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```
% Project 3
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% 02/25/16
```

Problem 1

```
% In order to load the data into Matlab I first saved the excel file onto
% my desktop. I then used the Matlab tool "Import Data" found under the
% home tab to import the data into Matlab before renaming each vector
% to a more discriptive name.
```

```
load('Project_3_data');
```

```
% In order to then produce the data as a function of how many days out
% the prediction was made I turned it into a data structure. To do this
% I made the data structure 'janurary' and made 9 different fields for
% each type of data. Then I set each of the rows of each field equal to
% 31 day intervals. This creates it so that each row of each field
% is a 31 x 1 vector of all of the predictions for so many days out. For
% example the first row consists of all of the same day predictions, the
% second row the next day predictions, so on and so on.
```

```
k = 1;
```

```
for i = 1:31
```

```
    janurary(i).AvgHi = AvgHi(k:k+30,1);
    janurary(i).AvgLo = AvgLo(k:k+30,1);
    janurary(i).High = High(k:k+30,1);
    janurary(i).Low = Low(k:k+30,1);
    janurary(i).Precip = Precip(k:k+30,1);
    janurary(i).Snow = Snow(k:k+30,1);
    janurary(i).TruHi = TruHi(k:k+30,1);
    janurary(i).TruLo = TruLo(k:k+30,1);
    janurary(i).TruPrecip = TruPrecip(k:k+30,1);
```

```
k = k+31;
```

```
end
```

Problem 2

%The following for loop creates a second data structure named DiffTemp
%with two fields 'prediction' and 'average'. The prediction field holds
%the difference in the Accuweather predicted hightemps and the actual
%high temperature, while the average field holds the difference in the
%historical average high temperatures and the actual high temperatures.
%This is more of a random or know nothing prediction which can then be
%compared to the Accuweather predictions.

```
for i = 1:31
    h = i;
    j= 1;
    for k = h:31
        DiffTemp(i).prediction(j,1) = abs(janurary(j).High(k,1)-...
            janurary(j).TruHi(k,1));

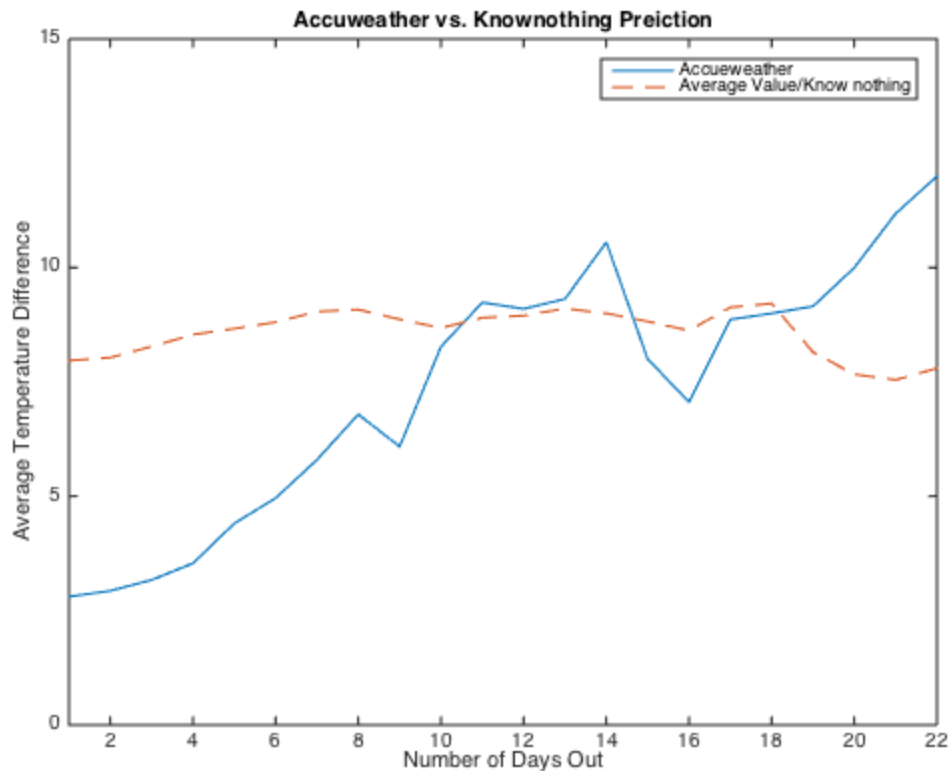
        DiffTemp(i).average(j,1) = abs(janurary(j).AvgHi(k,1)-...
            janurary(j).TruHi(k,1));

        j = j+1;
    end
end

AvgDiffTempprediction = zeros(31,1);
AvgDiffTempaverage= zeros(31,1);

