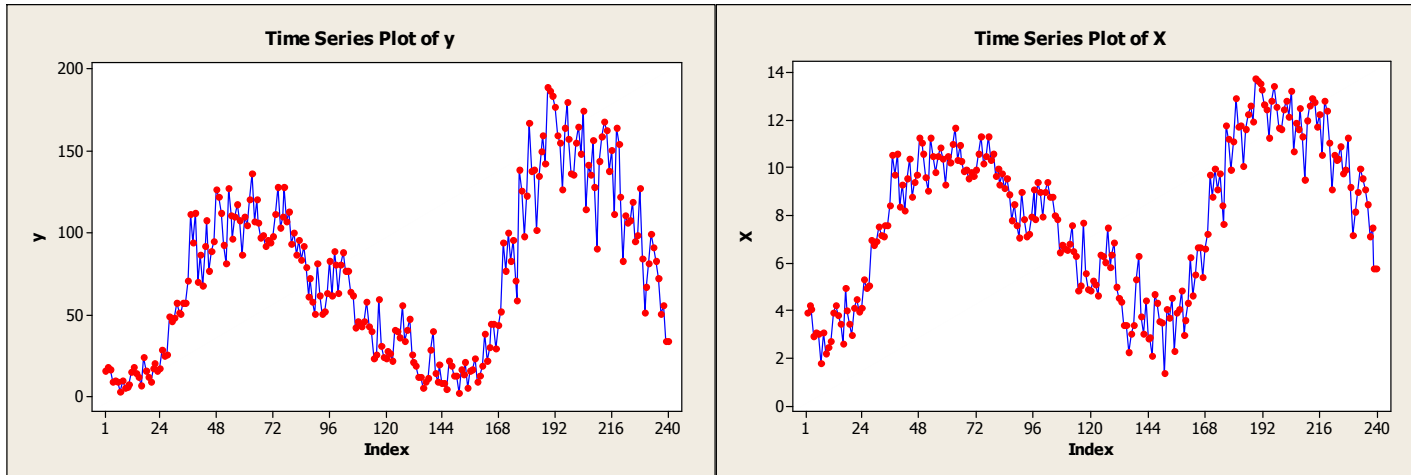


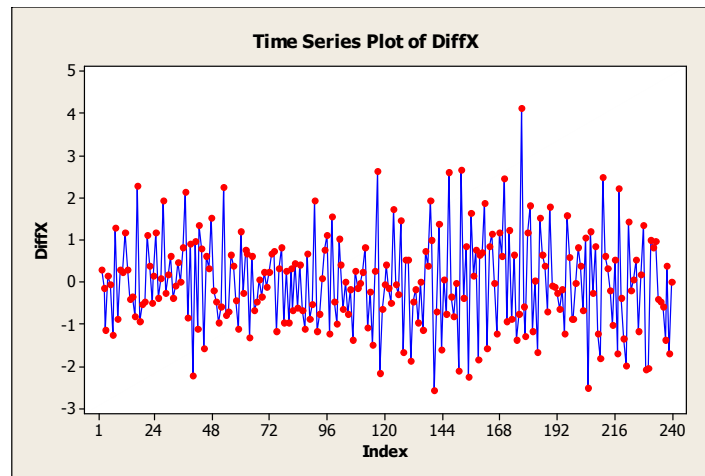
Minitab Project Report - Assignment 6

6.1.1 Sunspot data

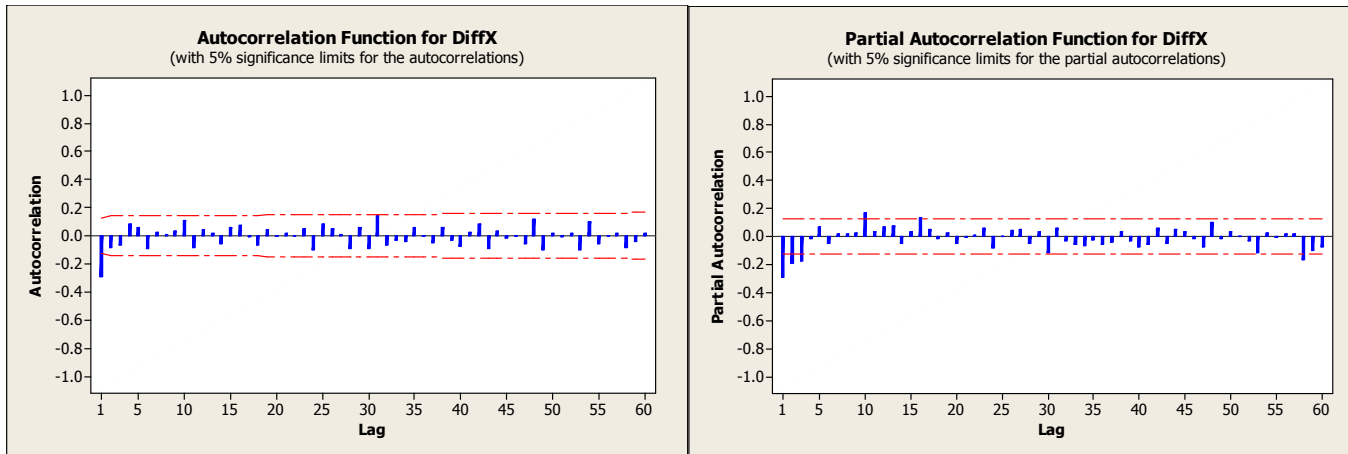


The data have a wavy pattern. However, they do not show any seasonality. There seem to be an increasing variability at the higher levels of the observations.

A power transformation (Box-Cox transformation performed in Minitab) stabilizes the variance.



DiffX: the differenced data represent the monthly changes in the transformed (by square root) numbers of sunspots. DiffX series looks stationary with zero mean.



Autocorrelation Function: DiffX

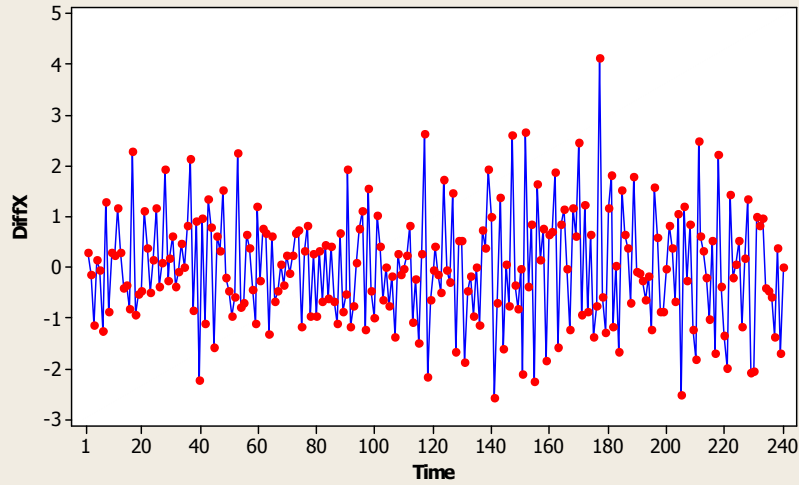
Lag	ACF	T	LBQ
1	-0.294591	-4.55	21.00
2	-0.086777	-1.24	22.83
3	-0.069790	-0.99	24.02
4	0.086939	1.23	25.87
5	0.062427	0.88	26.83
6	-0.095245	-1.33	29.08

