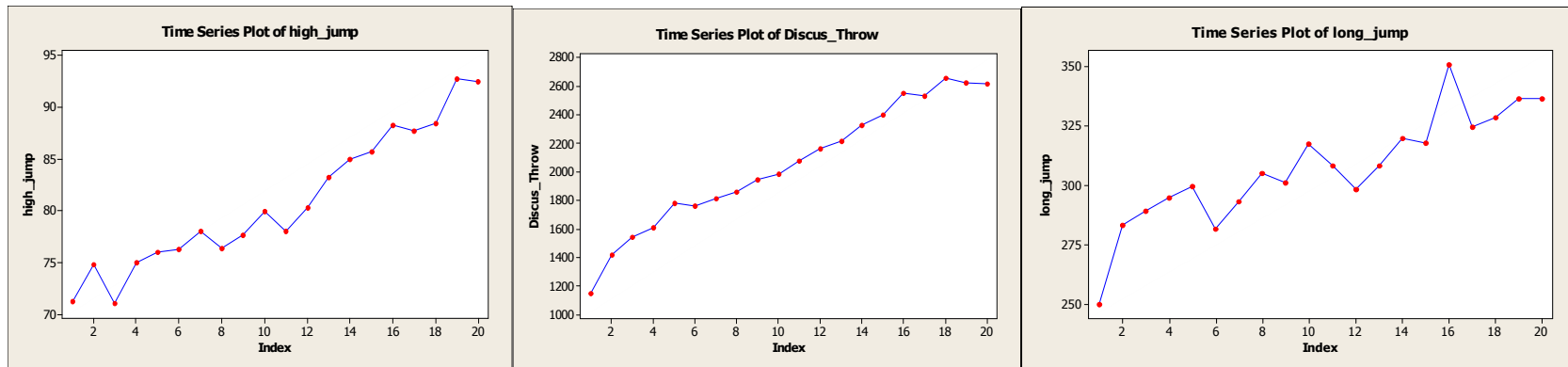


Minitab Project Report – Assignment 1

1.3 Time series plot



All the data for the three Olympic events have **no seasonality** by their nature, however the plots indicate strong **increasing trend** in each series.

None of the series has a noise component increasing in time.

There are no sudden peaks or clear turning points.

The Discus Throw is most 'smooth', while High Jump and Long Jump results, although increasing, have jumps up and down.

1.4 Trend analysis

1.4.1 Regression model fit

Accuracy measures

MAPE Mean Absolute Percentage Error (MAPE) measures the accuracy of fitted time series values. It expresses accuracy as a percentage.

$$\text{MAPE} = \frac{1}{n} \sum |(y_t - \hat{y}_t) / y_t| 100$$

where y_t equals the actual value at time t , \hat{y}_t equals the fitted value, and n equals the number of observations.

MAD Mean Absolute Deviation (MAD) measures the accuracy of fitted time series values. It expresses accuracy in the same units as the data, which helps conceptualize the amount of error.

$$\text{MAD} = \frac{1}{n} \sum |y_t - \hat{y}_t|$$

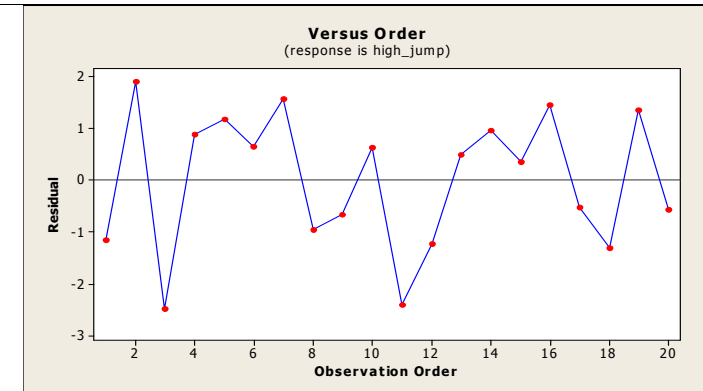
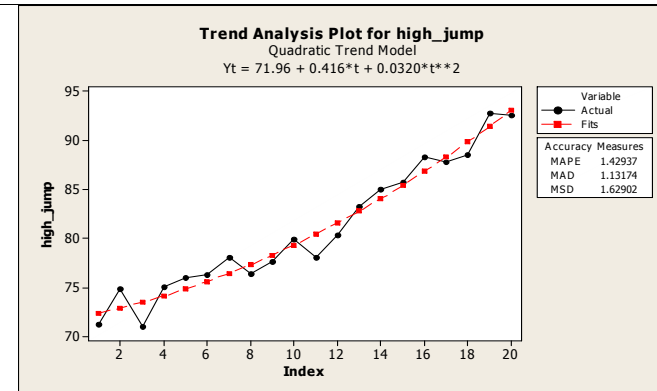
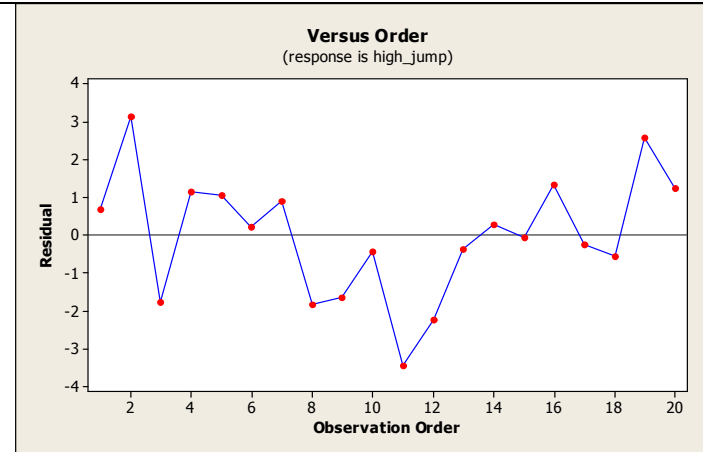
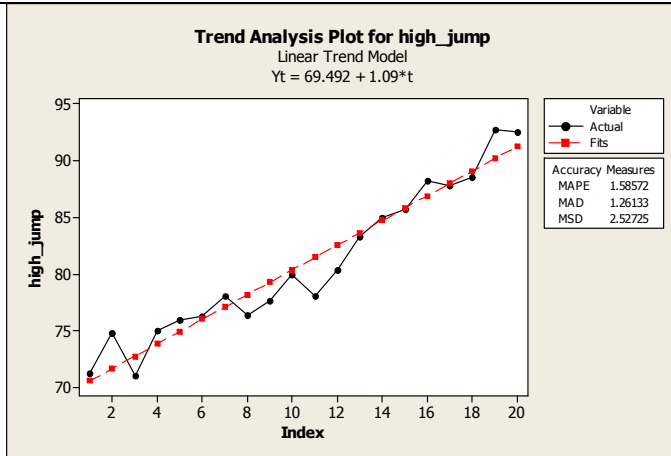
where y_t equals the actual value at time t , \hat{y}_t equals the fitted value, and n equals the number of observations.

