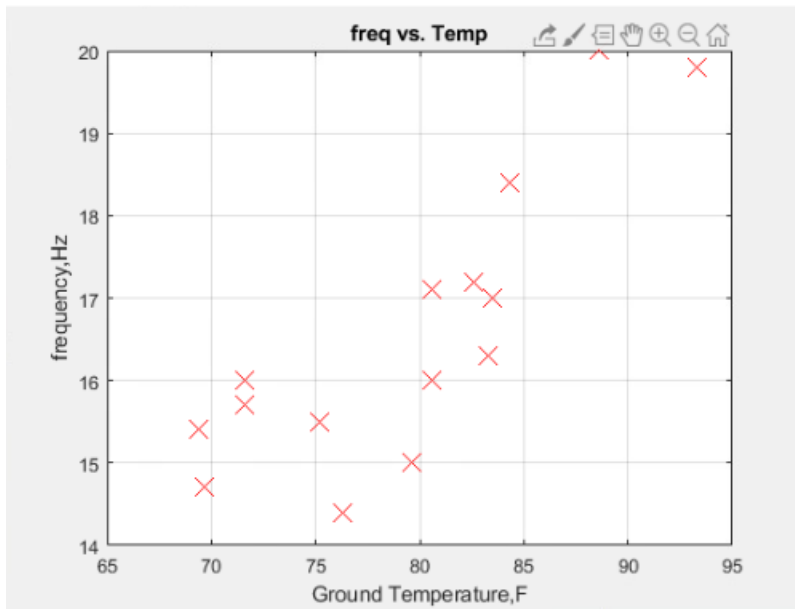
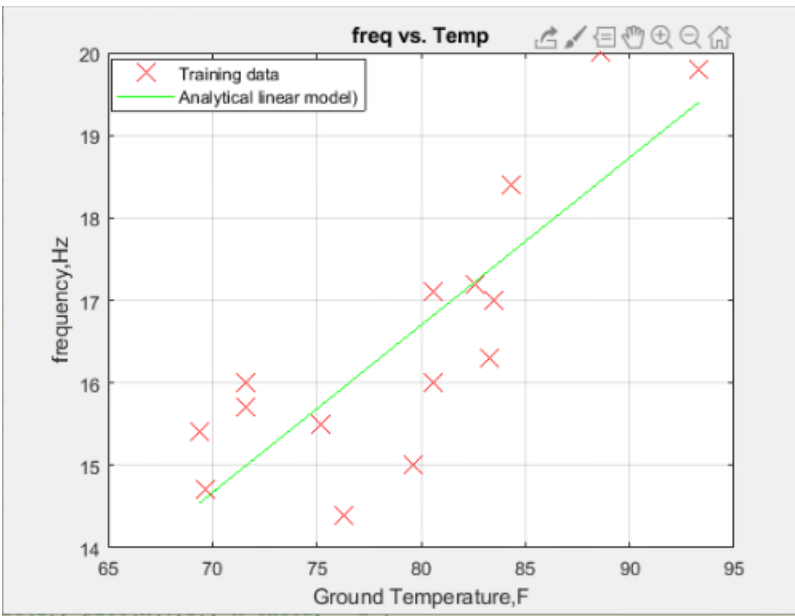


1)

1)

