

# Computational Methods HW- 9

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[https://github.com/AasimZahoor/Comp\\_methods.git](https://github.com/AasimZahoor/Comp_methods.git)

## Question 1.

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### Given

This problem has 4 different tasks and a bonus task towards at the end. Here we are given a file *SN-m-tot-V2.0* which has sunspot numbers observed at different years starting from 1749. From observation it is known that there is a yearly cycle and a cycle of 11 years.

### Task-1

This task asks us to elaborate on the AR model I will be using. The model I have chosen has noise,  $\mu$  (average) dependence, and has terms which account for dependence on previous value, yearly cycle and the 11 year cycle. Mathematically, my model can be written as;

$$X_t = c + \phi_1 X_{t-1} + \phi_{12} X_{t-12} + \phi_{132} X_{t-132} + N(0, \sigma_z)$$

So my AR model has a p value of 3. Here, c accounts for  $\mu$ , the  $\phi$  values account for different dependencies and  $N$  accounts for noise.

$$\text{Model} = c + \phi_1 X_{t-1} + \phi_{12} X_{t-12} + \phi_{132} X_{t-132}$$

$$\text{Noise} = N(0, \sigma_z)$$

$$\text{AR model} = c + \phi_1 X_{t-1} + \phi_{12} X_{t-12} + \phi_{132} X_{t-132} + N(0, \sigma_z)$$

**Task-2**

For this task I made three functions.

- **lnprior(theta):**

This function returns prior for the given values of theta. My function favors phi values  $\leq 1$ .

Parameters-

theta : Array of parameters.

Returns : Value of log-prior

- **lnlike(theta,data):**

This function returns loglikelihood for given parameters of the model. The model is described in this part of the code.

Parameters-

theta : Array of parameters

data : data

Returns-

loglikelihood of the model with the given parameters.

- **lnprob(theta, data):**

This function gives the probability of the model with given parameters. This function calls the previous two functions

Parameters-

theta : Array of parameters

data : data

Returns-

Probability of the model with the given parameters.

So after defining the functions I used the emcee package to find the parameter values. Using those parameter values I found the fit values(for model and AR model) and noise. The equation I used is:

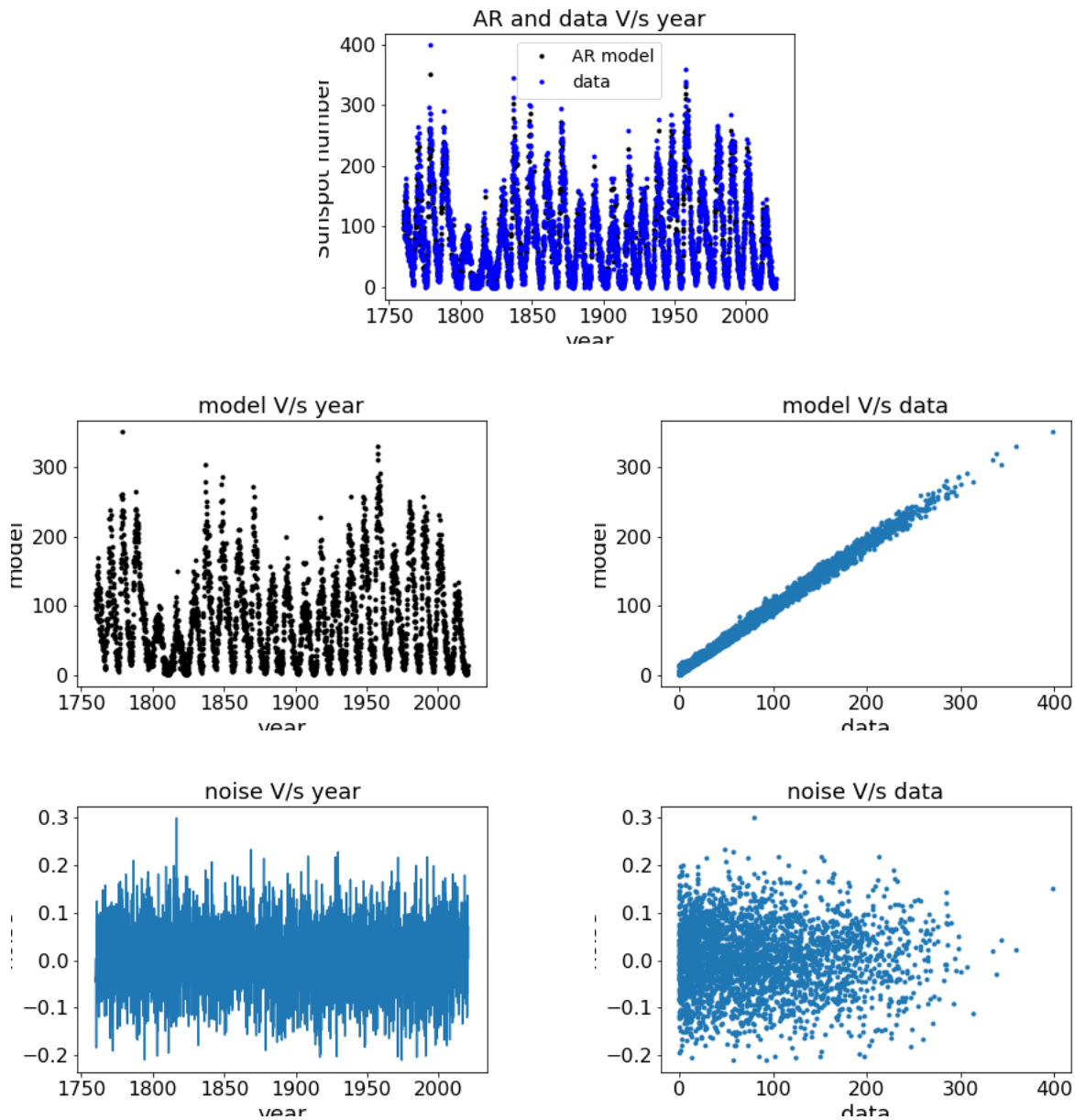
$$\text{Model} = c + \phi_1 X_{t-1} + \phi_{12} X_{t-12} + \phi_{132} X_{t-132}$$

