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# **Tutorial 2**

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TIME SERIES & DATA ANALYSIS

NASSP – UCT

DUE 9 MARCH 2020 - 11AM

You may use any programming language.

Submit both the code (can be as Jupyter notebook) and a pdf on [cloudcape.sao.ac.za](https://cloudcape.sao.ac.za)

## Question 1

- 1.1 Code your own discrete Fourier transform (DFT).
- 1.2 Create and plot a fake lightcurve containing a sinusoidal variation of 2.5 days. (You can use `create_data` function as below, or your own.)
- 1.3 Run your DFT on the lightcurve. Plot resulting Fourier transform as a function of frequency.
- 1.4 Plot  $|F(\omega)|^2$  as a function of frequency.
- 1.4 Indicate on your plot the position of the 2.5 day input period.
- 1.6 Compare your result to the `numpy` implementation of the Fast Fourier Transform.

## Question 2

Use the Kepler lightcurve of an eclipsing binary `09514070.00.lc.data`.

- 2.1 Generate and plot a Lomb-Scargle periodogram for this lightcurve. Motivate your choice of  $f_{\min}$ ,  $f_{\max}$  and  $N_{\text{eff}}$ .
- 2.2 Find the peak frequency and fold the lightcurve at this frequency. Is this the orbital period of the binary? Substantiate your answer with text and more plots if needed.
- 2.3 Write your own bootstrapping code to determine if the measured period is significant at the 99% level.
- 2.4 How does the above result compare with the Horne & Baliunas false alarm probability?
- 2.5 A number of peaks are visible in the Lomb-Scargle periodogram. Explain what they are in terms of aliasing and/or harmonics.
- 2.6 Estimate the error on your period determination – explain your estimate.

## Question 3

Repeat Question 2 for the lightcurve `lc100099.dat`.

## Question 4

- 4.1 Using Jake Vanderplas's `create_data` function below, show that the peak width of the Lomb-Scargle periodogram is insensitive to the both the number of data points in the lightcurve, as well as the signal-to-noise in the lightcurve.

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```
1  import numpy as np
2
3  def create_data(N, T=4, signal_to_noise=5, period=1.0, random_state=None):
4      rng = np.random.RandomState(random_state)
5      t = T * rng.rand(N)
6      dy = 0.5 / signal_to_noise * np.ones_like(t)
7      y = np.sin(2 * np.pi * t / period) + dy * rng.randn(N)
8      return t, y, dy
```

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- 4.2 Discuss whether the significance of the peak is sensitive to either the number of data points in the lightcurve or the signal-to-noise.