Partial Autocorrelation Function: DiffX

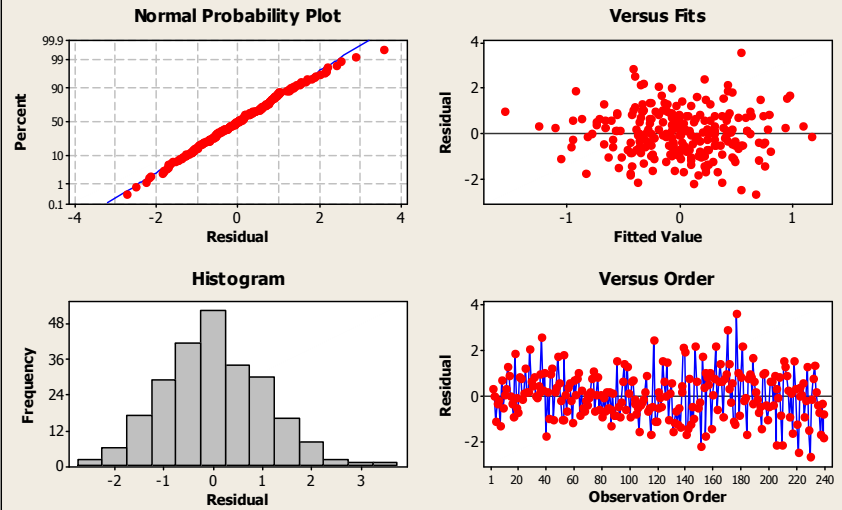
Lag	PACF	T
1	-0.294591	-4.55
2	-0.190054	-2.94
3	-0.177453	-2.74
4	-0.016479	-0.25
5	0.067342	1.04
6	-0.048177	-0.74

The sample ACF and PACF suggest an MA(1) model because the ACF cuts off at lag 1 and the PACF tails off. Also, $\hat{\rho}(1)$ is negative and all the significant sample PACF values are negative – this supports the choice of MA(1) with a negative estimate of θ .

Time Series Plot for DiffX

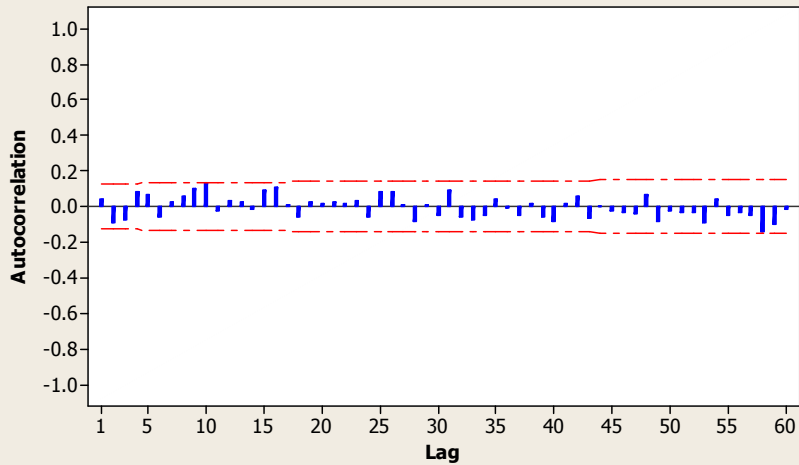


Residual Plots for DiffX



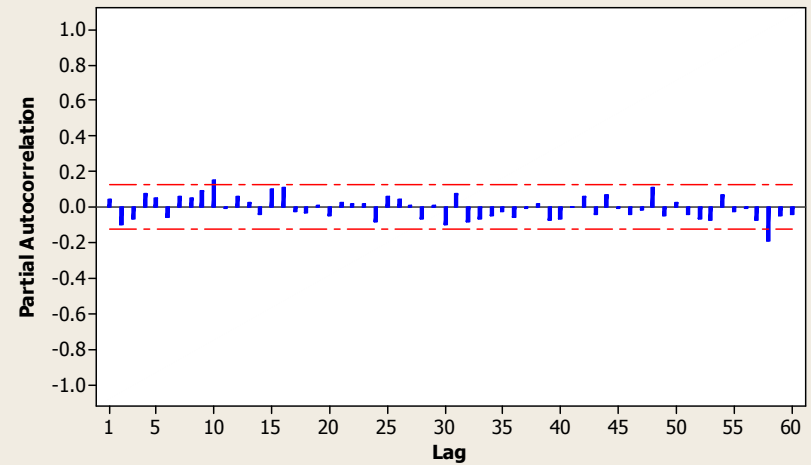
ACF of Residuals for DiffX

(with 5% significance limits for the autocorrelations)