$$\text{Noise} = N(0, \sigma_z)$$

$$\text{AR model} = c + \phi_1 X_{t-1} + \phi_{12} X_{t-12} + \phi_{132} X_{t-132} + N(0, \sigma_z)$$

Using the above equations I plotted these graphs and found the following best fit parameters.

## Plots for task 2



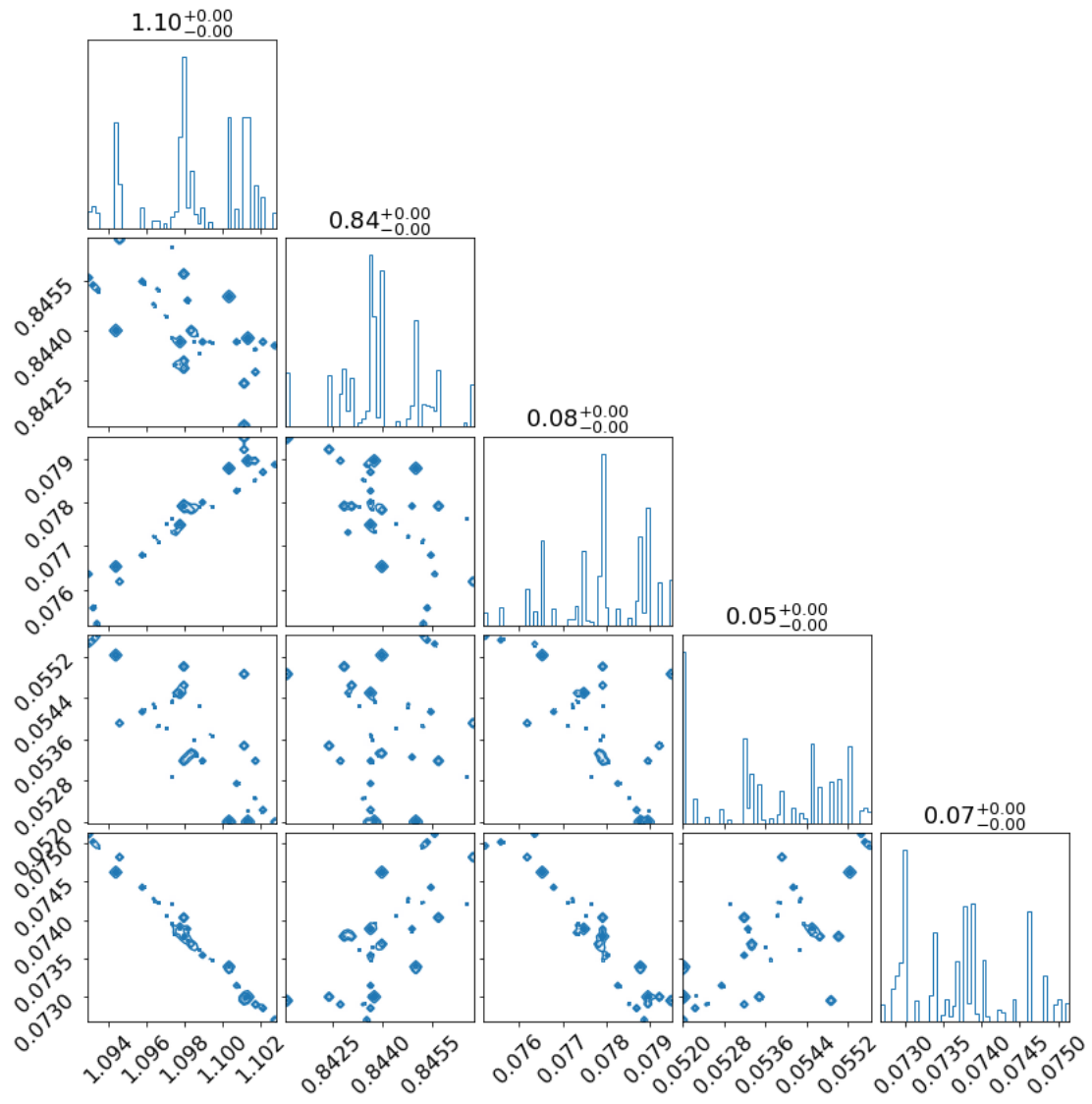
```

In [18]: runcell('-----plotting the model +noise
values-----', '/Users/aj3008/Desktop/MS_3rd_Sem/
Comp_methods_in_AST/Comp_methods/Hw_9/pb1.py')
c= 1.098479015918067 phi= 0.8439781586629621 phi12= 0.07793689909117908 phi132=
0.05369646245989251 sigma2= 0.07371053832247128

In [19]: runcell(3, '/Users/aj3008/Desktop/MS_3rd_Sem/Comp_methods_in_AST/Comp_methods/
Hw_9/pb1.py')
<Figure size 432x288 with 0 Axes>

In [20]:

```



**Task-3**

Here we are asked to plot the spectrum. I used `np.fft.fft` to find the fourier transform. Using the fourier transform I calculated the Amplitude. The frequency was calculated using a frequency function described below:

- **freq(time):**

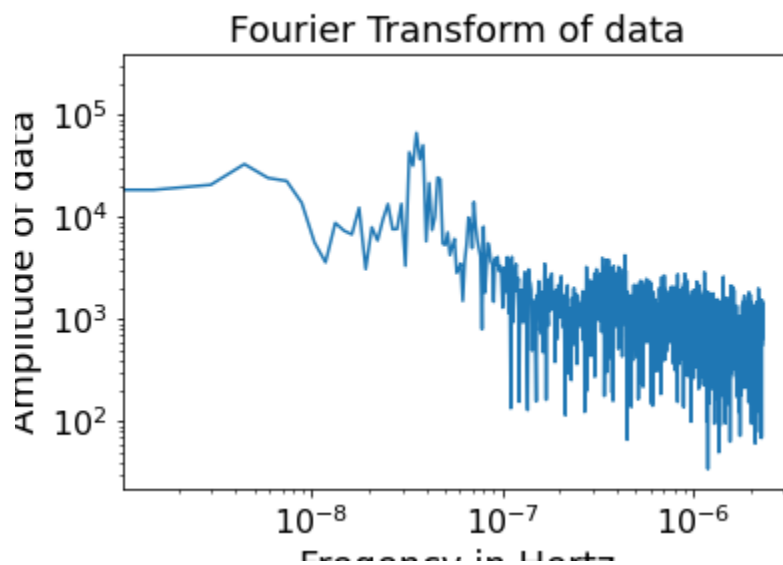
This finds the frequency for the fourier transform using the time at which data was taken.

Parameters:

time : time at which each sunspot value was taken(in days).