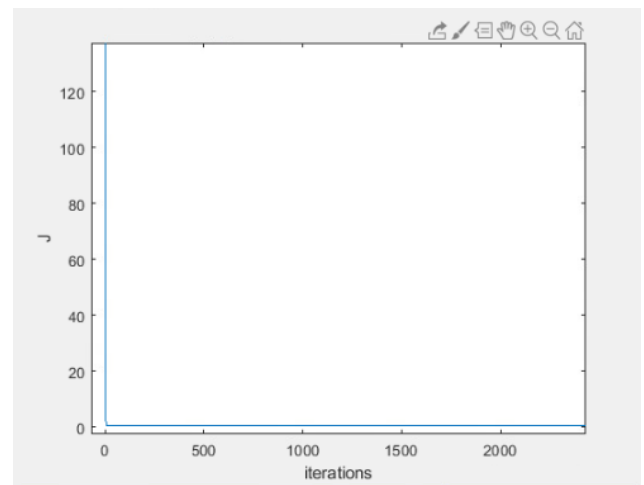
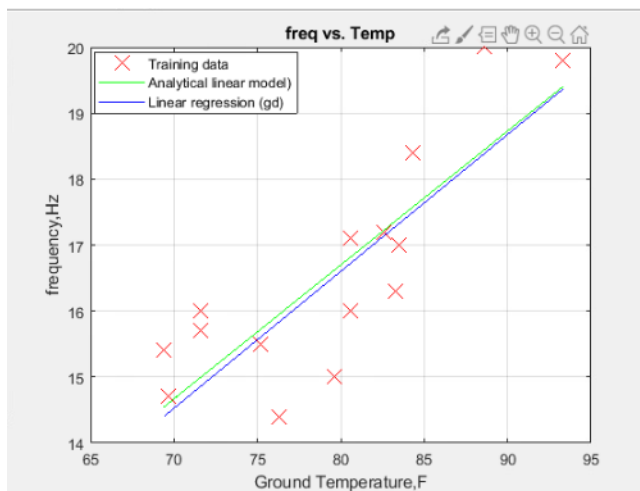


2)



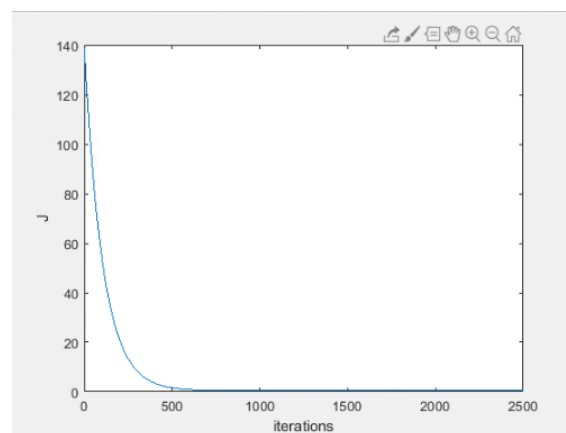
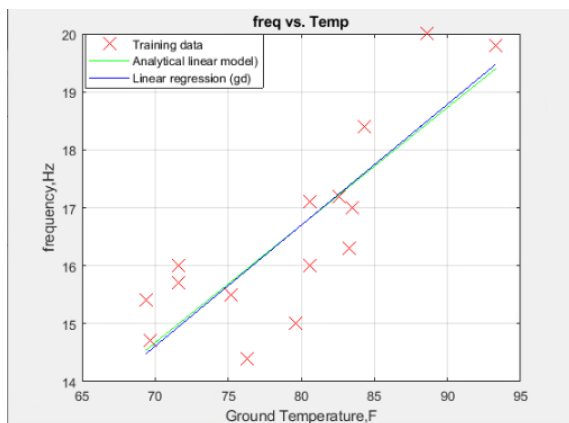
3) Comparison of the Theta found by linear regression (with  $\alpha = 5e-6$ , iterations=2500) to that found analytically:

Theta found Analytically: 0.459315 0.203000  
 Theta found by gradient descent: 0.002943 0.207485



Comparison of the Theta found by linear regression (with  $\alpha = 5e-8$ , iterations=2500) to that found analytically:

Theta found Analytically: 0.459315 0.203000  
 Theta found by gradient descent: 0.002617 0.208700