PACF of Residuals for DiffX

(with 5% significance limits for the partial autocorrelations)



Modelling MA(1)

Type	Coef	SE Coef	T	P
MA 1	0.4341	0.0588	7.38	0.000

Modified Box-Pierce (Ljung-Box) Chi-Square statistic

Lag	Chi-Square	DF	P-Value
12	16.3	11	0.132
24	24.2	23	0.390
36	36.6	35	0.395
48	45.2	47	0.54

The model parameter θ is statistically significant ($p < 0.000$), $\hat{\theta} = -0.4341$ (note that Minitab is using Box and Jenkins' (1976) notation for MA(1), that is $\tilde{X}_t = Z_t - \theta Z_{t-1}$).

Hence, the MA(1) model is $\tilde{X}_t = Z_t - 0.4341 Z_{t-1}$, where \tilde{X}_t denotes the transformed and differenced series, Z_t denotes a white noise.

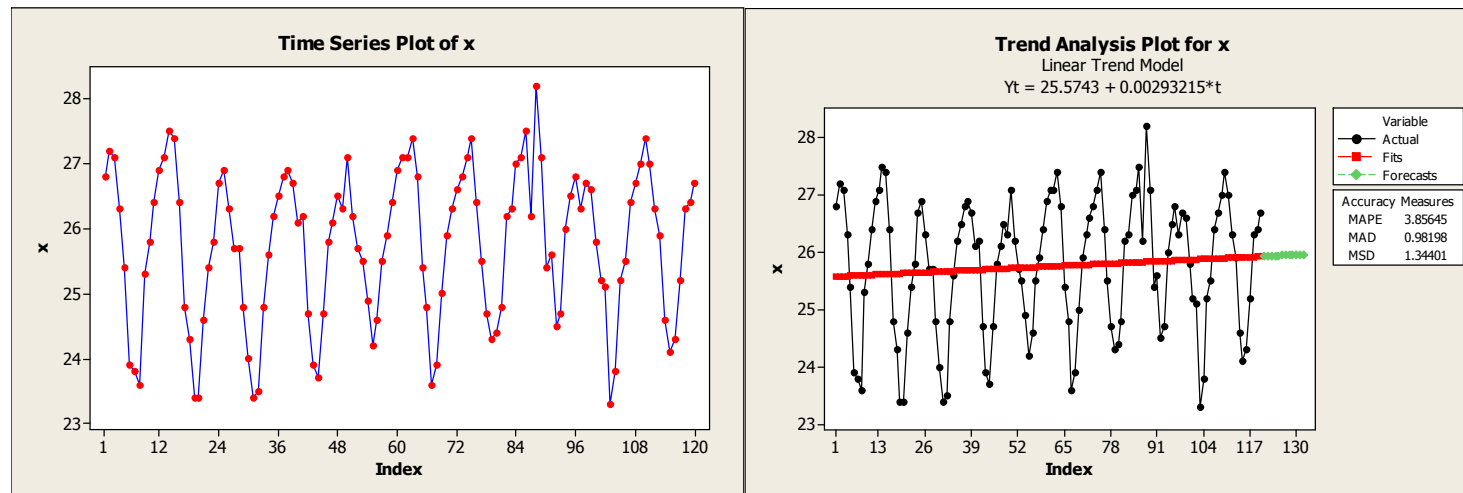
The residuals look like Gaussian White Noise.

Also, the Box-Pierce (Ljung-Box) Chi-Square statistics show that the residuals of this model (in groups of up to 48 values) are uncorrelated.

MA(1) fits well the transformed and differenced sunspots data.