Returns:

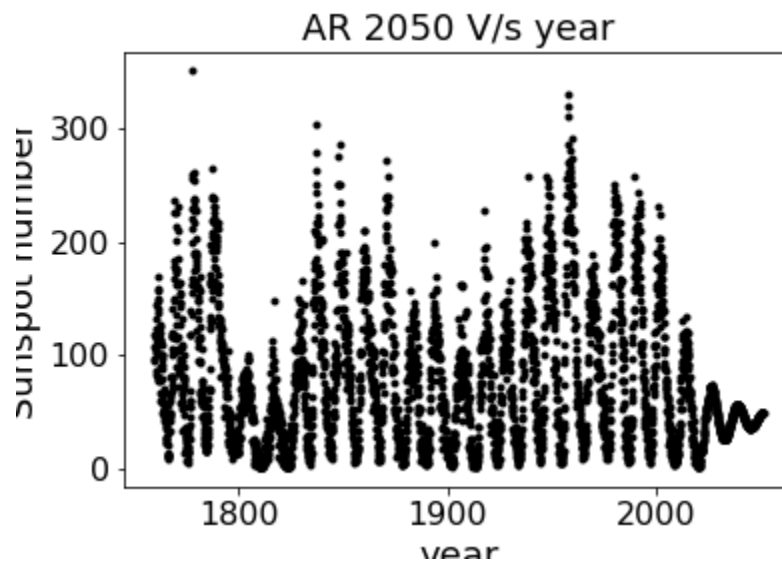
frequency in Hertz.

**Results of task-3**

From the plot it we can see there are two peaks. I think the two peak represent the yearly periodicity and 11 year periodicity. I imagine the yearly periodicity to be a wave enveloped by the 11 year periodicity wave.

**Task-4**

We are asked to plot till year 2050. To do that I first made a new time array which goes from 1749 - 2050. Then using the below equation and parameter values found earlier I found the sunspot values at different values. I felt my model is good for predicting this far in time.

**Plot**

**Bonus**

This problem asks us to find AIC for three models, including the model I used for earlier tasks , values using the equation;

$$AIC = 2k - 2\ln(L_{max})$$

Here, I used the following three AR models.Mathematically;

$$\text{Model-1} = c + \phi_1 X_{t-1} + \phi_{12} X_{t-12} + \phi_{132} X_{t-132} + N(0, \sigma_z)$$

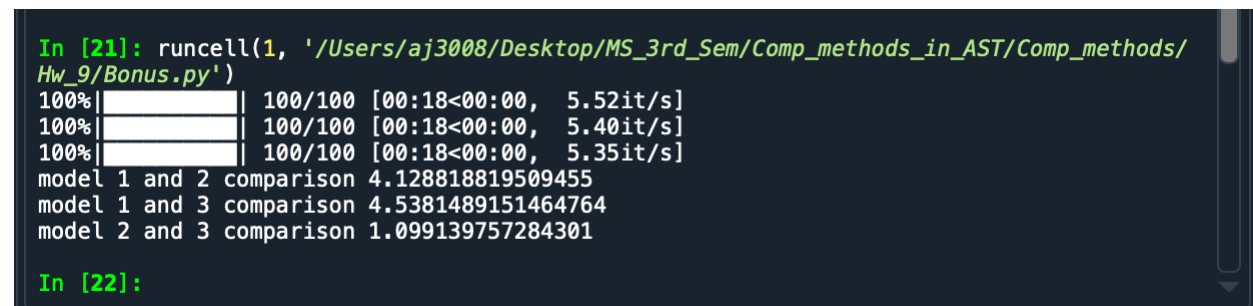
This model accounts for previous value dependence, yearly cycle and 11 year cycle. k value is 5.

$$\text{Model-2} = c + \phi_1 X_{t-1} + \phi_{12} X_{t-12} + N(0, \sigma_z)$$

This model accounts for previous value dependence and yearly cycle. k value is 4.

$$\text{Model-3} = c + \phi_1 X_{t-1} + N(0, \sigma_z)$$

This model accounts for just previous value dependence. k value is 3.



```
In [21]: runcell(1, '/Users/aj3008/Desktop/MS_3rd_Sem/Comp_methods_in_AST/Comp_methods/Hw_9/Bonus.py')
100%|██████████| 100/100 [00:18<00:00, 5.52it/s]
100%|██████████| 100/100 [00:18<00:00, 5.40it/s]
100%|██████████| 100/100 [00:18<00:00, 5.35it/s]
model 1 and 2 comparison 4.128818819509455
model 1 and 3 comparison 4.5381489151464764
model 2 and 3 comparison 1.099139757284301
In [22]:
```

Comparison was done by taking the ratio of AIC values.

I think it makes sense that model 1 is the best since it accounts for all dependencies and cycles. Model 2 accounts for yearly cycle so it is somewhat better than model 3.