# The assignment 2 report

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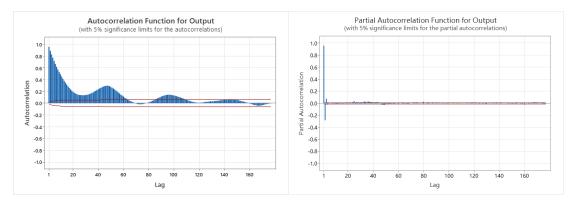
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## **Dataset of Clements Gap Wind Farm Output**

#### 1.1 Find the best ARMA(p, q) model for the 2011 data.

The first step is to check whether the dataset suit for ARMA model via ACF and PACF. The result like the pictures below.



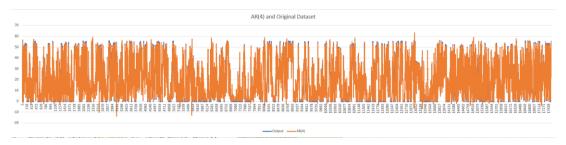
According to the graphs, it's reasonable to use ARMA model for this dataset. Next step is to find a proper ARMA(p, q) model for the dataset. There are five models that could be found for the dataset. The parameters like the picture right.

For selecting a proper model, we should compare the mean of squared error (MSE) and p-value of the parameters for each model, the smaller the better. The table below lists all the MSEs.

	ARMA(3,3)	ARMA(4,1)	AR(4)	AR(3)	AR(2)
mean of square error	overflow	85.89976922	24.75701538	24.76769416	24.92197907

According to the mean of square error, the performance of AR(3) and AR(4) are similar, I select the AR(4) as the best model. The details of AR(4) result like the picture below.

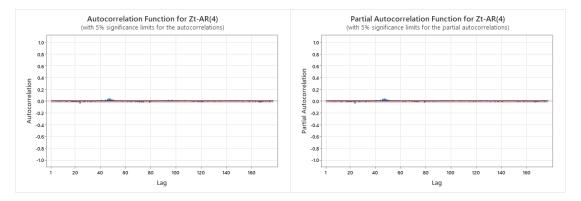
S	T	U	V	W							
	AR/A	RMA for O	utput								
Final Estin	nates of Pa	irameters									
Туре	Coef	SE Coef	T-Value	P-Value							
AR 1	-0.4246	0.0144	-29.55	0							
AR 2	0.5552	0.0225	24.64	0							
AR 3	0.6599	0.021	31.37	0							
MA 1	-1.6758	0.0118	-142.12	0							
MA 2	-1.155	0.0186	-62.26	0							
MA 3	-0.23619			0							
Constant	4.189	0.153	27.4	0							
Mean	19.999										
Final Estimates of Parameters											
Type	Coef	SE Coef	T-Value	P-Value							
AR 1	1.87	0.221		0							
AR 2	-1.159	0.275	-4.21	0							
AR 3	0.3351	0.0809	4.14	0							
AR 4	-0.0663	0.0157	-4.21	0							
MA 1	0.62	0.221									
Constant	0.4044		28.3	0							
Mean	20.004	0.707									
Final Estir	nates of Pa										
Type	Coef		T-Value	P-Value							
AR 1	1.24913	0.00753	165.83	0							
AR 2	-0.3786	0.0117									
AR 3	0.0787										
Constant	1.0162			0							
Mean	19.999	0.74									
Final Estimates of Parameters											
Final Estir		rameters									
Туре	Coef	rameters SE Coef	T-Value	P-Value							
Type AR 1	Coef 1.25076	SE Coef 0.00755	T-Value 165.57	0							
Type AR 1 AR 2	Coef 1.25076 -0.3865	SE Coef 0.00755 0.0121	T-Value 165.57 -32.01	0							
Type AR 1 AR 2 AR 3	1.25076 -0.3865 0.1046	SE Coef 0.00755 0.0121 0.0121	T-Value 165.57 -32.01 8.67	0							
Type AR 1 AR 2 AR 3 AR 4	Coef 1.25076 -0.3865 0.1046 -0.02077	SE Coef 0.00755 0.0121 0.0121 0.00756	T-Value 165.57 -32.01 8.67 -2.75	0 0 0.006							
AR 1 AR 2 AR 3 AR 4 Constant	1.25076 -0.3865 0.1046 -0.02077 1.0374	0.00755 0.0121 0.00756 0.0376	T-Value 165.57 -32.01 8.67 -2.75 27.59	0							
Type AR 1 AR 2 AR 3 AR 4 Constant Mean	Coef 1.25076 -0.3865 0.1046 -0.02077 1.0374 19.999	0.00755 0.0121 0.00756 0.0376 0.0376	T-Value 165.57 -32.01 8.67 -2.75 27.59	0 0 0.006							
Type AR 1 AR 2 AR 3 AR 4 Constant Mean Final Estir	1.25076 -0.3865 0.1046 -0.02077 1.0374 19.999 nates of Pa	0.00755 0.0121 0.0121 0.0125 0.0376 0.725 0.725	T-Value 165.57 -32.01 8.67 -2.75 27.59	0 0 0 0.006							
Type AR 1 AR 2 AR 3 AR 4 Constant Mean Final Estir	Coef 1.25076 -0.3865 0.1046 -0.02077 1.0374 19.999 mates of Pa Coef	0.00755 0.0121 0.0121 0.0125 0.0376 0.0376 0.725 rameters	T-Value 165.57 -32.01 8.67 -2.75 27.59	0 0 0.006 0							
Type AR 1 AR 2 AR 3 AR 4 Constant Mean Final Estir Type AR 1	Coef 1.25076 -0.3865 0.1046 -0.02077 1.0374 19.999 mates of Pa Coef 1.22693	0.00755 0.0121 0.0121 0.0125 0.0376 0.0376 0.725 rameters SE Coef 0.00725	T-Value 165.57 -32.01 8.67 -2.75 27.59 T-Value 169.26	0 0 0.006 0							
Type AR 1 AR 2 AR 3 AR 4 Constant Mean Final Estir Type AR 1 AR 2	Coef 1.25076 -0.3865 0.1046 -0.02077 1.0374 19.999 mates of Pa Coef 1.22693 -0.2821	SE Coef 0.00755 0.0121 0.0121 0.00756 0.0376 0.725 rameters SE Coef 0.00725 0.00725	T-Value 165.57 -32.01 8.67 -2.75 27.59 T-Value 169.26 -38.91	0 0 0.006 0 P-Value							
Type AR 1 AR 2 AR 3 AR 4 Constant Mean Final Estir Type AR 1	Coef 1.25076 -0.3865 0.1046 -0.02077 1.0374 19.999 mates of Pa Coef 1.22693	SE Coef 0.00755 0.0121 0.0121 0.00756 0.0376 0.725 rameters SE Coef 0.00725 0.00725	T-Value 165.57 -32.01 8.67 -2.75 27.59 T-Value 169.26 -38.91 29.25	0 0 0.006 0							