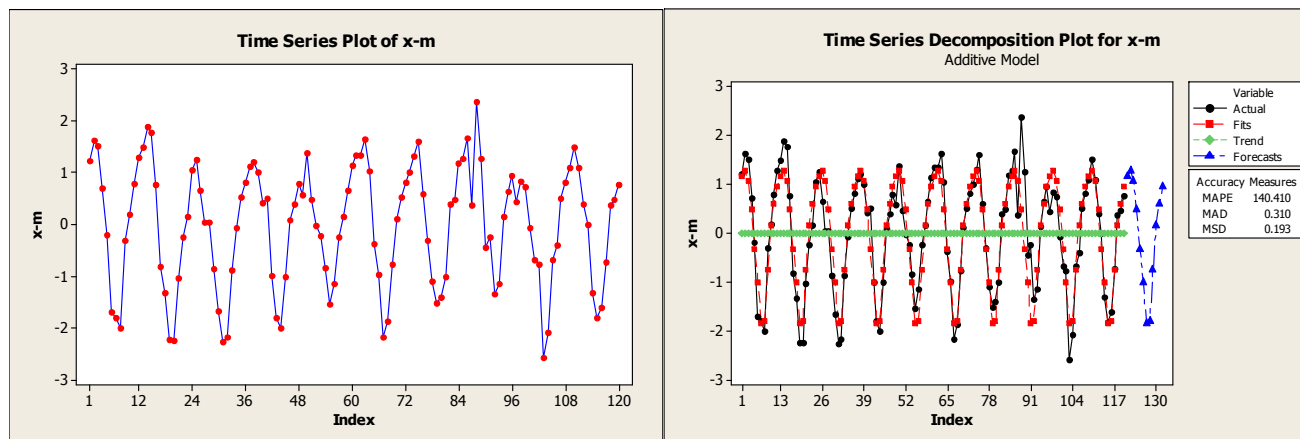
Another model, such as AR(3) might also be considered.

6.1.2 Recife Data

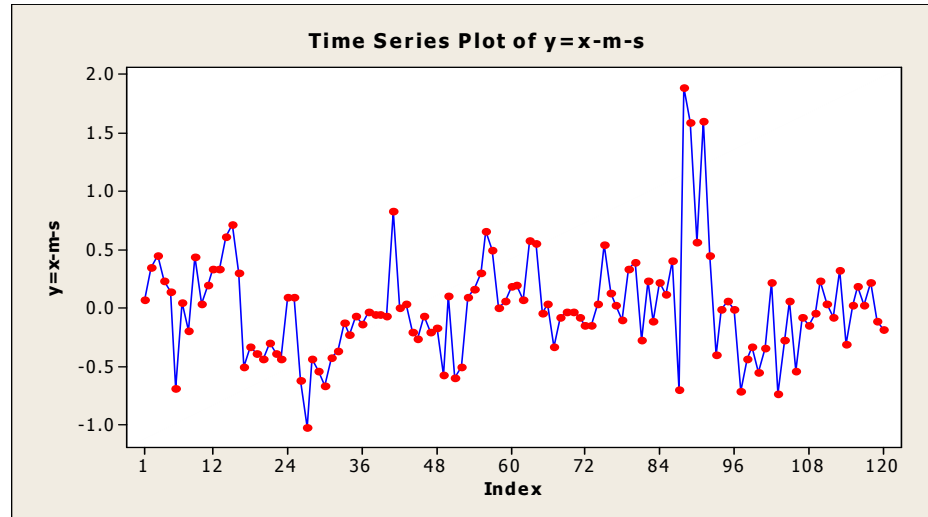


Monthly average temperature in Recife in Celsius in years 1953-62. The temperature varies between about 23 – 28 degrees C. The warmest months are December-March and least warm are June-August. There is clear seasonality in the data and not very clear, perhaps slightly upward, trend. Unusual pattern occurred in the eighth year of the recorded data.

