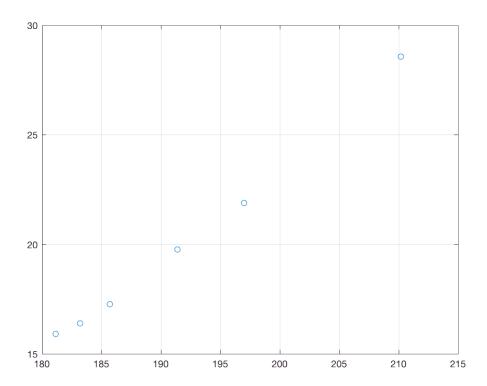
## ASE-4046 Exercise 3 Solutions

## **Problem 1**

One should always look at the data before fitting a curve!

```
x=[210.2 \quad 197 \quad 191.4 \quad 185.7 \quad 183.2 \quad 181.15]'; % column vector v=[28.559 \quad 21.892 \quad 19.758 \quad 17.267 \quad 16.385 \quad 15.919]'; plot(x,v,'o'); grid on
```



The points appear to lie on a slightly curved line, so an exponential fit seems reasonable.

An exponential model can be fitted by fitting a straight line  $y \approx b_1 x + b_2$  to the log-transformed data. The formulas on slide 6 give the coefficients

```
y=log(v);
n=length(y); Sx=sum(x); Sx2=sum(x.^2); Sy=sum(y); Sxy=sum(x.*y);
b=[n*Sxy-Sx*Sy; Sx2*Sy-Sx*Sxy ]/(n*Sx2-Sx^2)

b =
    0.0204
    -0.9309
```

The same answer is obtained by solving the normal equation as on slide 15:

```
X=[x ones(n,1)];
bb=X\y
```

```
0.0204
```

The fitted exponential model is  $v \approx e^{b_1 x + b_2} = e^{b_2} e^{b_1 x}$ , where

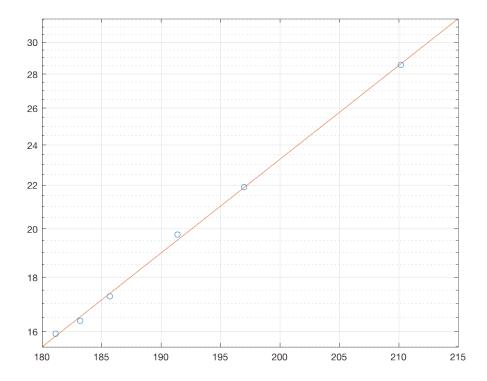
```
eb2=exp(b(2)), b1=b(1)

eb2 = 0.3942

b1 = 0.0204
```

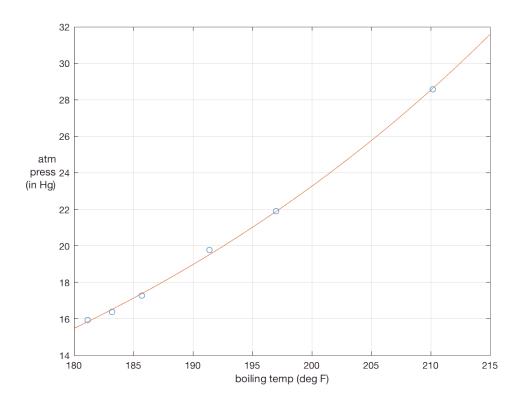
Plotting on semilog axes we see that the straight line is a good fit:

```
xx=180:215;
yy=b(1)*xx+b(2);
vv=exp(yy);
semilogy(x,v,'o',xx,vv); grid on
```



On linear axes we see that the fitted curve has some curvature:

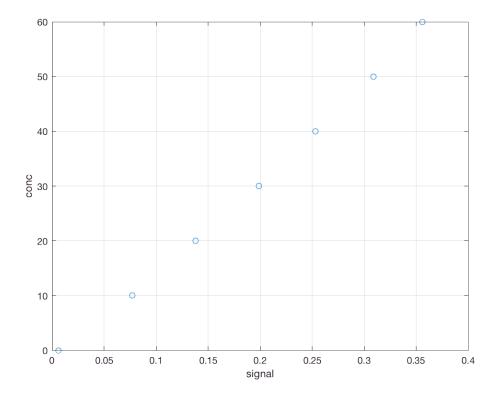
```
plot(x,v,'o',xx,vv);
grid on
xlabel('boiling temp (deg F)')
ylabel({'atm';'press';'(in Hg)'},'rot',0,'horiz','right')
```