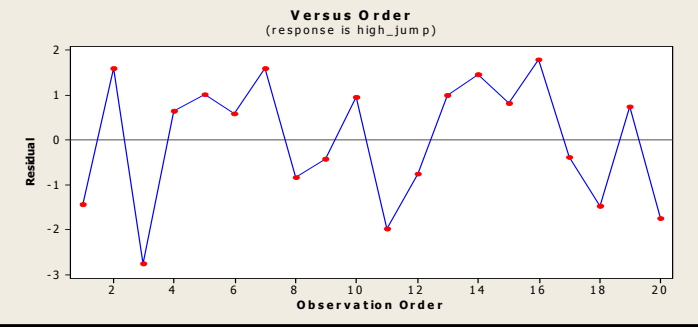
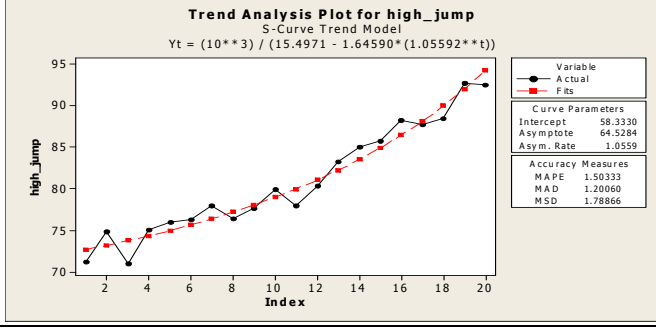
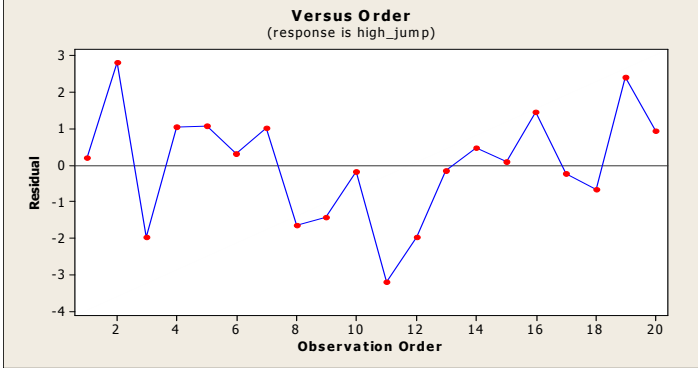
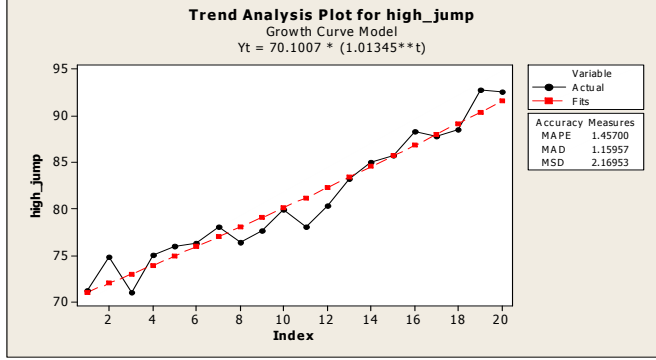
MSD Mean Squared Deviation (MSD) is always computed using the same denominator, n , regardless of the model. MSD is a more sensitive measure of an unusually large forecast error than MAD.

$$\text{MSD} = \frac{1}{n} \sum (y_t - \hat{y}_t)^2$$

where y_t equals the actual value at time t , \hat{y}_t equals the forecast value, and n equals the number of forecasts.

Trend Analysis for high_jump





Fitted Trend Equation

$$Y_t = 69.492 + 1.09 * t$$

Accuracy Measures

MAPE 1.58572
MAD 1.26133
MSD 2.52725

Fitted Trend Equation

$$Y_t = 71.96 + 0.416 * t + 0.0320 * t ** 2$$

Accuracy Measures

MAPE **1.42937**
MAD **1.13174**
MSD **1.62902**

Fitted Trend Equation

$$Y_t = 70.1007 * (1.01345 ** t)$$

Accuracy Measures

MAPE 1.45700
MAD 1.15957
MSD 2.16953

Fitted Trend Equation

$$Y_t = (10 ** 3) / (15.4971 - 1.64590 * (1.05592 ** t))$$

Accuracy Measures

MAPE 1.50333
MAD 1.20060
MSD 1.78866

Comments:

Model chosen: $X_t = \beta_0 + \beta_1 t + \beta_2 t^2 + Y_t$ where Y_t denotes random noise.

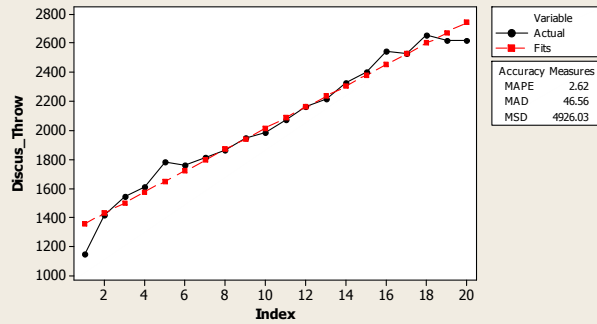
The smallest values of the model accuracy measures are for the quadratic trend, the residuals seem to be scattered about zero with no indication of trend or of non-constant variance.