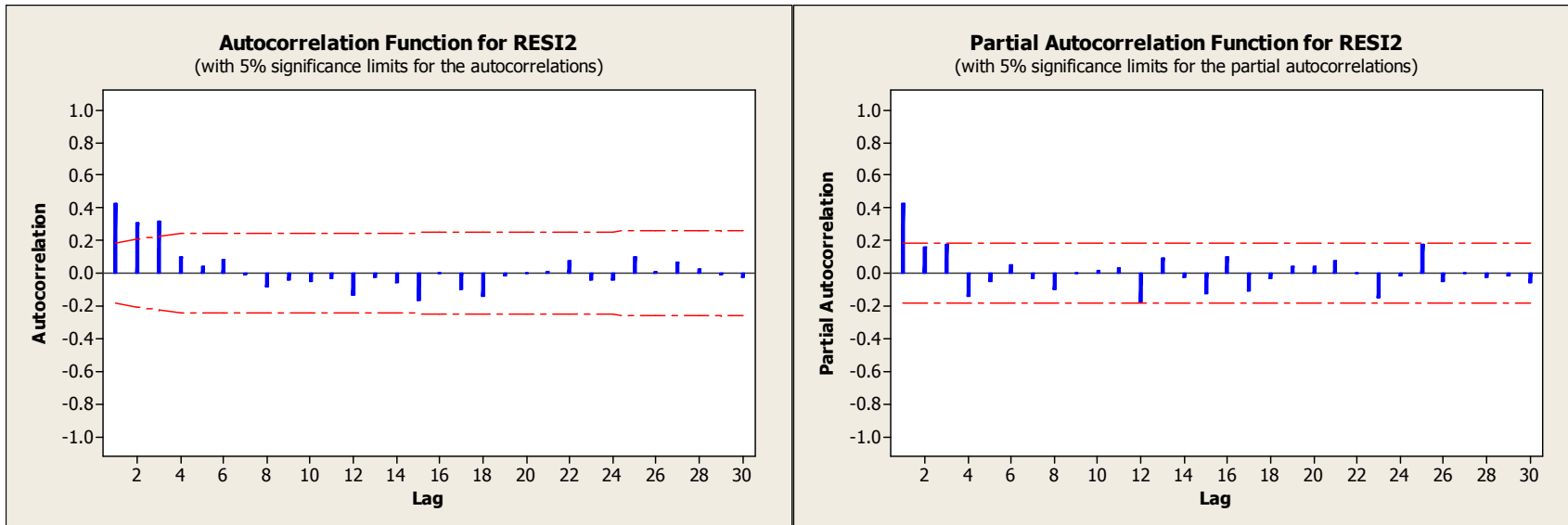
The linear trend fit shows a small increase in temperature over the years.



The detrended data exhibit seasonality but no trend, the 12 forecasted values repeat the pattern of the seasonal effects.