%Find the average of both the technology based and the know nothing
%predictions.
for i = 1:31
    AvgDiffTempprediction(i,1) = mean(DiffTemp(i).prediction);
    AvgDiffTempaverage(i,1) = mean(DiffTemp(i).average);
end

%Plot up the two predictions to graphically compare them.
figure(1)
plot([1:31],AvgDiffTempprediction,[1:31],AvgDiffTempaverage,'--')
axis([1,22,0,15])
xlabel('Number of Days Out');
ylabel('Average Temperature Difference')
title('Accuweather vs. Knownothing Preiction')
legend('Accuweather','Average Value/Know nothing')
```



Problem 3

%Find the Standard Deviation of the average of of both predictions and plot
 %this on the graph in the form of error bars to see the trend of the error
 %as the predicitons get further out.

```
stdDiffTempprediction = zeros(31,1);
stdDiffTempaverage = zeros(31,1);
for i = 1:31
    sigmaDiffTempprediction(i,1) = std(DiffTemp(i).prediction/sqrt((32-i)));
    sigmaDiffTempaverage(i,1) = std(DiffTemp(i).average/sqrt((32-i)));
end
```

```
figure(2)
```

```
plot([1:31],AvgDiffTempprediction,'b',[1:31],AvgDiffTempaverage,'r--')
axis([1,22,0,15])
xlabel('Number of Days Out')
ylabel('Average Difference in Temperature')
title(['Error Comparison Between Accuweather and Historical Average',...
       ' Predictions'])
legend('Accuweather','Average Value/Know nothing')
```

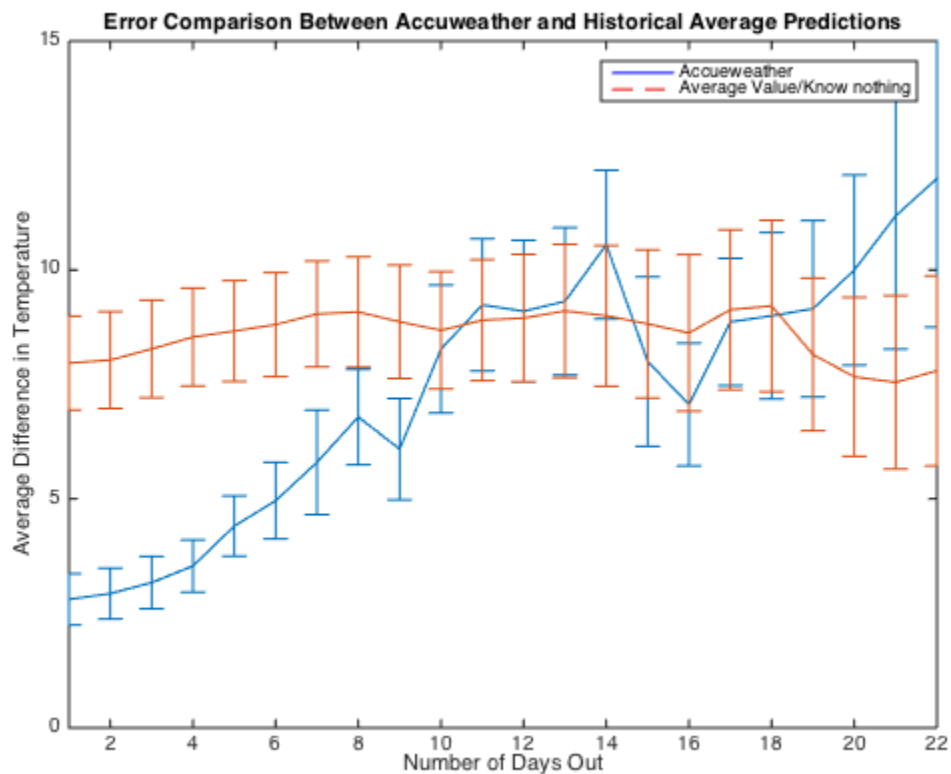
```
hold on
errorbar([1:31],AvgDiffTempprediction,sigmaDiffTempprediction,...
```