Visualization of AR(4) with original dataset

#### 1.2 Take the noise Zt from that model and check its SACF.

The residuals based on the model AR(4) could get, and name the residuals as Zt-AR(4), then, we could see the SACF according to the ACF and PACF, like the pictures below.

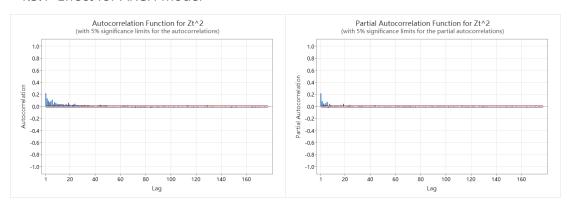


According to the result from ACF and PACF, the Zt-AR(4) is not suit for ARMA model.

#### 1.3 Calculate squared noise and show that it has the ARCH effect.

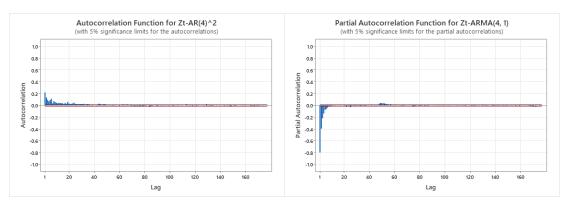
For calculating the **ARCH** effect, it should be separated into two parts, one for **ARCH** model, another for **GARCH** model.

#### 1.3.1 Effect for ARCH model



According to the SACF above, there is ARCH effect for ARCH model.

#### 1.3.2 Effect for GARCH model



According to the SACF above, there is also ARCH effect for GARCH model.

#### 1.4 Find the best ARCH or GARCH model for it.

For this part, I try to find all possible ARCH and GARCH models, and then try to compare the results for finalizing the model.

#### 1.4.1 For ARCH model

According to the squared residuals for ARCH model, five **AR** models for the dataset could be found, the parameters like the right picture. The coverage rate could be calculated like the picture below based on the parameters of ARCH models.