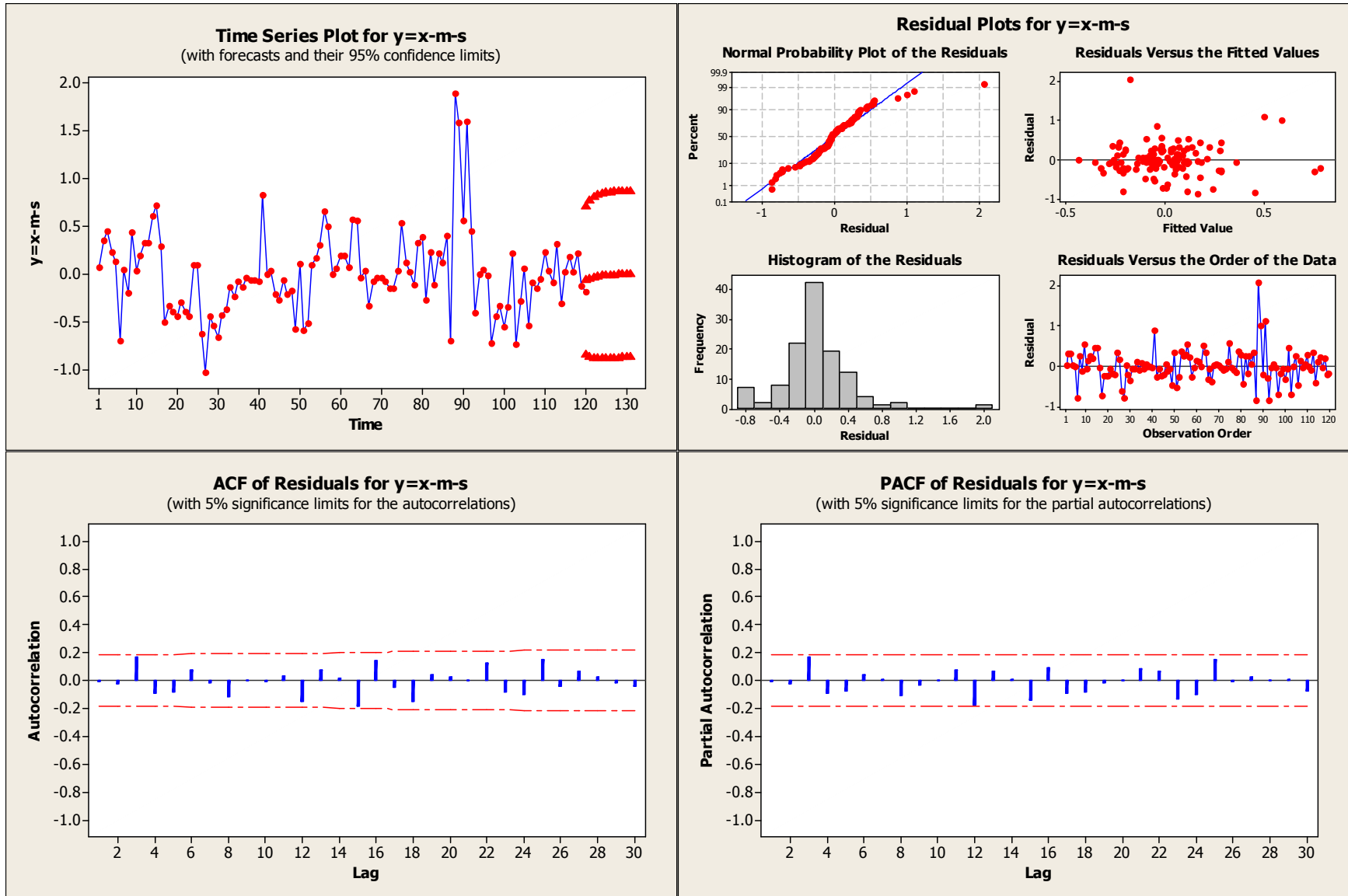
The seasonally adjusted and detrended data seem to be correlated with a few unusual values reflecting the pattern in the eighth year of the recorded data.



