

Time Series Test Questions

Given the data below answer the following questions:

x_1	6
x_2	8
x_3	13
x_4	12
x_5	10
x_6	7
x_7	4
x_8	2

a. Calculate $\hat{\gamma}_0$

$$x = c(6,8,13,12,10,7,4,2)$$
$$\text{var}(x) * 7/8$$

c. Calculate $\hat{\rho}_0$

$$\hat{\rho}_0 = 1$$

b. Calculate $\hat{\rho}_1$ and how many pairs were used to find this estimate?

$$\text{plotts.sample.wge}(x)$$
$$\hat{\rho}_1 = .546$$

7 pairs

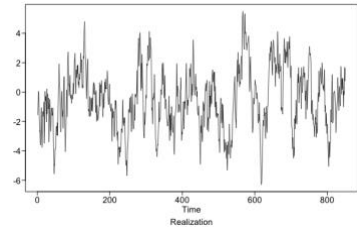
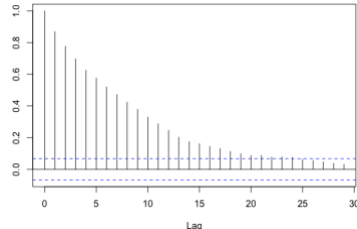
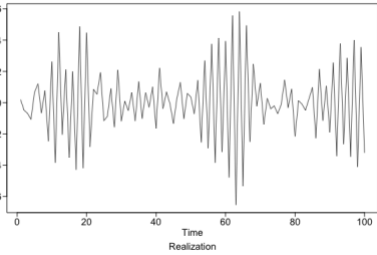
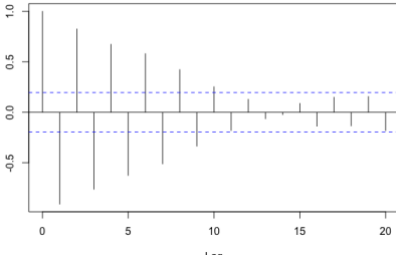
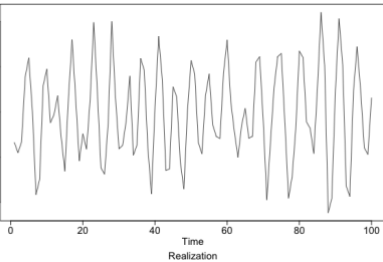
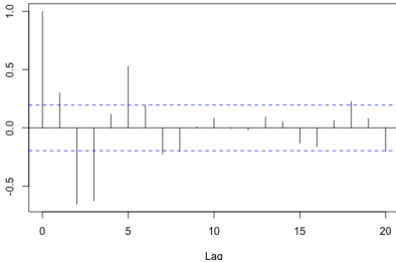
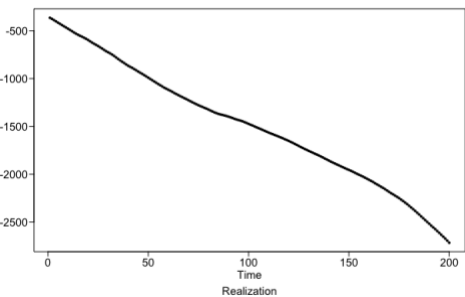
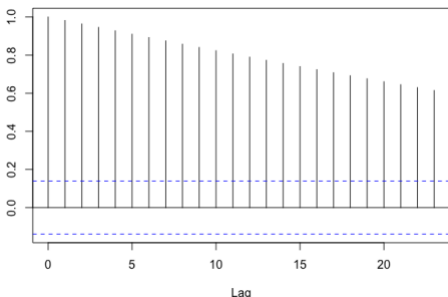
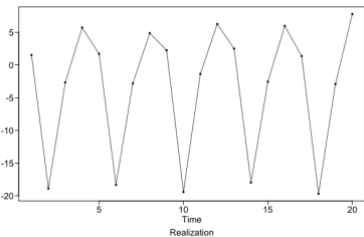
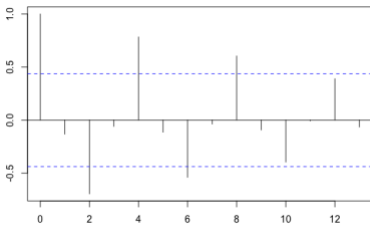
c. Which pairs would be used to calculate $\hat{\rho}_6$? (2,8) and (4,6)

f. Given the model: $(1 - .5B)X_t = a_t$ Calculate ("by hand" and show the steps) $X_{8+1} = X_9$