```

        sigmaDiffTempprediction)
    errorbar([1:31],AvgDiffTempaverage,sigmaDiffTempaverage,...
        sigmaDiffTempaverage)
    hold off

```

%Form observing this graph it becomes clear that as the predictions get further out the standard deviation between the predicted high temp. and the actual high temp gets larger and larger. This shows that as the predictions get further out they get significantly less reliable. This graph also shows that the Accuweather error becomes equal to the Historical Average error after about 7 days showing that the technology that Accuweather uses is really only helpful to about 7 days out.



Problem 4

```

%Create a vector of just the 30 next day predictions.
NextDayPred = DiffTemp(2).prediction;

%Calculate the Residuals each of these Predictions
Residual = NextDayPred-mean(NextDayPred);

figure(3)
plot(Residual)
xlabel('Day in Janurary')
ylabel('Residual')
title('Residual of the Next Day Weather Predictions in Juanrury')

```

```

legend('Residual')
%Based on this graph it would appear that the residuals are very random.
%This shows that the data is probably independent.

%Calculate the Variance and Covariance
Covar = dot(Residual(1:29,1),Residual(2:30,1))/(29-1);
Var = dot(Residual,Residual)/(30-1);

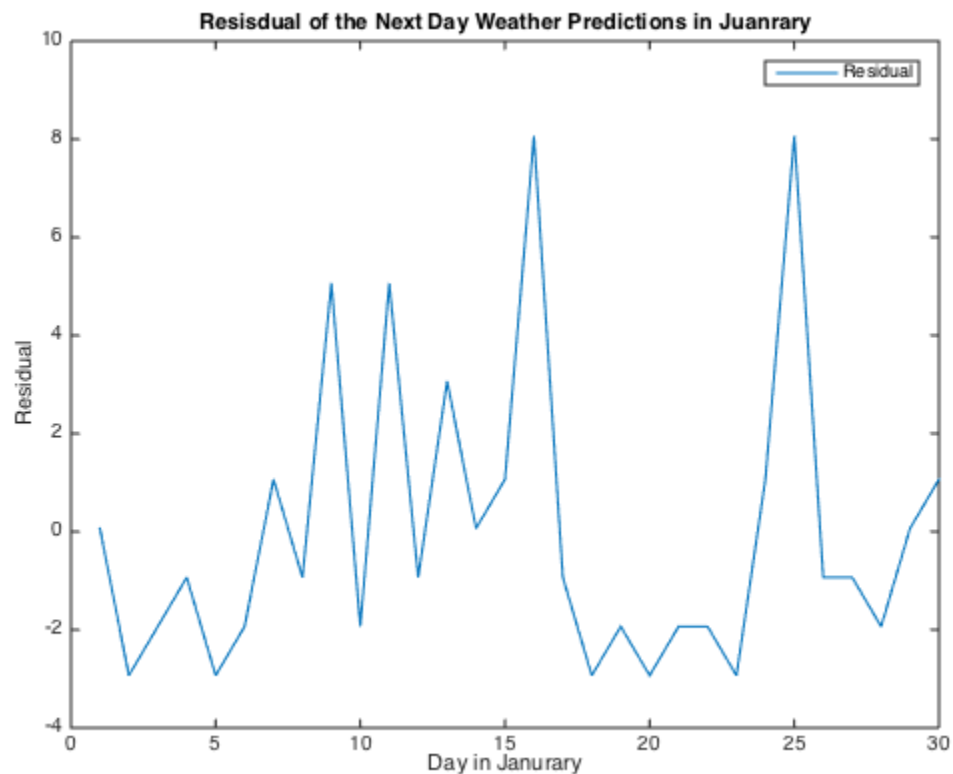
Ratio = Covar/Var;

disp(Ratio)

% As you can see the ratio between the Covariance and Variance is very
% small. This shows that the Covariance is very low compared to the
% Variance shows the days are not dependent on eachother and shows that
% the data is infact random.

0.0453

```



Problem 5

```

%In order to turn the results from part 3 into a probability I first
%calculated the T statistic and then used a normal cumulative distribution
%function.By plotting this with a semilogy scale I was properly able to
%plot the probability that you are better off with the historical

```

```

%average prediction rather than the Accuweather prediction.

sigmaTstat = (sigmaDiffTempprediction.^2 + sigmaDiffTempaverage.^2).^5;

Tstat=(AvgDiffTempprediction-AvgDiffTempaverage)./sigmaTstat;

Tstat(31,1) = 0; %The error for the prediction will be 0 because there is
                %Only one measurement for this value.