None of the models might be good for prediction of far future values of the high jump, as they all sharply increase, what may not be realistic, though they may be fine for prediction of the value for next Olympic game. We would expect the increase in the performance to 'slow down' simply due to limitations of human body.

Trend Analysis for Discus_Throw

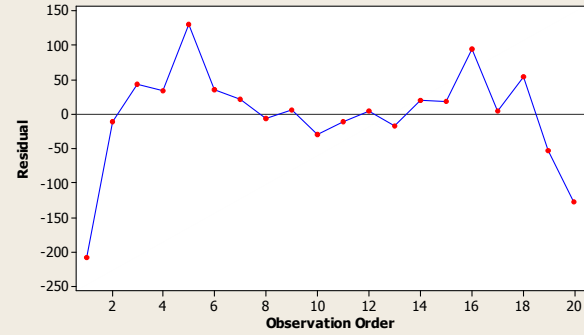
Trend Analysis Plot for Discus_Throw

Linear Trend Model
 $Y_t = 1282.9 + 73.4 * t$



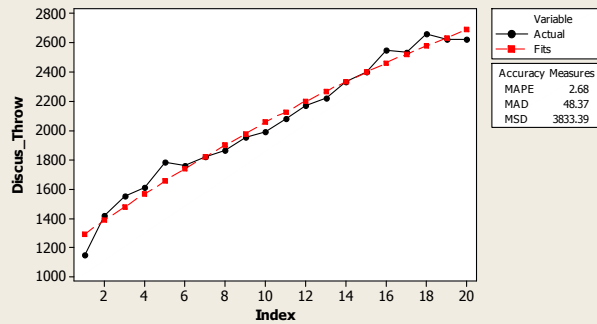
Versus Order

(response is Discus_Throw)



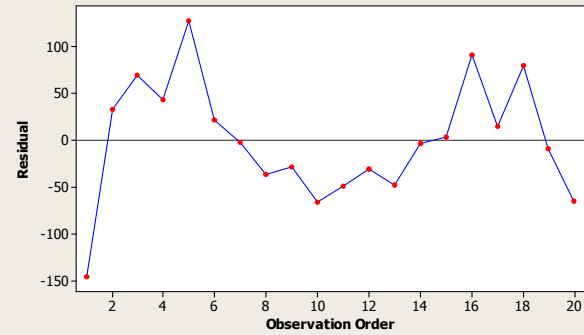
Trend Analysis Plot for Discus_Throw

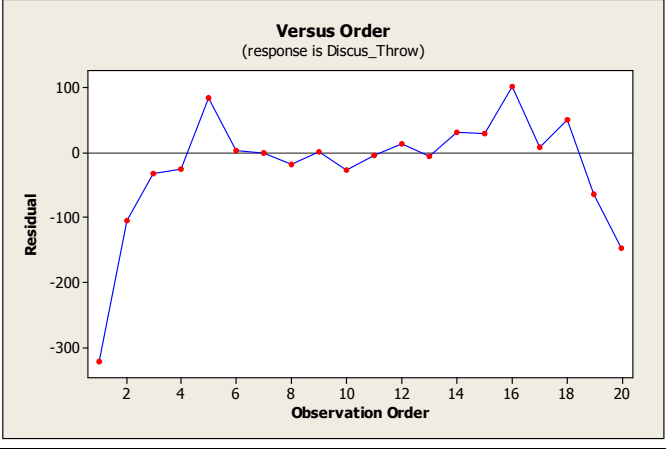
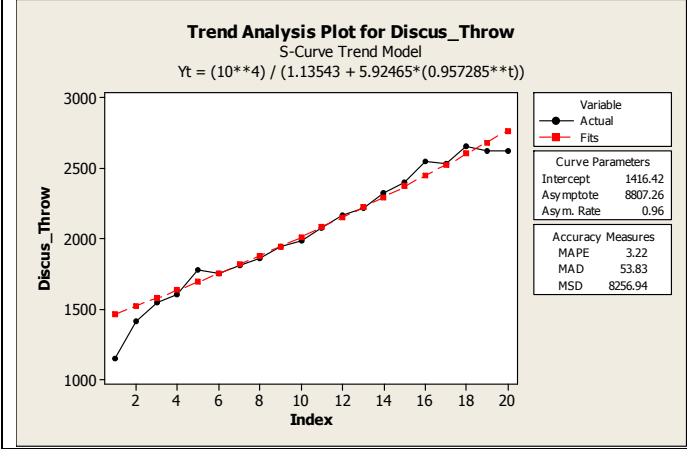
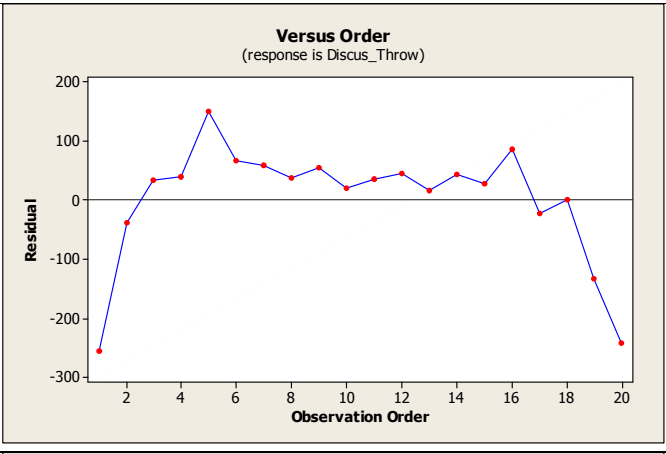
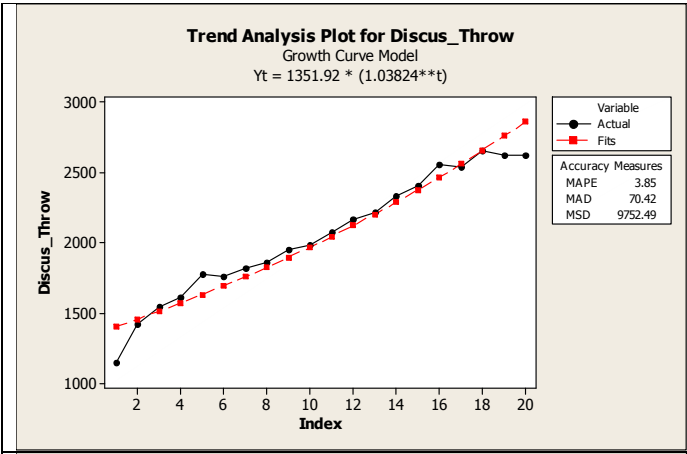
Quadratic Trend Model
 $Y_t = 1197.0 + 96.8 * t - 1.116 * t^2$