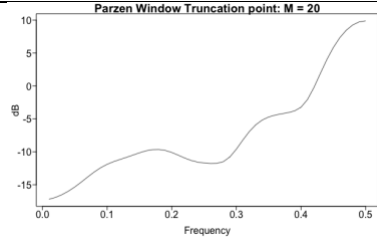
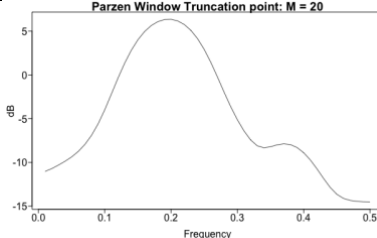
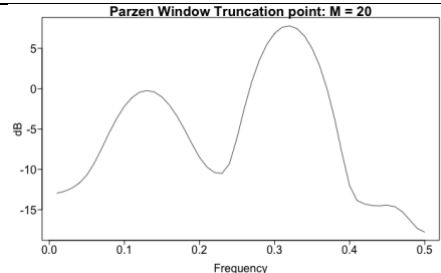
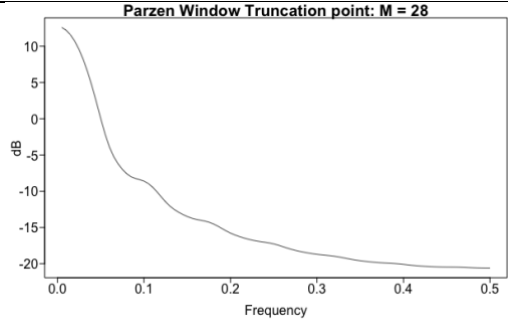
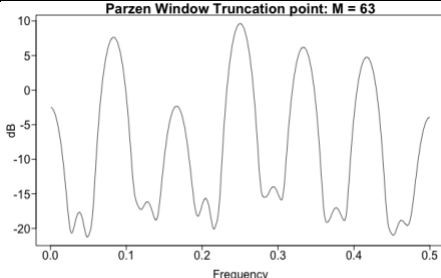
$$X_9 = .5(2) + (1 - .5)(7.75) = 4.875$$

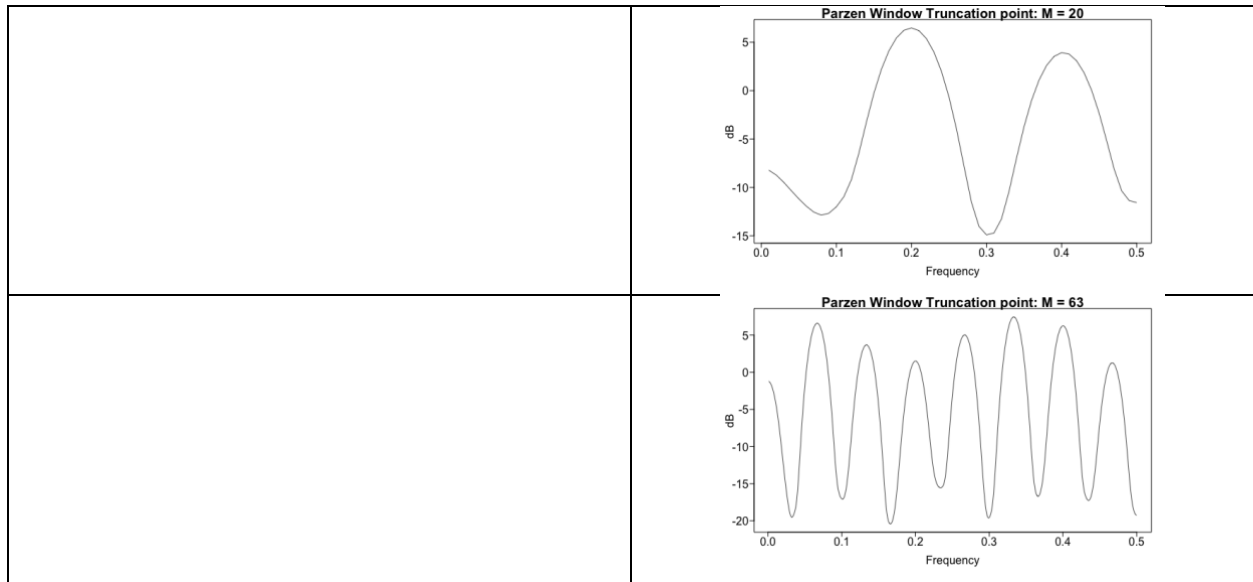
Check: `fore.arma.wge(x,phi = .5, n.ahead = 2)`