## Problem 2

First, look at the data.

```
y=(0:10:60)'; % concentration, column vector
t=[0.006 0.077 0.138 0.199 0.253 0.309 0.356]'; % signal
plot(t,y,'o'); grid on; xlabel('signal'), ylabel('conc')
```



There appears to be a small amount of curvature. As instructed, we fit a degree 2 polynomial; this can be done by adapting the code on slide 16.

```
n=length(t); tt=(2*t-t(1)-t(n))/(t(n)-t(1)); \quad \text{$\%$ transform to $[-1,1]$} \\ d=2; \quad \text{$\%$ polynomial degree} \\ X=ones(n,d+1); \quad \text{for $k=d:-1:1, $X(:,k)=tt.*X(:,k+1); end $\%$ design matrix $bb=X\y; \quad \text{$\%$ solve LS problem}
```

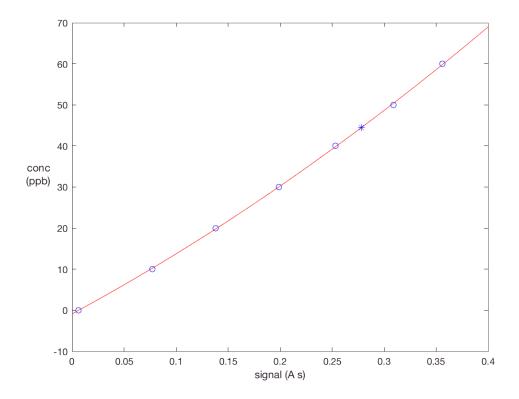
The model predicts the following lead concentration of tap water at "your home" for the given instrument reading:

```
stap=0.278; % instrument reading for your tap water sstap=(2*stap-t(1)-t(n))/(t(n)-t(1)); % transform to [-1,1] scale ctap=polyval(bb,sstap) % evaluate the polynomial
```

ctap = 44.4465

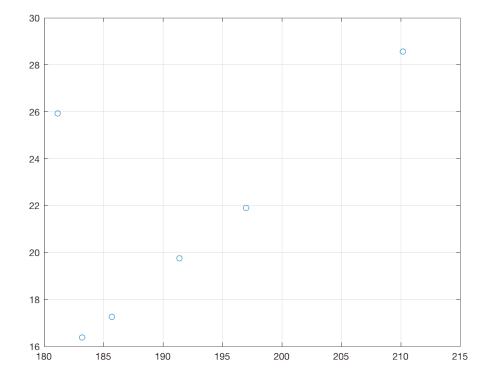
Plot the data and the curve. The lead concentration at "your home" is shown with a blue asterisk.

```
T=0:0.01:0.4; % mesh for plotting
TT=(2*T-t(1)-t(n))/(t(n)-t(1)); % transformed mesh
plot(T,polyval(bb,TT),'r-',t,y,'bo',stap,ctap,'b*')
xlabel('signal (A s)')
ylabel({'conc';'(ppb)'},'rot',0,'horiz','right')
```



## Problem 3

First, we look at the data. The outlier is clearly visible on the left.



Here we fit a least-squares model to the log-transformed data:

```
n=length(x);
y=log(v);
X=[x ones(n,1)];
bLS=X\y
bLS =

0.0119
0.7794
```

Next we fit the least-absolute residuals model to the log-transformed data, as in slide 21. For some versions of Matlab there is a warning message about the algorithm.

```
 f = [0 \ 0 \ ones(1,2*n)]'; \ % \ cost = f' * [b; u; v] = sum(u) + sum(v) \\ lb = [-inf -inf zeros(1,2*n)]; \ % \ constraints u >= 0, v >= 0 \\ Aeq = [X \ eye(n) \ -eye(n)]; \ beq = y; \ % \ constraint \ X*b + u - v = y \\ designvar = linprog(f,[],[],Aeq,beq,lb); \ % \ designvar = [b; u; v]
```

Warning: Your current settings  $\underline{\text{will run a different algorithm}}$  ('dual-simplex') in a future release.

Optimization terminated.

```
bLAD = designvar(1:2); % coefficients
```

The fitted exponential model is  $v \approx e^{b_1x+b_2} = e^{b_2}e^{b_1x}$ , where

```
eb2=exp(bLAD(2)), b1=bLAD(1)
```

```
eb2 = 0.4142

b1 = 0.0201
```

In the following plot we see that the LAD fit "ignores" the outlier and has a better fit to the remaining points.

```
xx=180:215; % dense mesh
yyLS=polyval(bLS,xx);
yyLAD=polyval(bLAD,xx);
plot(xx,exp(yyLAD),'-',xx,exp(yyLS),'--',x,v,'o')
legend('LAD','LS','location','southeast')
```