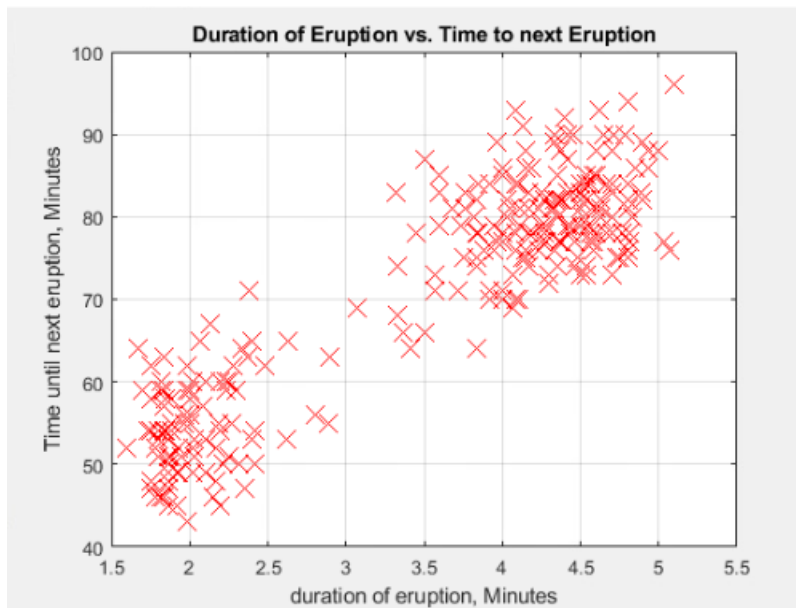


We can see that we did not need all 2500 iterations to reach an acceptable value for our cost function

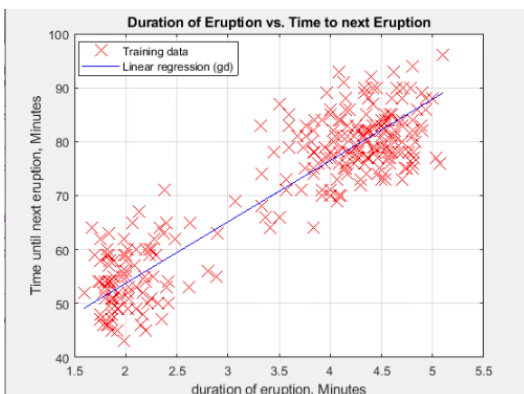
- 4) For temperature = 91 degrees, we predict a freq of 18.994332 Hz  
 For temperature = 77 degrees, we predict a freq of 16.072530 Hz
- 5) For temperature = 50 degrees, we predict a freq of 10.437625 Hz

2)

1)



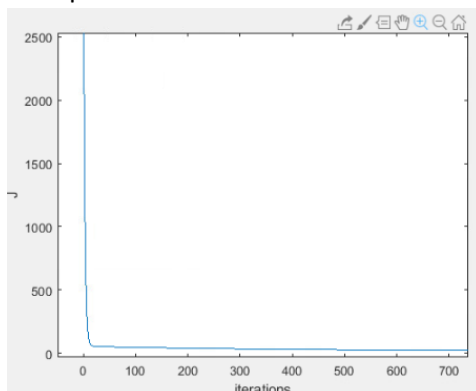
2)



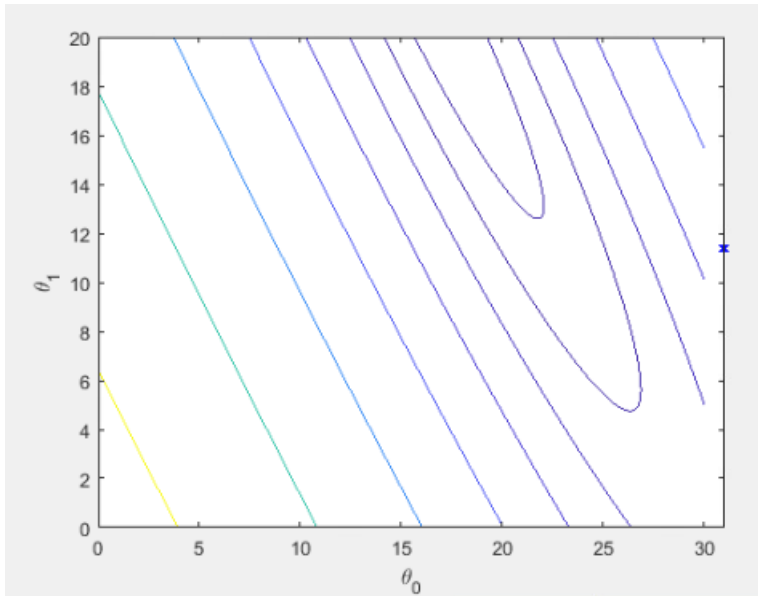
3)