Versus Order

(response is Discus_Throw)





Fitted Trend Equation

$$Y_t = 1282.9 + 73.4*t$$

Accuracy Measures

MAPE	2.62
MAD	46.56
MSD	4926.03

Fitted Trend Equation

$$Y_t = 1197.0 + 96.8*t - 1.116*t^{**2}$$

Accuracy Measures

MAPE	2.68
MAD	48.37
MSD	3833.39

Fitted Trend Equation

$$Y_t = 1351.92 * (1.03824^{**t})$$

Accuracy Measures

MAPE	3.85
MAD	70.42
MSD	9752.49

Fitted Trend Equation

$$Y_t = (10^{**4}) / (1.13543 + 5.92465*(0.957285^{**t}))$$

Accuracy Measures

MAPE	3.22
MAD	53.83
MSD	8256.94

Comments

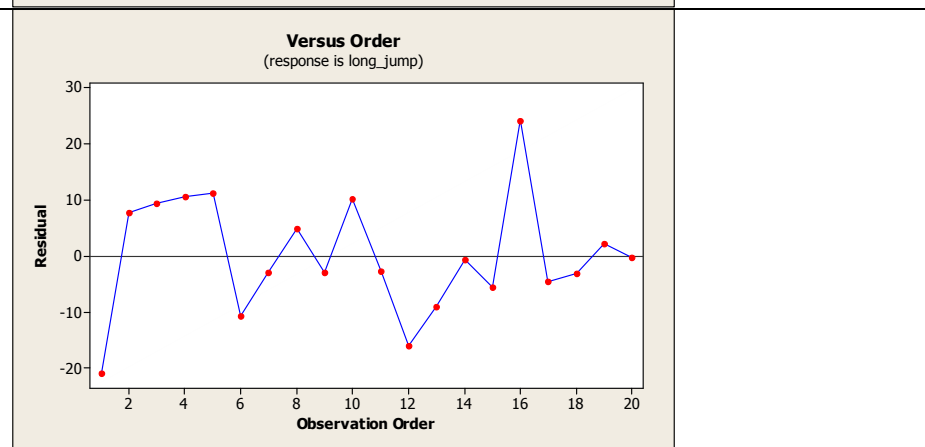
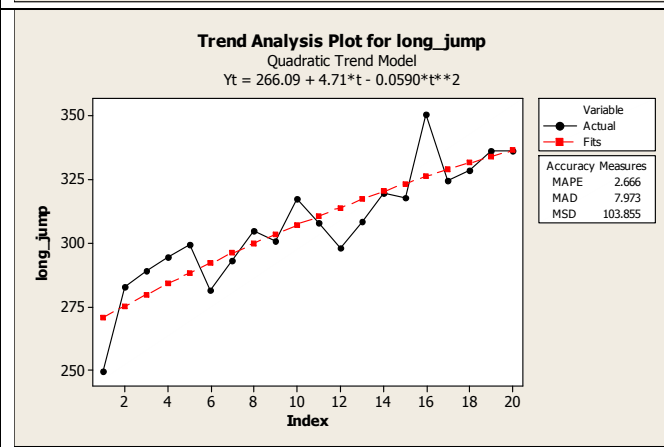
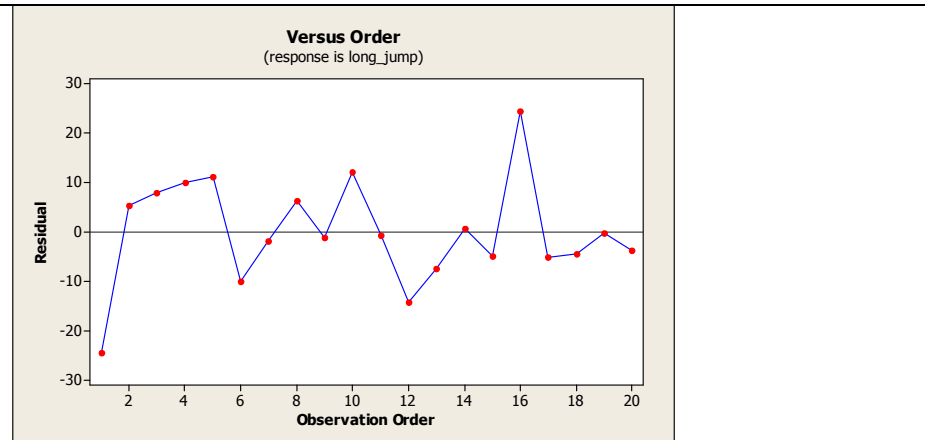
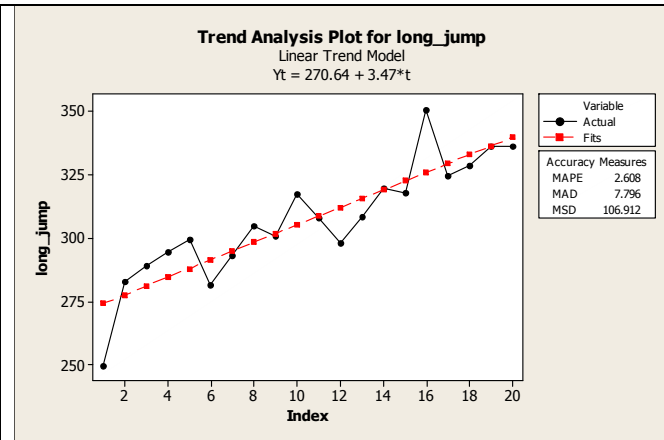
Model chosen: $X_t = \beta_0 + \beta_1 t + \beta_2 t^2 + Y_t$ where Y_t denotes random noise.