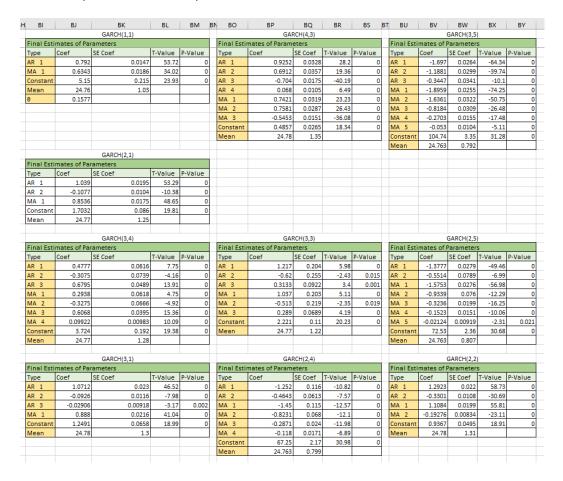
AG	AH	Al
ARCH(5)	Bounds Coefficient	1.96
ARCH(3)	Coverage Rate	0.89578
ARCH(4)	Bounds Coefficient	1.96
ARCH(4)	Coverage Rate	0.895786
ARCH(3)	Bounds Coefficient	1.96
Anch(5)	Coverage Rate	0.895792
ARCH(2)	Bounds Coefficient	1.96
ARCH(2)	Coverage Rate	0.895798
ARCH(1)	Bounds Coefficient	1.96
ARCH(I)	Coverage Rate	0.895804

According to the result, all the coverage rates for the five model are similar, and they all approach 89.6% for score 1.96, which is much less than 95%.

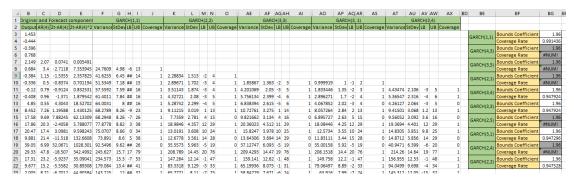
ARCH(5)										
Final Estimates of Parameters										
Туре	Coef	SE Coef	T-Value	P-Value						
AR 1	0.19027	0.00754	25.22	0						
AR 2	0.07629	0.00768	9.94	0						
AR 3	0.04172	0.00769	5.42	0						
AR 4	0.0215	0.00768	2.8	0.005						
AR 5	0.05725	0.00754	7.59	0						
Constant	15.178	0.589	25.79	0						
Mean	24.761	0.96								
ARCH(4)										
Final Estimates of Parameters										
Туре	Coef	SE Coef	T-Value	P-Value						
AR 1	0.19213	0.00755	25.44	0						
AR 2	0.07894	0.00768	10.27	0						
AR 3	0.04624	0.00768	6.02	0						
AR 4	0.0325	0.00755	4.3	0						
Constant	16.1	0.59	27.31	0						
Mean	24.762	0.907								
ARCH(3)										
Final Estin	nates of Pa	rameters	5							
Туре	Coef	SE Coef	T-Value	P-Value						
AR 1	0.19384	0.00755	25.69	0						
AR 2	0.08159	0.00766	10.65	0						
AR 3	0.05253	0.00755	6.96	0						
Constant	16.641	0.59	28.22	0						
Mean	24.762	0.878								
ARCH(2)										
Final Estin	nates of Pa									
Туре	Coef	SE Coef	T-Value	P-Value						
AR 1	0.19867	0.00752	26.4	0						
AR 2	0.09202	0.00752	12.23	0						
Constant	17.564	0.591	29.74	0						
Mean	24.763	0.833								
ARCH(1)										
Final Estin	nates of Pa	rameters	5							
Туре	Coef	SE Coef	T-Value	P-Value						
AR 1	0.21881	0.00737	29.68	0						
Constant	19.345	0.593	32.62	0						
Mean	24.763	0.759								

#### 1.4.2 For GARCH model

According to the squared residuals for GARCH model, 10 **ARMA** model for the dataset could be found, the parameters like the picture below.



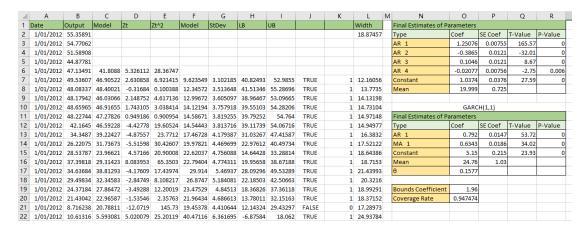
For some models will occurs negative values, which will lead to the specified model unavailable. The coverage rate for each model like the picture below.