For duration = 1.5 minutes, we predict the next eruption in 48.061149 minutes  
For duration = 3 minutes, we predict the next eruption in 65.118113 minutes  
For duration = 5 minutes, we predict the next eruption in 87.860732 minutes

4) see computeCost.m



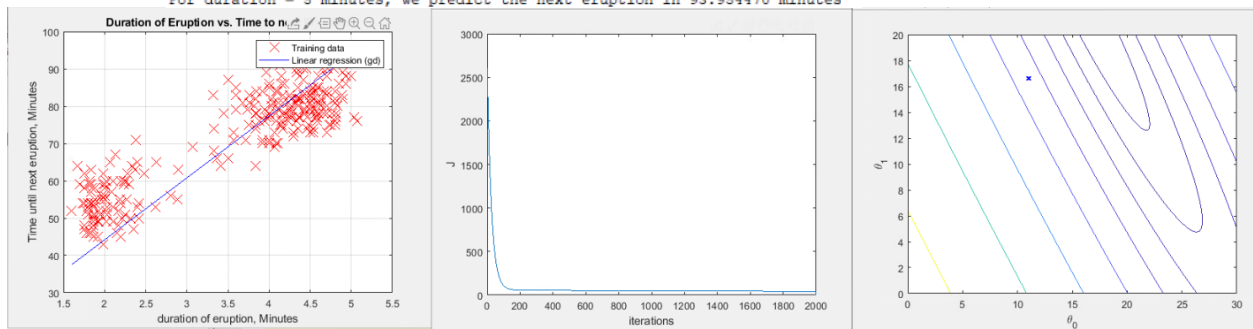
5)



6) previous graphs were using alpha = 5e-5

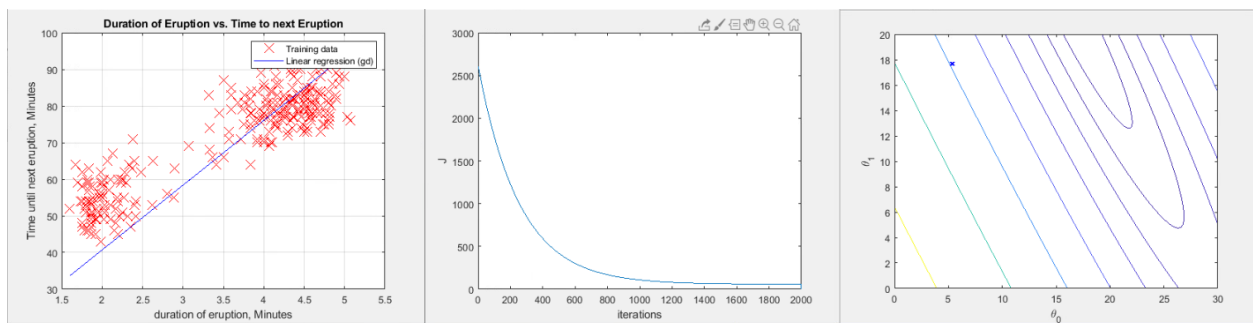
Theta found by gradient descent: 11.002468 16.590400  
 For duration = 1.5 minutes, we predict the next eruption in 35.888068 minutes  
 For duration = 3 minutes, we predict the next eruption in 60.773669 minutes  
 For duration = 5 minutes, we predict the next eruption in 93.954470 minutes

Alpha = 5e-6



Theta found by gradient descent: 5.352990 17.672474  
 For duration = 1.5 minutes, we predict the next eruption in 31.861700 minutes  
 For duration = 3 minutes, we predict the next eruption in 58.370411 minutes  
 For duration = 5 minutes, we predict the next eruption in 93.715358 minutes

Alpha=5e-7



3)

1) the dimensions of theta in this case are 3x1

Generally speaking, it is  $n+1$ , where  $n$  is the number of parameters

And we add 1 for  $\theta_0$

2) with  $\alpha = 0.001$ , we get our optimal theta after around 1250 iterations