Match the Realization with the ACF:

Realization	ACF
	
	
	
	
	

Match each model on the left with a Spectral Density on the right ... each spectral density will be used only once and process of elimination may need to be used.

AR(1) $\rho_1 = -0.95$	
AR(2)	
AR(4)	
ARIMA (1,2,1)	
ARIMA (0,0,0) with $s = 12$	



Multiple Choice:

What type of filter is the 10-point moving average filter?

- A. Low Pass
- B. High Pass
- C. Neither
- D. Could be either

What type of filter is the first difference filter?

- A. Low Pass
- B. High Pass
- C. Neither
- D. Could be either

What type of filter is the Butterworth filter?

- A. Low Pass
- B. High Pass
- C. Neither
- D. Could be either

Which is not true about an MA(q) process?

- a. It creates dips in the autocorrelation function.
- b. It is already in GLP form.
- c. It's autocorrelations (ρ_k) are 0 for $k > q$.
- d. It is always stationary.
- e. They are invertible if the roots are outside the unit circle.

Factor Tables

Forecasts

What type of models will oscillate above and below the sample mean and eventually converge to the sample mean?

- A. AR(1) positive ϕ
- B. AR(1) negative ϕ
- C. AR(2) complex conjugate roots.
- d. AR(4) with two sets of complex conjugate roots
- e. airline models
- f. ARIMA(0,1,0) models
- g. signal + noise models

What type of models will simply repeat forecast X_{t+1} to be X_t ?

- A. AR(1) positive ϕ
- B. AR(1) negative ϕ
- C. AR(2) complex conjugate roots.
- d. AR(4) with two sets of complex conjugate roots
- e. airline models
- f. ARIMA(0,1,0) models
- g. signal + noise models

Assuming the Sunspot data is stationary (Base R dataset: sunspot.year), what is the model ID (ARMA(p,q)) that is most favored by the AIC?

```
> aic5.wge(sunspot.year)
-----WORKING... PLEASE WAIT...
```

```
Five Smallest Values of aic
  p  q  aic
15  4  2  5.621074
18  5  2  5.622962
11  3  1  5.623829
10  3  0  5.624369
 8  2  1  5.626398
```

Which model do you think is most appropriate/useful for forecasting the Sunspot data?

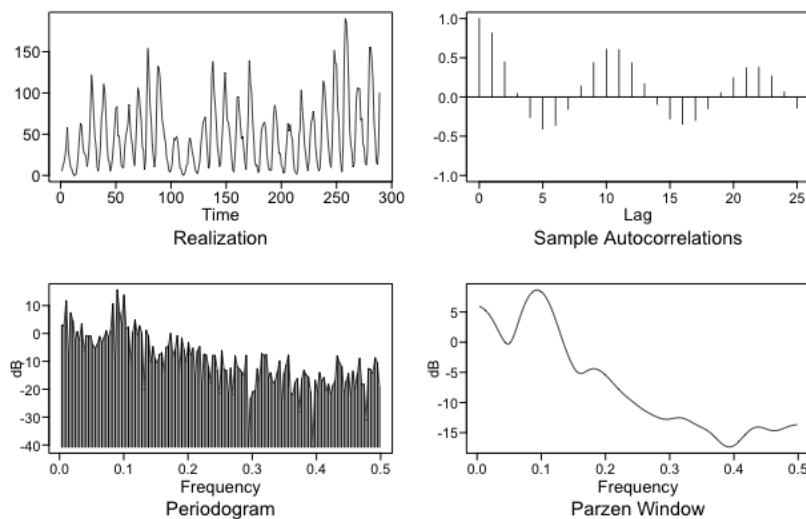
Provide at least 2 arguments as to why the model you selected is more useful than the other two in predicting the next 10 years of sunspots.