According to the result the best model for the residuals is *GARCH(1,1)*, the coverage could be 99.15%.

#### 1.5 Apply all models to the 2012 output data.

According to previous steps, we got two models for the dataset. One is AR(4), another one is GARCH(1,1). All the models will apply to 2012 dataset. The result likes the picture below.

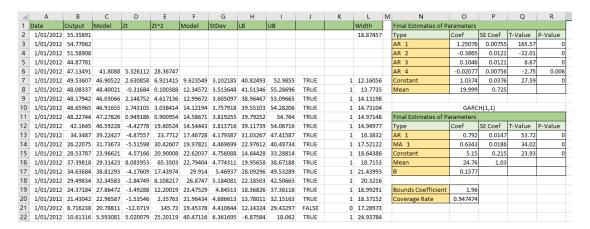


According to the result of coverage also approaches 95% that means quite well.

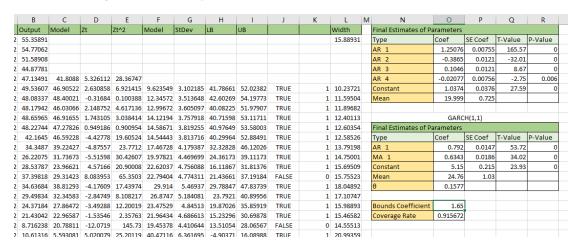
#### 1.6 Evaluate the performance.

The score of 90% is about 1.65 and the score of 95% is about 1.96. The 95% result like the picture below.

The 95% coverage result like the picture below.



The 90% coverage result like the picture below.



The statistical summary for the residuals of 2012 data likes the right picture. According to the statistical summary, the standard deviation is about 5.05. The mean prediction interval width of 95% coverage is about 18.87, for 90% coverage is about 15.89. The real coverage with 1.96 score is 94.75%, and 91.57% for 1.65 score. So we could get the formulas like below:

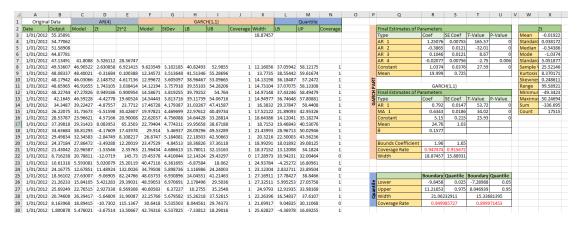
94.75% ≈ 95% 91.57% > 90% 5.05\*1.96\*2=19.8>18.87 5.05\*1.65\*2=16.665>15.89

The results suggest that the model is quiet well for the dataset.

#### Τ U Mean -0.019220.038172 Standard -0.54386 Median -1.0374Mode Standard 5.051877 Sample 25.52146 Kurtosis 8.370171 Skewnes 0.248611 99.58921 Range Minimur -49.3423 Maximui 50.24694 -336.695 17515 Count

#### 1.7 Compare the results.

The picture below shows two types of error bounds, the GARCH(1,1) and Quantile.



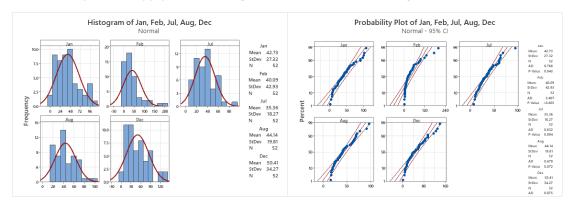
The right picture focusses on the results of the two approaches. According to the results the performance of two approaches is similar.

	Bounds Coefficient	1.96	1.65		
	Coverage Rate	0.947474	0.915672		
	Width	18.87457	15.88931		
		Boundary	Quantile	Boundary	Quantile
ë	Lower	-9.8458	0.025	-7.28988	0.05
antile	Upper	11.21653 0.975		8.046939	0.95
듄					
Quan	Width	21.062	32911	15.336	81395
	Width Coverage Rate	21.062 0.9499			81395 971453

#### **Dataset of Melbourne Airport Rain**

### 2.1 Test normality for December, January, February, July, August

For normality test, the pp-plot and histogram could be used for testing.