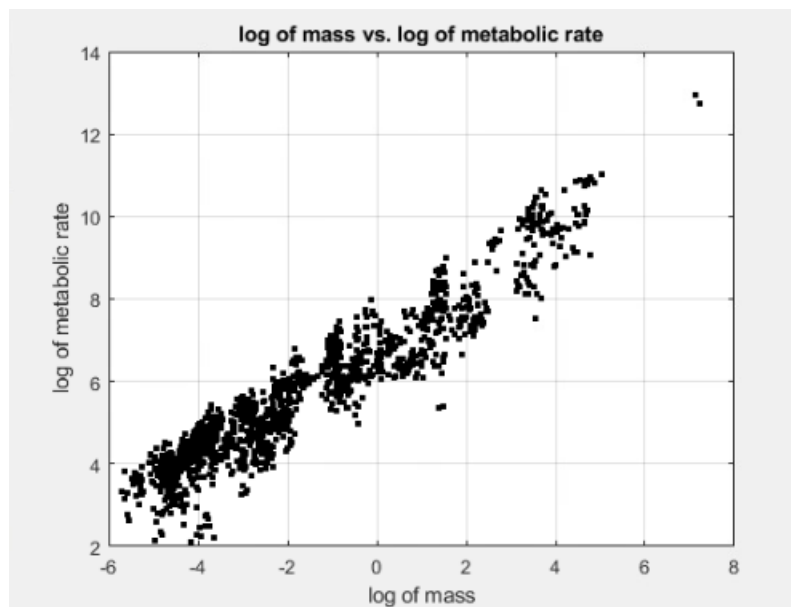
Which give us the following predictions: For House with 1800 sf and 5 rooms, the predicted cost is \$423134.13

3) comparison of analytical vs linear regression:

```
Theta found by gradient descent: 340314.303042 110344.568321
For House with 1800 sf and 5 rooms, the predicted cost is $423134.13
Theta Analytical linear model: 340412.659574 110631.050279
For House with 1800 sf and 5 rooms, the predicted cost is $423342.51
```

4)

1)



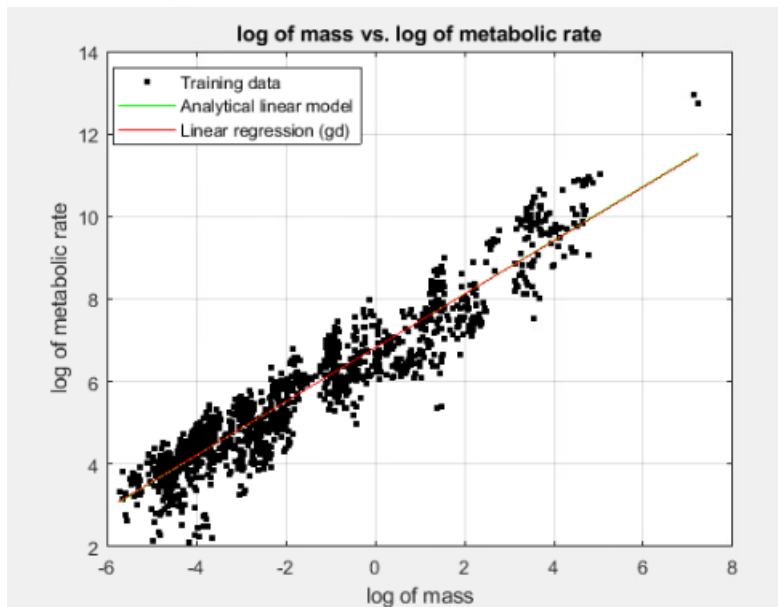
```
2 and 3 ) %log(htheta) = theta0+log(X)*theta;
           %htheta = exp(log_htheta);
           %can also be expressed as: htheta = theta0*year^theta1
           %so we can find optimal theta by looking at the linear relationship between
           %the logs and then calculating our prediction by taking e to the power of
           %our hypothesis function
```

It is a log-log linear regression

So: our hypothesis function would be:  $y = \theta_0 * (\text{year}^{\theta_1})$

But, if we keep the cost function and linear descent we've been using to this point, we can get theta's that are optimal for log of our desired result/prediction (i.e.  $\exp(\theta_0 * (\text{year}^{\theta_1}))$ )