$$(1 - .723B - .283B^2 + .519B^3)X_t = (1 + .60B)a_t$$

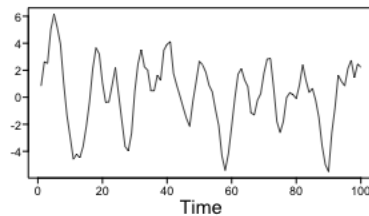
$$(1 - B^{12})(1 - B)X_t = a_t$$

$$(1 - 1.06B + .4B^2)(1 - B^{10})X_t = a_t$$

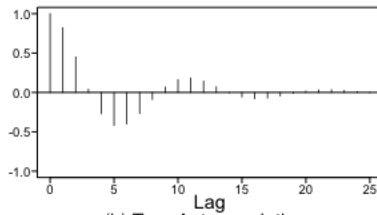
Sunspot Data:



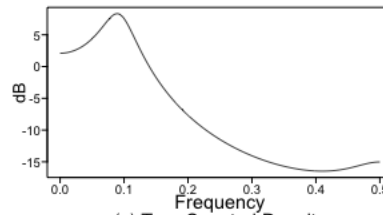
$$(1 - .723B - .283B^2 + .519B^3)X_t = (1 + .60B)a_t$$



(a) Realization

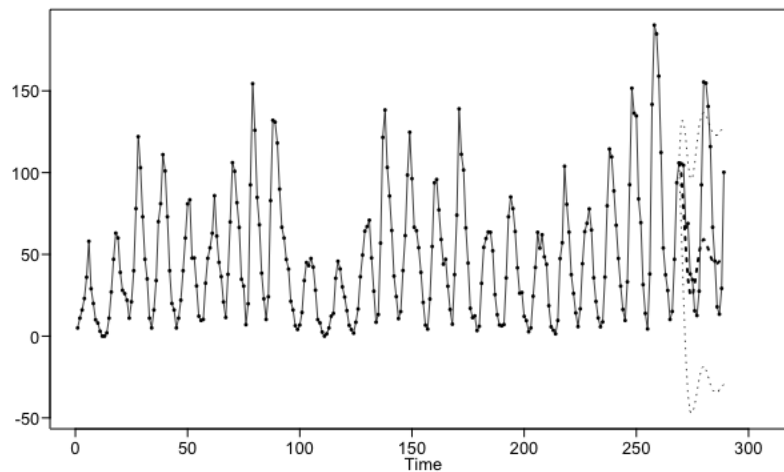


(b) True Autocorrelations



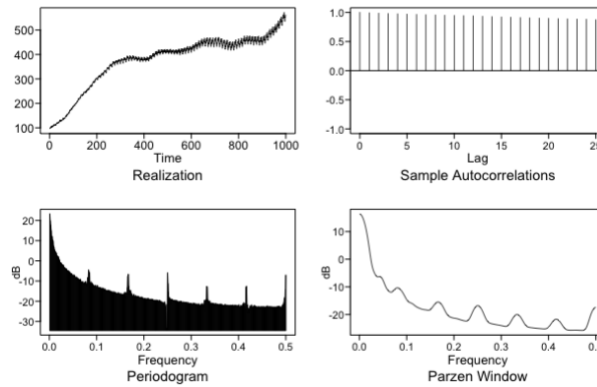
(c) True Spectral Density

plotts.true.wge(100,phi = c(0.723, 0.283, -0.519), theta = -.6)