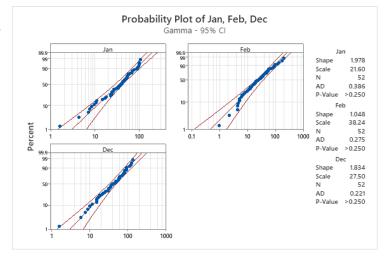
According to the histogram all the distribution of the months are right skewed. According to the pp-plot result, we reject January, February and December datasets follow normal distribution because of p-value is less than 0.05.

#### 2.2 Test for Gamma fit.

There are two steps for this question. The first step is to test the gamma distribution, and then to calculate the  $\alpha$  and  $\beta$  parameters. Another step is to get the distribution and visualize them. According to previous step, the datasets of January, February and December will be processed.

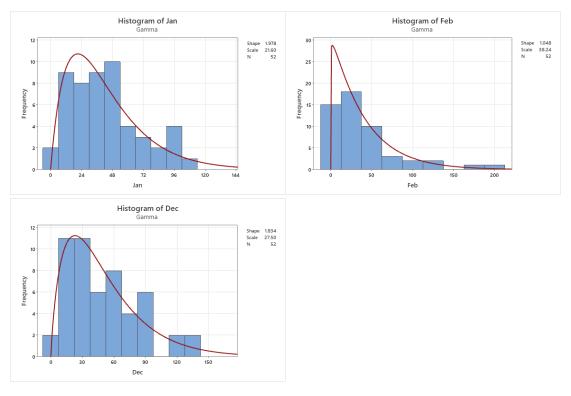
#### 2.2.1 Gamma distribution test

According to the right graph, the p-values are all greater than 0.05, it means we cannot reject the datasets follow gamma distribution. And the parameters of gamma distribution for each dataset are also listed on the graph.



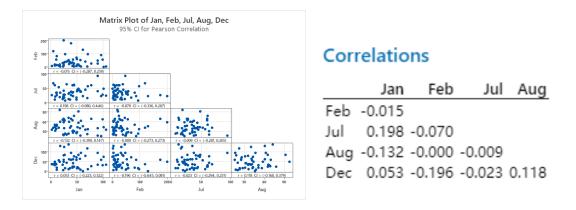
#### 2.2.2 Visualize the distribution.

This part will show the gamma distribution for each month.



#### 2.3 Test the correlation.

For the correlation test, the Pearson correlation is used here. The Matrix Plot and correlation matrix like the pictures below.

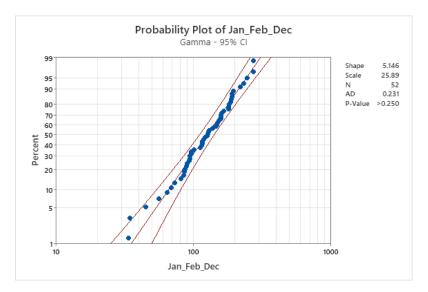


According to the correlation matrix, the data of Jan with July have the highest correlation, even the highest correlation is very week, so we regard there is no correlation among the datasets.

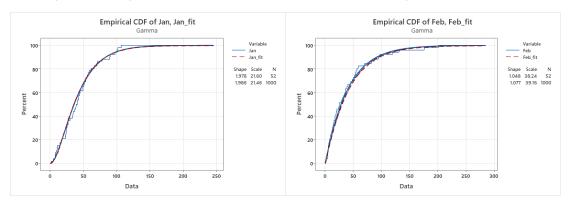
# 2.4 Synthetic and empirical CDFs for January, February, December and total.

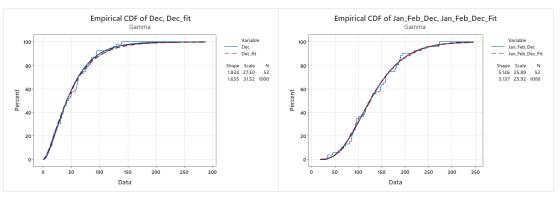
The first step is to get the total data then to test gamma distribution for it. The second step is to generate synthetic data for the datasets that follow gamma distribution, then to do empirical CDFs for them.

Make sum of January, February, and December to form the total data, then to do gamma test like the picture below.



According to this graph above, it can not to reject the total data follow gamma distribution. So, next step is to do empirical CDFs for the four datasets, details like the pictures below.