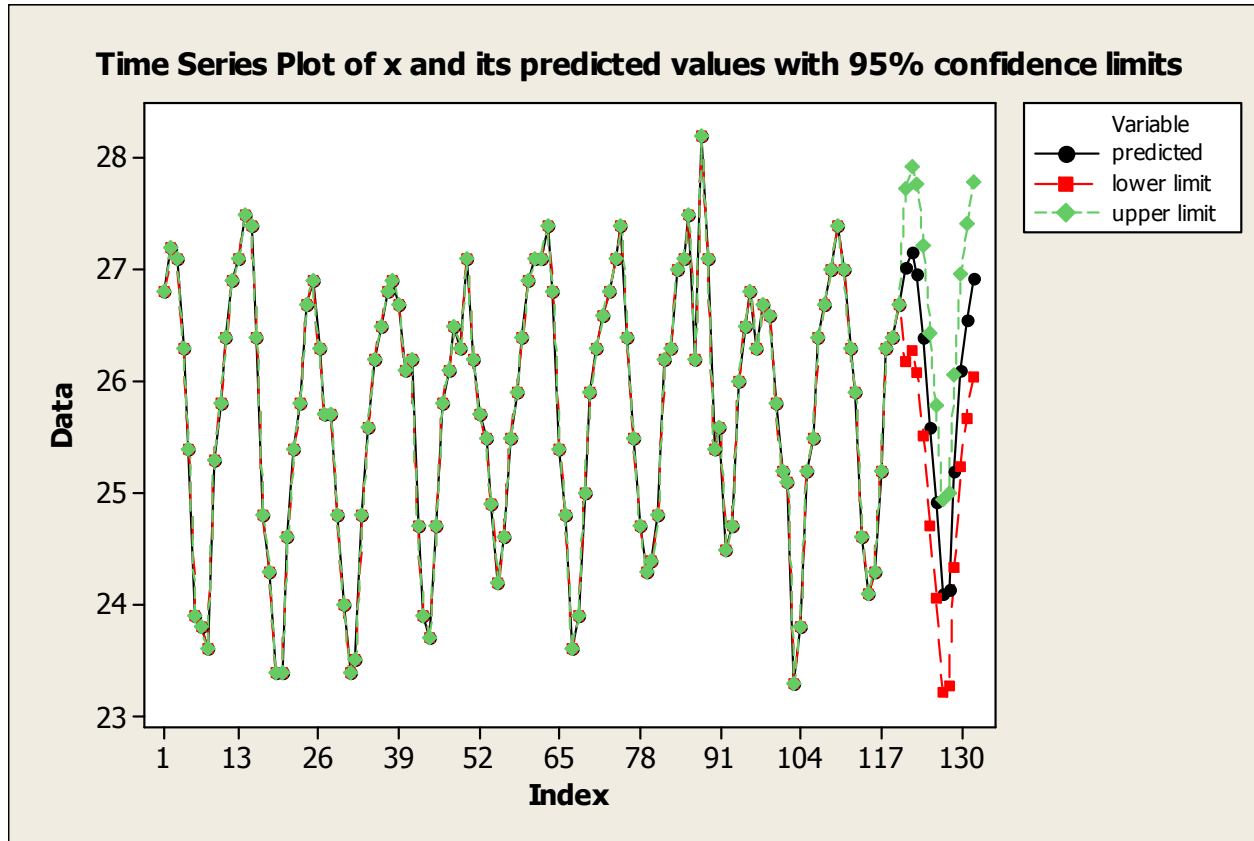
The sample ACF and PACF suggest an ARMA process. Trying various models we have obtained a reasonable fit with ARMA(1,1):