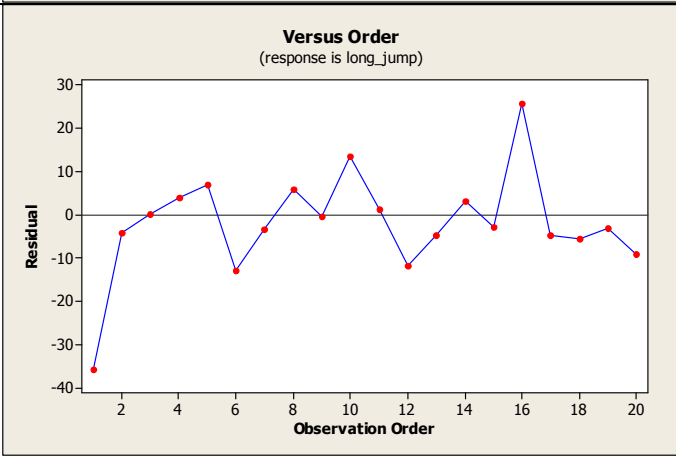
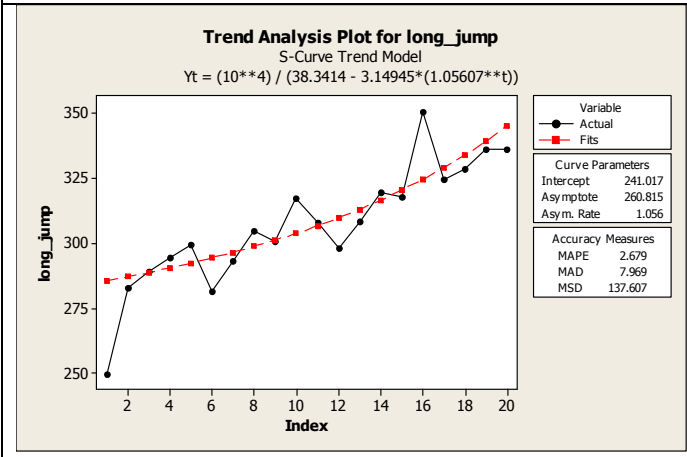
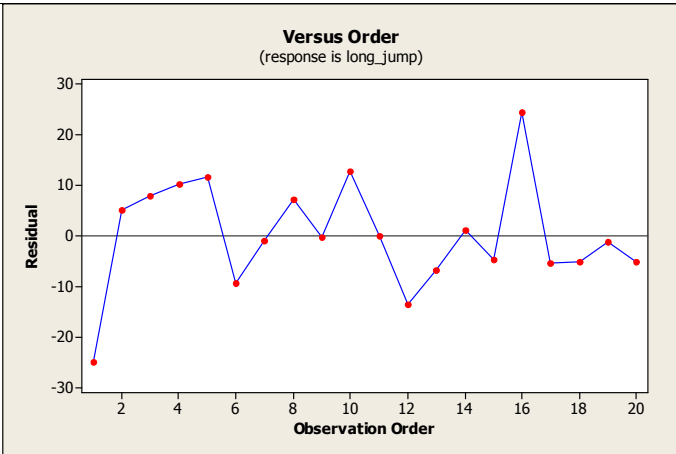
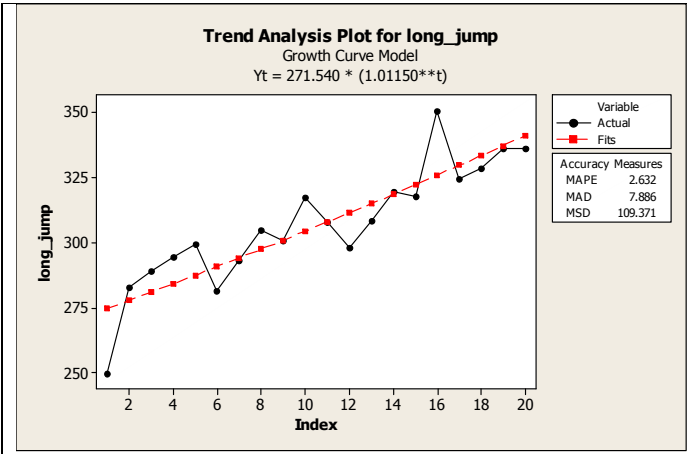
None of the trend models has all three accuracy measures best. Linear trend has smallest MAPE and MAD, while the quadratic trend has smallest MSD. The other two trend models have larger measures, particularly MSD.

Comparing the linear and quadratic trends, we see that

- there is not much difference between MAPE and MAD criteria,
- MSD is much lower for the quadratic model,
- the quadratic trend fits better at the ends of the series, particularly at the last value. It may give better prediction values for near future,
- in general, its residuals are less varied.

Trend Analysis Plot for long_jump





Fitted Trend Equation

$$Y_t = 270.64 + 3.47*t$$

Accuracy Measures

MAPE	2.608
MAD	7.796
MSD	106.912

Fitted Trend Equation

$$Y_t = 266.09 + 4.71*t - 0.0590*t^{**2}$$

Accuracy Measures

MAPE	2.666
MAD	7.973
MSD	103.855

Fitted Trend Equation

$$Y_t = 271.540 * (1.01150^{**t})$$

Accuracy Measures

MAPE	2.632
MAD	7.886
MSD	109.371

Fitted Trend Equation

$$Y_t = (10^{**4}) / (38.3414 - 3.14945*(1.05607^{**t}))$$

Accuracy Measures

MAPE	2.679
MAD	7.969
MSD	137.60

Comments

Model chosen: $X_t = \beta_0 + \beta_1 t + \beta_2 t^2 + Y_t$ where Y_t denotes random noise.