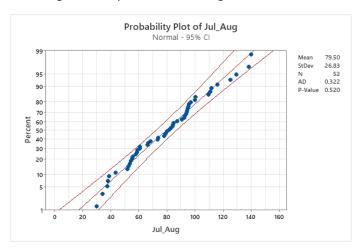


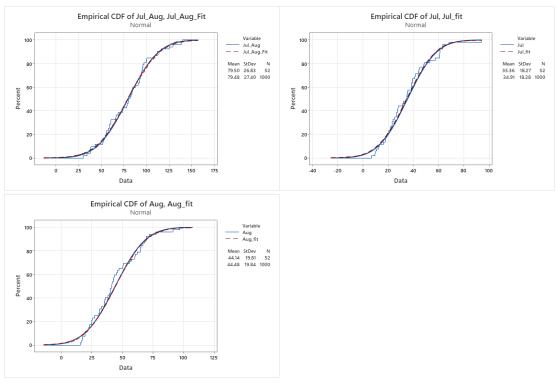


#### 2.5 Synthetic and CDFs for July, August.

The process for this question is the same with chapter 2.4. The difference is the datasets of July and August follow normal distribution. When generate Synthetic data using normal distribution.

Test the total value of July and August, then to test the normality for it, details like the picture below. According to the result, we cannot reject the total data of July and August follows normal distribution because of the p-value is greater than 0.05. The next step is to do empirical CDFs for the three datasets, details like the pictures below.





Empirical CDF for July, August, and Sum of them

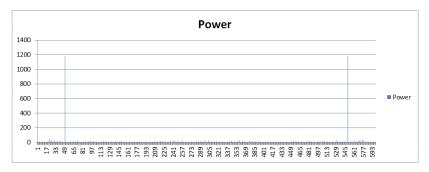
#### Datasets of Mt Gambier By Months Temperature and Mt Gambier

#### Rainfall

#### 3.1 Model the seasonality and then get residuals.

There are three steps for this question. The first step is to find the best frequencies, the second step is to find the proper parameters for seasonality, and the last step is to visualize the seasonality result.

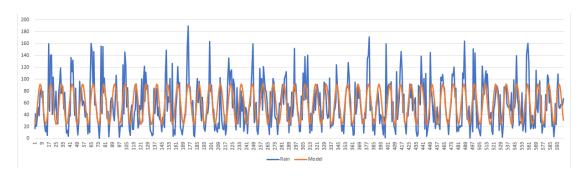
The picture right is the DFT result for the dataset. The 50 and 550 is the best for the dataset.



The right picture is the seasonality parameters using the frequencies got from step1. And also we got the final model and the residuals.

	Α	В	С	D	E	F	G	Н	1	J
1					1/year	11/year		n=	600	
2				Frequencies	0.523598776	5.759587		SME	520408.3	
3	Mean	57.868		alpha	-13.568258	-13.5683				
4	Variance	38.18741		beta	-10.495358	10.49536				
5										
6										
_										
7	Voor	Month	Time	Pain	Frequer	ncies	Model	DiffA2	Posiduals	
_	Year	Month	Time	Rain	Frequer 50	ncies 550	Model	Diff^2	Residuals	
7	Year 1950	Month 1	Time 1	Rain 17.1	50		Model 23.87173	Diff^2 45.85632		
7		Month 1	Time 1 2		50 -16.9981352	550	23.87173		-6.77173	
7 8 9	1950	Month 1 2 3	1	17.1	-16.9981352 -15.8733757	550 -16.9981	23.87173 26.12125	45.85632 248.969	-6.77173 15.77875	
7 8 9 10	1950 1950	1 2	1 2	17.1 41.9	50 -16.9981352 -15.8733757 -10.495358	550 -16.9981 -15.8734 -10.4954	23.87173 26.12125	45.85632 248.969 268.2154	-6.77173 15.77875 -16.3773	

The picture below is the visualization for the final model of seasonality.

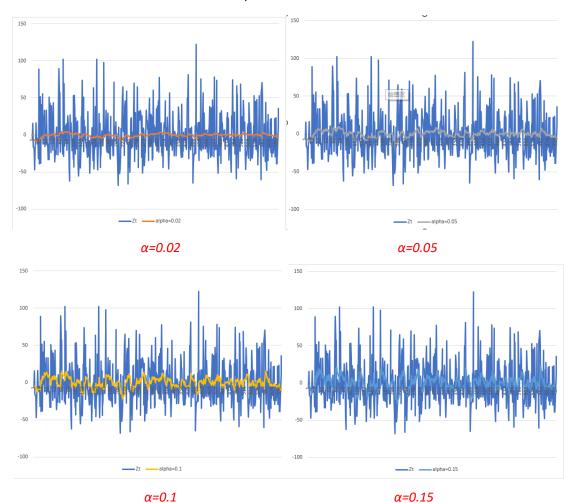


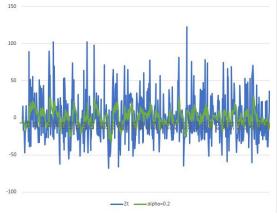
## 3.2 Use exponential smoothing to see the overall trend

For this question, I will the dataset calculated using exponential smoothing with various values, then to visualize them. The datasets get from exponential smoothing with different values like the pictures below.