If we plot the analytical linear model vs the linear regression, we can see they almost overlap:



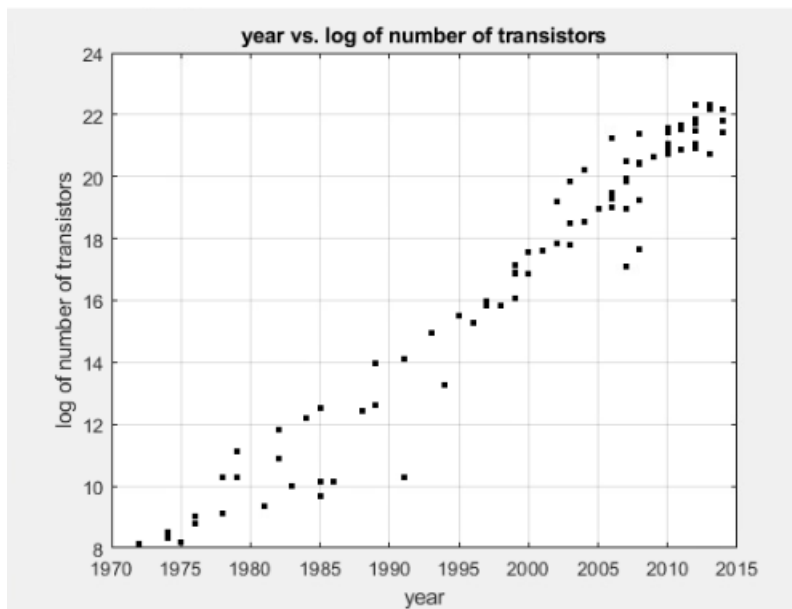
4 and 5) `predicted calories for a mammal weighing 15 kg is 1265.01`  
`estimated weight for a mammal that needs 2.5 kJoul per day is : 1.143892e-04 kilograms`

5)

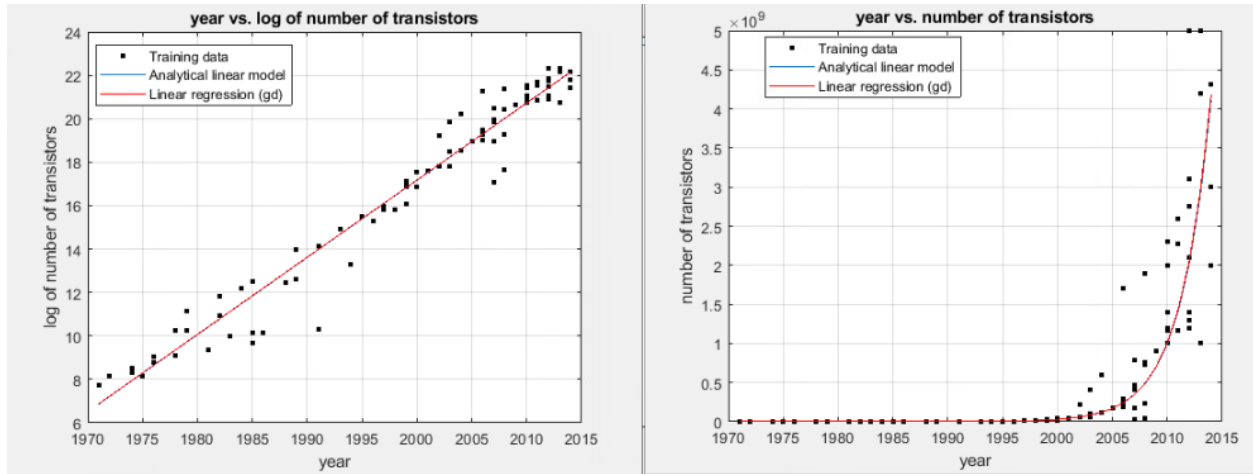
1) We can see there is exponential growth in the number of transistors

Therefore, the transformation is from  $y$  to  $\log(y)$

(Our hypothesis function is:  $\log(y) = x \cdot \theta$ )



2)



3)

The predicted number of transistors in 2018 is is 1.749930e+10