Prob = normcdf(Tstat,0,1);

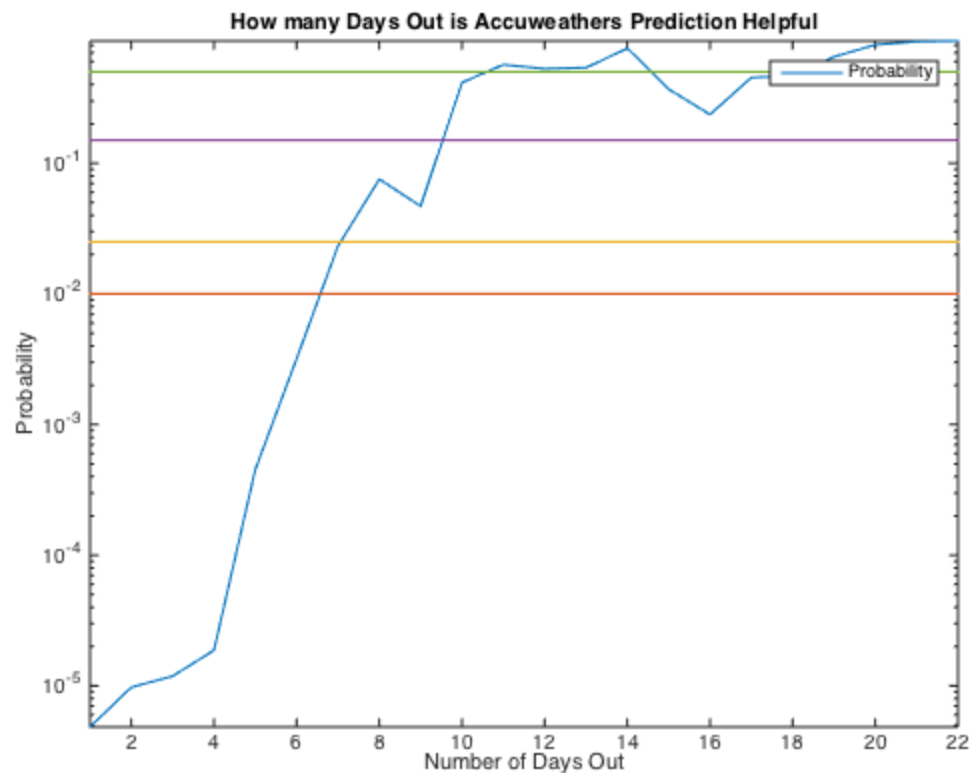
% Create Lines at 1 percent, 2.5 percent, 15 percent and 50 percent.
days=linspace(1,22,22);
prob1 = 0.01*ones(1,22);
prob2 = 0.025*ones(1,22);
prob3 = 0.15*ones(1,22);
prob4 = 0.5*ones(1,22);

figure(4)

semilogy(Prob)
axis([1,22,0,inf]);

hold on
xlabel('Number of Days Out')
ylabel('Probability')
title('How many Days Out is Accuweathers Prediction Helpful')
legend('Probability')
plot(days,prob1,days,prob2,days,prob3,days,prob4)
hold off
% By observing this graph it becomes clear that the Accuweather prediction
% is in the 99% confidence up to about 6 days out. Once the prediction
% gets to about 10 days out the probability of it
% being more accurate than the Know-nothing prediction becomes a coinflip
% showing that the predictioin is more or less becomes useless.

```



Problem 6

%This for loop is necessary because there is at least one value for
 %precipitation that is over 1. What the for loop does is it creates a new
 %datastructure Precip1 that will put a 1 for any day that the prediction
 %on if it would precipitate or not was incorrect and a 0 for any day that
 %the prediction was correct. This lets us see how accurate Accuweather
 %predicted the percipitation as a function of how far in advanced they
 %were predicting it.