	Α	В	С	D	E	F	G	Н	1	J	K	L	M	N	0	P	
1					1/year	11/year		n=	600						Exponent	tial Smoothing	
2				Frequencies	0.523598776	5.759587		SME	520408.3						alpha	0.05	
3	Mean	57.868		alpha	-13.568258	-13.5683											
4	Variance	38.18741		beta	-10.495358	10.49536											
5																	
6	Year	Month	Time	Pain	Freque	ncies	Seasonality	Diff^2	Residuals	E	xponen	tial Sm	oothin	g			
7	Tear	WIOTIET	IIIIIe	Italii	50	550	Model	DIII 2	Nesiduais	0.02	0.05	0.1	0.15	0.2			
8	1950	1	1	17.1	-16.9981352	-16.9981	23.87172968	45.85632	-6.77173								
9	1950	2	2	41.9	-15.8733757	-15.8734	26.1212486	248.969	15.77875	-6.77	-6.77	-6.77	-6.77	-6.77			
10	1950	3	3	20.5	-10.495358	-10.4954	36.87728392	268.2154	-16.3773	-6.32	-5.64	-4.52	-3.39	-2.26			
11	1950	4	4	35.7	-2.30511767	-2.30512	53.25776466	308.2751	-17.5578	-6.52	-6.18	-5.7	-5.34	-5.08			
12	1950	5	5	52.3	6.502777119	6.502777	70.87355424	344.9769	-18.5736	-6.74	-6.75	-6.89	-7.17	-7.58			
13	1950	6	6	38.2	13.56825803	13.56826	85.00451606	2190.663	-46.8045	-6.98	-7.34	-8.06	-8.88	-9.78			
14	1950	7	7	74.1	16.99813516	16.99814	91.86427032	315.5693	-17.7643	-7.78	-9.31	-11.9	-14.6	-17.2			
15	1950	8	8	83.6	15.8733757	15.87338	89.6147514	36.17723	-6.01475	-7.98	-9.74	-12.5	-15	-17.3			
16	1950	9	9	69.6	10.49535804	10.49536	78.85871608	85.72382	-9.25872	-7.94	-9.55	-11.9	-13.7	-15			
17	1950	10	10	78.9	2.305117668	2.305118	62.47823534	269.6744	16.42176	-7.96	-9.54	-11.6	-13	-13.9			
18	1950	11	11	38.4	-6.50277712	-6.50278	44.86244576	41.76321	-6.46245	-7.48	-8.24	-8.8	-8.61	-7.82			
19	1950	12	12	24.1	-13.568258	-13.5683	30.73148394	43.97658	-6.63148	-7.45	-8.15	-8.57	-8.29	-7.55			
20	1951	13	13	11.7	-16.9981352	-16.9981	23.87172968	148.151	-12.1717	-7.44	-8.07	-8.37	-8.04	-7.37			
21	1951	1/1	1/1	21 /	-15 9722757	-15 973/	26 1212/186	22 29019	_A 70105	-7 52	-8 28	-2 75	-8 66	-8 33			

According to the picture above, five different values of alpha are used to calculate. Next step will show the overall trend based on the five parameters.





*α=0.2* 

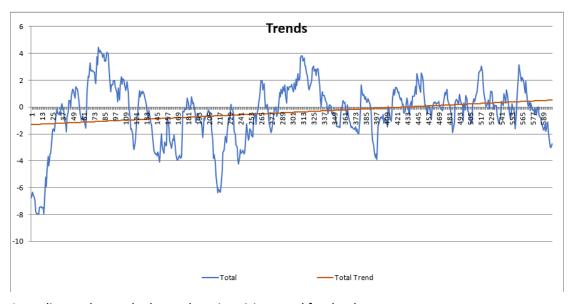
According to the graphs, it is much easier to extract the trend when the  $\alpha$ =0.02. For the first several data, the trend is rising, the following data has no obvious changing.

# 3.3 Find the trends for the smoothed data of whole series and various sections.

I will set the parameter  $\alpha$  equals 0.02 of smoothed data, and then to process the smoothed data. There are 3 steps to do. The first step is to find the trend of the whole dataset. The second step is to split the dataset into multiple sections, and the last step is to find the trends for each section.