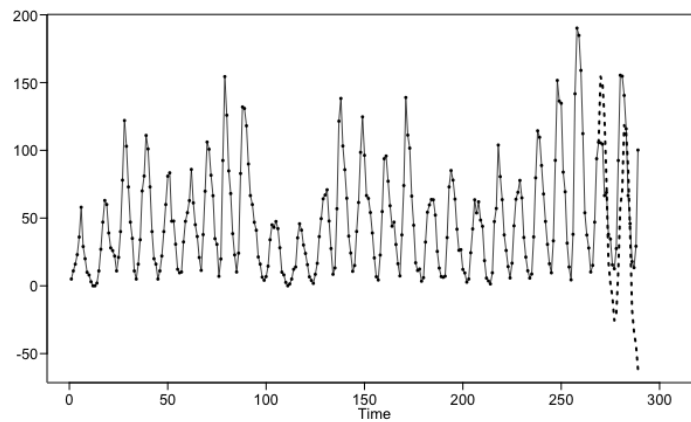


```
f.arma3_1 = fore.aruma.wge(sunspot.year,phi = c( 0.723, 0.283, -0.519), theta = -.6, n.ahead =
20, lastn = TRUE ) #Poor Visual Forecasts
aseARMA3_1 = sum(f.arma3_1$f - sunspot.year[(length(sunspot.year)-
19):length(sunspot.year)])^2
> aseARMA3_1
[1] 168439.9
```

$$(1 - B^{12})(1 - B)X_t = a_t$$

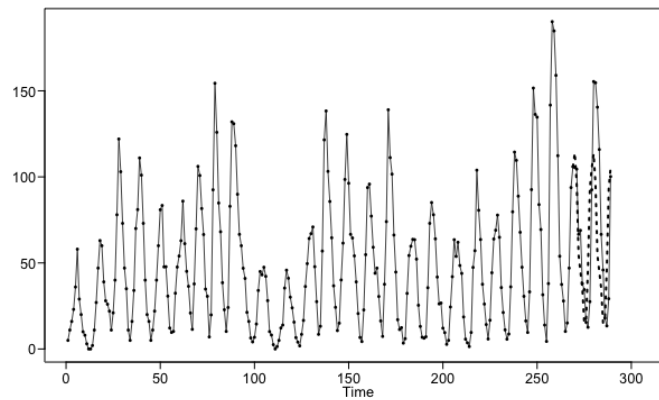


```
x = gen.aruma.wge(1000,d= 1, s= 12)
plots.sample.wge(x)
```

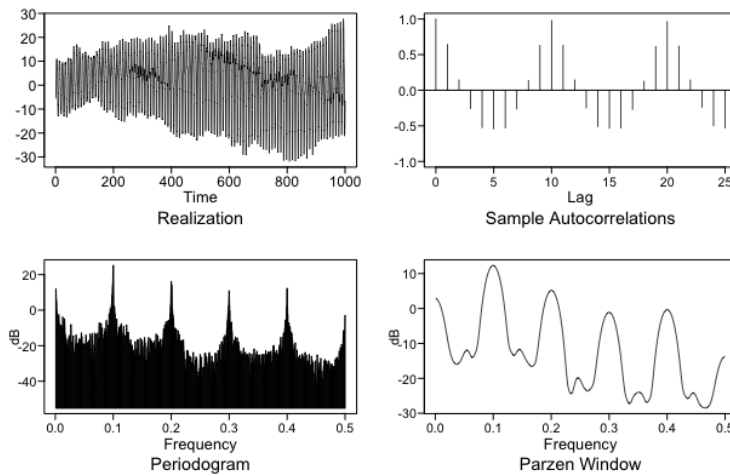


```
f.12.1 = fore.aruma.wge(sunspot.year,d= 1, s= 12, n.ahead = 20, lastn = TRUE, limits = FALSE )
ase.12.1 = sum(f.12.1$f - sunspot.year[(length(sunspot.year)-19):length(sunspot.year)])^2
> ase.12.1
[1] 353787
```


$$(1 - 1.06B + .4B^2)(1 - B^{10})X_t = a_t$$



```
f.10.ar2 = fore.aruma.wge(sunspot.year, phi = c(1.06, -.4), s = 10, n.ahead = 20, lastn = TRUE, limits = FALSE )
```



```
x = gen.aruma.wge(1000, phi = c(1.06, -.4), s = 10)
plots.sample.wge(x)
```

```
ase.10.ar2 = sum(f.10.ar2$f - sunspot.year[(length(sunspot.year)-19):length(sunspot.year)])^2
```

```
> ase
[1] 25391.44
```

The ARIMA(2,0,0) with $s = 10$ is the most thought to be the most useful model in terms of forecasting behavior, spectral density, ACF and ASE.