```
for i = 1:31
    h = i;
    j= 1;
    for k = h:31
        if janurary(j).Precip(k,1)>1
            janurary(j).Precip(k,1)=1;
            Precip1(i).error(j,1) = abs(ceil(janurary(j).Precip(k,1))-...
            ceil(janurary(j).TruPrecip(k,1)));
        elseif janurary(j).TruPrecip(k,1)>1
            janurary(j).TruPrecip(k,1)=1;
            Precip1(i).error(j,1) = abs(ceil(janurary(j).Precip(k,1))-...
            ceil(janurary(j).TruPrecip(k,1)));
        else
            Precip1(i).error(j,1) = abs(ceil(janurary(j).Precip(k,1))-...
            ceil(janurary(j).TruPrecip(k,1)));
        end
    end
end
```

```

        end
        j = j+1;
    end
end

%Create a Rate matrix that is the sum of all the missed predictions
%divided by the total number of predictions for each prediction range.
N = zeros(31,1);
Rate = zeros(31,1);

for i = 1:31
    N(i,1) = sum(Precip1(i).error);
    Rate(i,1) = N(i,1)./(32-i);
end

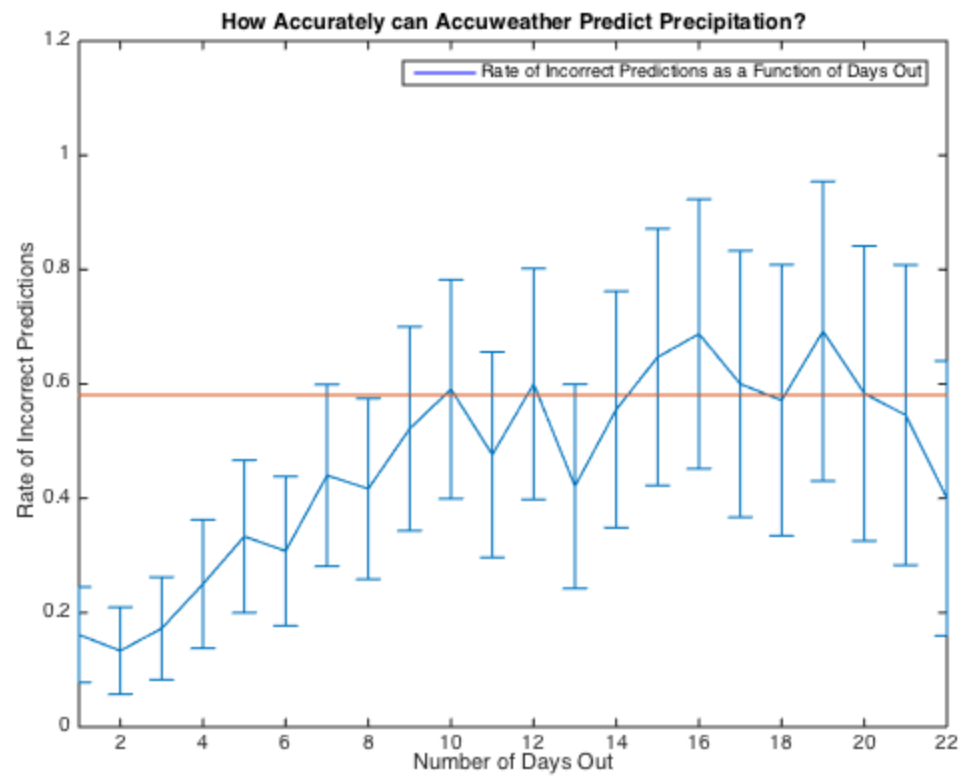
for i = 1:31
    sigmaRate(i) = (Rate(i).^(1-Rate(i)))/(32-i)).^.5;
end

averagefrac = (18/31)*ones(1,31); %There were 18 days in Janurary with
                                   %precipitation.

%By plotting the Rate of incorrect predictions up with errorbars and
%a line showing the average fraction of days in Janurary that it actually
%rained you can very well see that once again as the prediction gets
%further and further out it becomes less and less accurate. At right
%around 10 days the graph crosses the average percentage of times it
%actually rained in Janurary showing that at 10 days out or more
%Accuweathers prediction of the precipitation is not a very good one.
figure(5)

plot([1:31],Rate,'b')
axis([1,22,0,1.2])
xlabel('Number of Days Out')
ylabel('Rate of Incorrect Predictions')
title('How Accurately can Accuweather Predict Precipitation?')
legend('Rate of Incorrect Predictions as a Function of Days Out')
hold on
errorbar([1:31],Rate,sigmaRate,sigmaRate)
plot([1:31],averagefrac)
hold off

```

Published with MATLAB® R2014b