#### 3.3.1 Find the trend of whole dataset.

We could find the trend from the smoothed data and linear regression. The whole dataset is rising based the smoothed dataset got from the previous step. The next step is using linear regression approach to get, I use the univariate linear regression to model the trend. The result like the picture below.



According to the result above, there is a rising trend for the dataset.

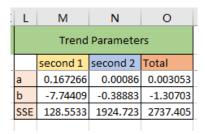
#### 3.3.2 Split whole dataset into multiple sections.

According the visualization of the dataset, the dataset could be split into two sections, the first section (from the beginning to 60) rise rapidly, and the second section oscillate around a variable. So, the dataset could be split into two sections like the picture below.

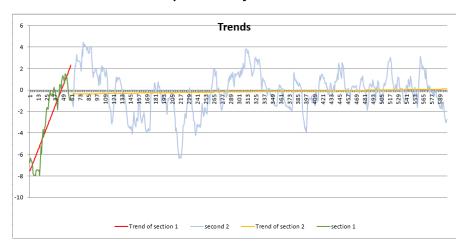
4	Α	В	С	D	E	F	G	Н	1	J	K	
1	Time		Exponential Trends					squared error				
2		Total	second1	second 2	second 1	second 2	Total	second1	second2	Total		
60	58	-0.959194581	-0.95919458		1.957334		-1.12994	8.506138		0.029154		
61	59	-0.939259604	-0.9392596		2.1246		-1.12689	9.387234		0.035204		
62	60	-0.709104091	-0.70910409		2.291866		-1.12383	9.005819		0.172001		
63	61	-1.020356603		-1.02036		-0.37164	-1.12078		0.420833	0.010085		
64	62	-1.076374443		-1.07637		-0.37078	-1.11773		0.497863	0.00171		
65	63	-1.586392632		-1.58639		-0.36992	-1.11467		1.479803	0.222518		
66	64	-0.707820073		-0.70782		-0.36906	-1.11162		0.114757	0.163055		
67	65	1.094865244		1.094865		-0.3682	-1.10857		2.140566	4.855117		
68	66	2.316877618		2.316878		-0.36734	-1.10551		7.205038	11.71277		

#### 3.3.3 Find the trends for the two sections

According to the smoothed data itself, we could find the first section has a rapid rising, and the second section data oscillate up and down around 0. Next step we could use univariate linear regression to model the trend, details like the picture below.



The trends parameters for the two sections



The trends for the two sections

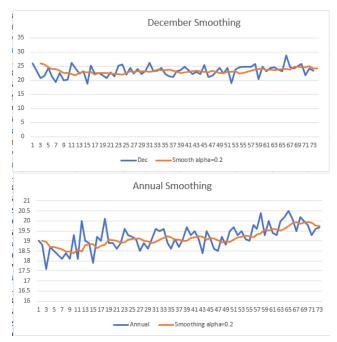
According to the results above, the two sections both have a rising trend, the first section has a rapid rising trend, the second section only has a slight rising trend that could almost be ignored.

# 3.4 Take the data for the month of December and the Annual mean temperature from MtGambierByMonthsTemperature.xlsx and find trend over time.

For finding trending there are two ways to do, the first way is using smoothing method, another approach is using linear regression. I will try to use the two approaches.

#### 3.4.1 Smoothing method

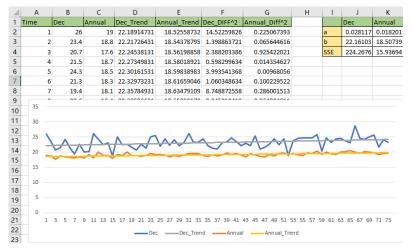
I set the alpha = 0.02 to smoothing the dataset like the picture right. According to the graph, the temperature of December does not have too much change, on the other hand, the temperature of Annual temperature has an obvious rise.



#### 3.4.2 Linear regression

The univariate linear regression to model trends for the two datasets. The results like the picture right.

According to the picture, the temperature of Annual and December are both rise, but the upward trend of Annual is more obvious.



#### 3.5 How much has the mean temperature changed over time in each case?

According to the smoothing approach, the mean temperature changed over time is about rise from 18.5 to 20 that approaches 1.5.

According to the linear regression result from the last step, the mean temperature changed over time should be calculated via the linear regression parameters with the whole 73 years on this dataset. The mean temperature changed over time for should be, a\*year =  $0.018201 * 73 \approx 1.32$ .