$$X_t - \phi X_{t-1} = Z_t + \vartheta Z_{t-1}, \quad Z_t \sim \text{WN}(0, \sigma^2).$$

Below are given plots of the detrended and deseasonalised data and their 12 forecasted values. Also, there are diagnostic plots for checking the white noise assumptions of the ARMA(1,1) model. As we can see the assumptions are approximately met.



Here is the plot of the original data and the 12 added forecasted values.



The numerical Minitab output:

Trend Analysis for x		Time Series Decomposition for x-m		ARIMA Model: y=x-m-s					
Fitted Trend Equation Yt = 25.5743 + 0.00293215*t		Additive Model		Type	Coef	SE Coef	T	P	
Period	Forecast	Period	Forecast	AR	1	0.7378	0.1278	5.77	0.000
121	25.9291	121	1.15694	MA	1	0.3911	0.1743	2.24	0.027
122	25.9320	122	1.27361	Modified Box-Pierce (Ljung-Box) Chi-Square statistic					
123	25.9349	123	1.06944	Lag		12	24	36	48
124	25.9379	124	0.48194	Chi-Square		11.1	28.2	35.9	56.5
125	25.9408	125	-0.32222	DF		10	22	34	46
126	25.9437	126	-0.99722	P-Value		0.353	0.171	0.378	0.139
127	25.9467	127	-1.83889	Forecasts from period 120					
128	25.9496	128	-1.79722	95 Percent Limits					
129	25.9525	129	-0.73889	Period	Forecast	Lower	Upper	Actual	
130	25.9555	130	0.15694	121	-0.067285	-0.840540	0.705970		
131	25.9584	131	0.59444	122	-0.049643	-0.868060	0.768775		
132	25.9613	132	0.96111	123	-0.036626	-0.878609	0.805357		
				124	-0.027022	-0.881561	0.827516		
				125	-0.019937	-0.881232	0.841358		
				126	-0.014709	-0.879661	0.850242		
				127	-0.010853	-0.877788	0.856083		
				128	-0.008007	-0.876020	0.860006		
				129	-0.005907	-0.874507	0.862692		
				130	-0.004358	-0.873277	0.864560		
				131	-0.003216	-0.872307	0.865876		
				132	-0.002372	-0.871559	0.866814		