None of the trend models has all three accuracy measures best. Linear trend has smallest MAPE and MAD, while the quadratic trend has smallest MSD. The other two trend models have larger values of the measures.

Comparing the linear and quadratic trends, we see that

- there is not much difference between the criteria,
- the quadratic trend fits better at the end of the series, particularly at the last value. It may give better prediction values for near future,
- in general, its residuals of all trend models look stationary, but the quadratic model gives residuals on a smaller range.

Trend models (careful with notation)

Linear Trend analysis by default uses the *linear trend* model:

$$Y_t = \beta_0 + \beta_1 t + e_t$$

In this model, β_1 represents the average change from one period to the next.

Quadratic The *quadratic trend model* which can account for simple curvature in the data, is:

$$Y_t = \beta_0 + \beta_1 t + \beta_2 t^2 + e_t$$

Exponential Growth The *exponential growth trend model* accounts for exponential growth or decay. For example, a savings account might exhibit exponential growth. The model is:

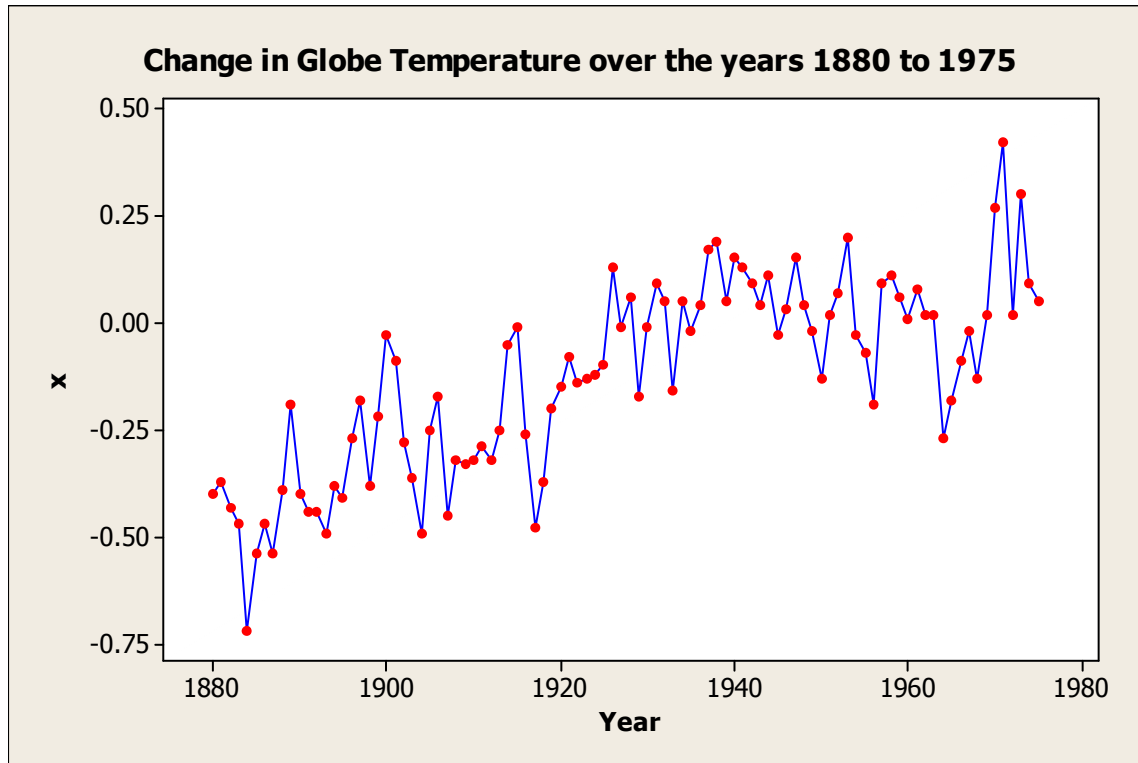
$$Y_t = \beta_0 * \beta_1^t * e_t$$

S-curve The *S-curve model* fits the Pearl-Reed logistic trend model. This accounts for the case where the series follows an S-shaped curve. The model is:

$$Y_t = 10^a / (\beta_0 + \beta_1 \beta_2^t) + e_t$$

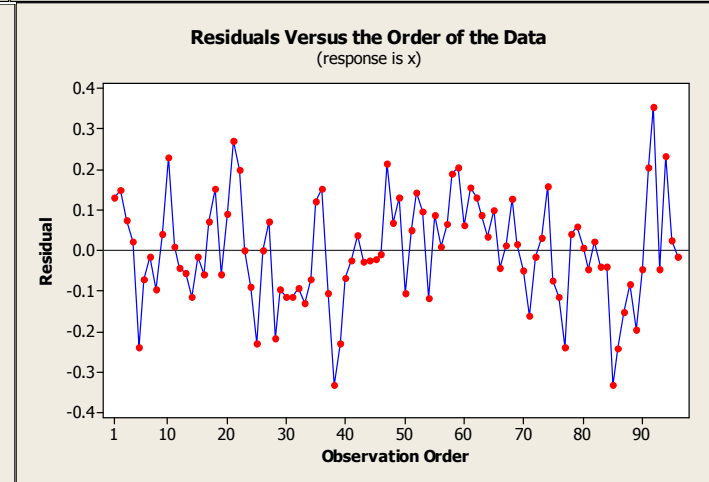
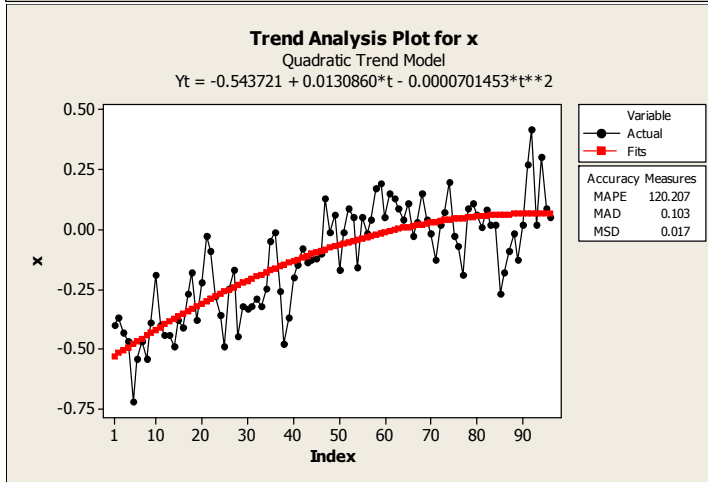
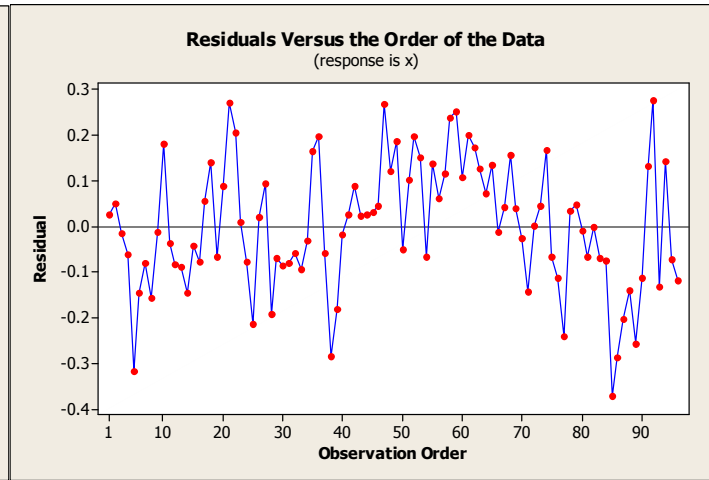
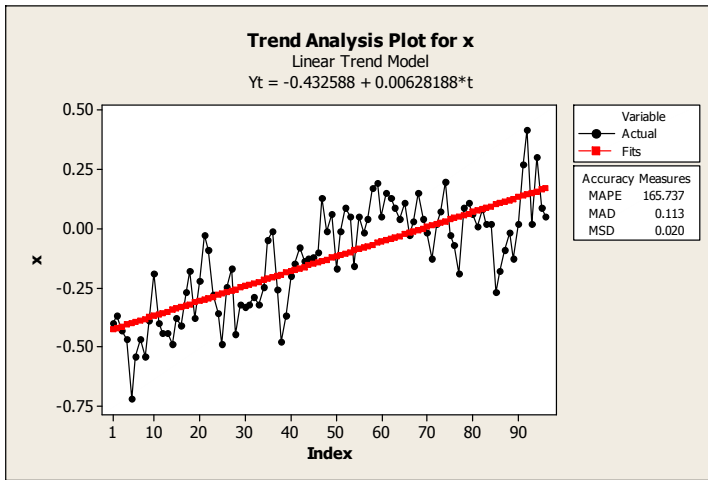
Note that the Exponential Growth model is multiplicative and so the residuals are the observations divided by the fitted values of the model. What we obtain as 'residuals' in Minitab is the difference between the two. The shape of the plot of the residuals is kept, but not the scale.

Change in Temperature Data



Comments:

There is steady upward trend in the change of temperature data.



Trend Analysis for x

$$Y_t = -0.432588 + 0.00628188*t$$

Accuracy Measures

MAPE 165.737
MAD 0.113
MSD 0.020

$$Y_t = -0.543721 + 0.0130860*t - 0.0000701453*t**2$$

Accuracy Measures

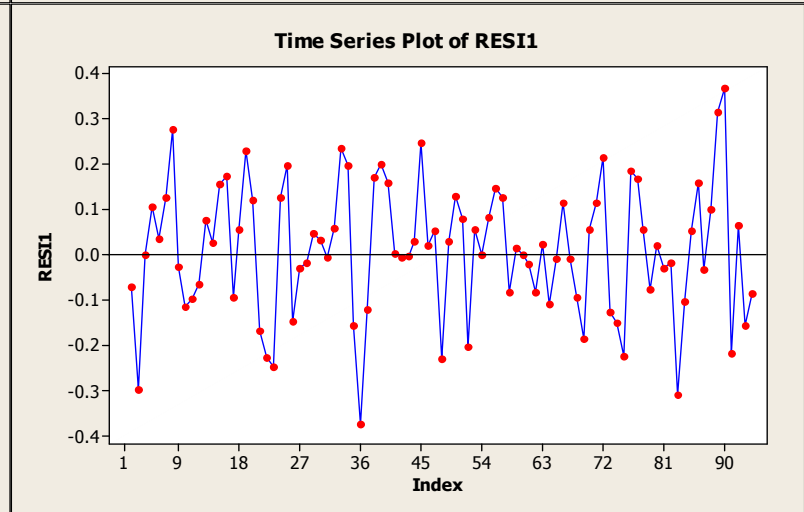
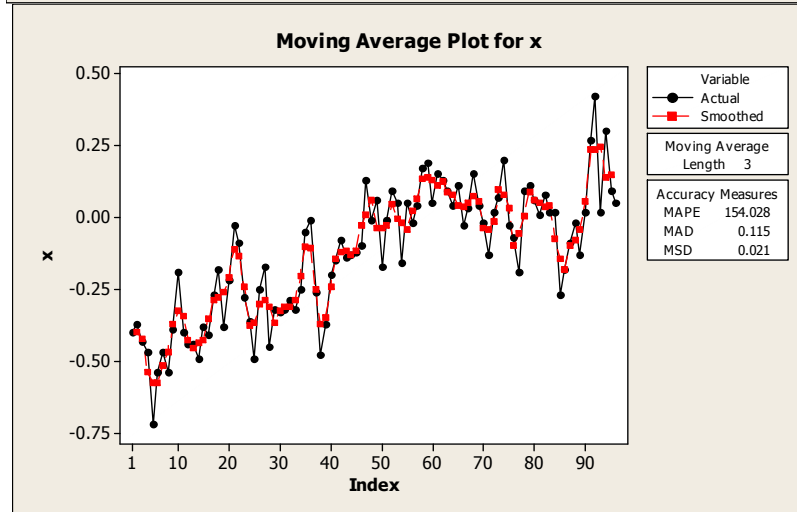
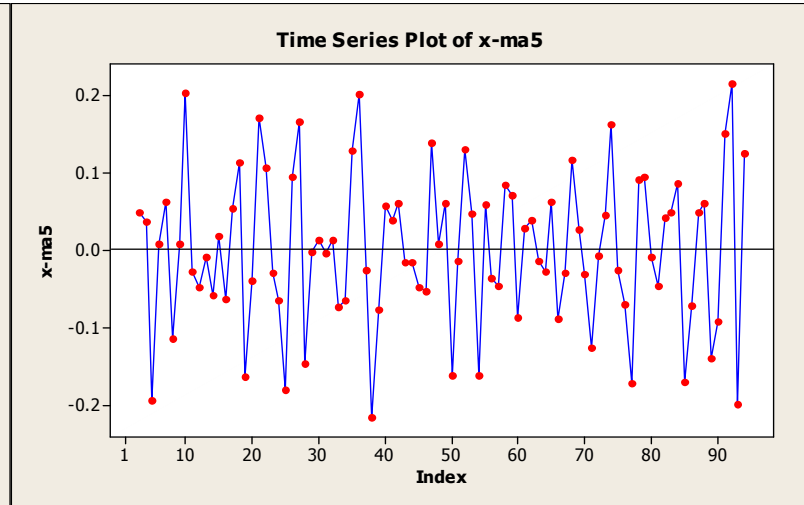
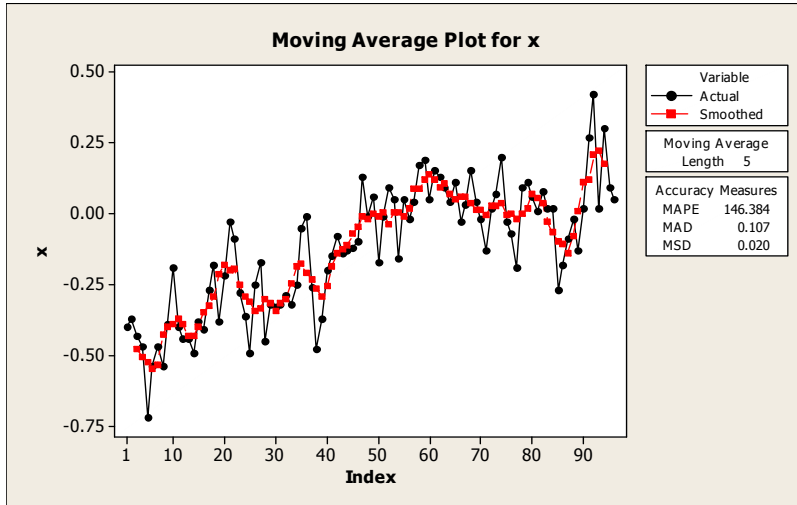
MAPE 120.207
MAD 0.103
MSD 0.017

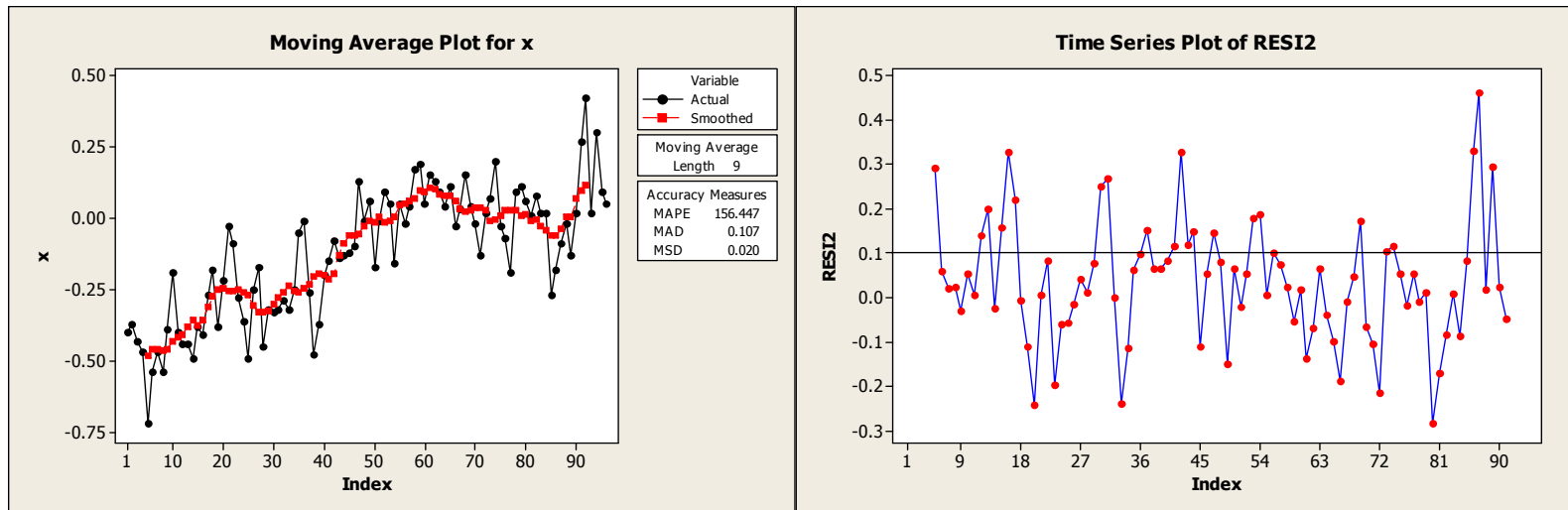
The accuracy measures give smaller values for the quadratic trend fit.

The growth curve and S-curve could not be fitted, as they assume that the values are all positive, while this TS has both negative and positive values.

Indeed, the quadratic trend seems to fit quite well, the residuals show less wavy pattern than those for the linear trend.

1.4.2 Moving Average Change in Temperature Data





Obviously, the smaller is the length of m.a. the closer it is to the original series, and so the residuals have smaller range and are less bursty.

The m.a. of length 9 is the smoothest among the three but gives most bursty residuals, reflecting the burstiness of the original TS.