

Why?

Given a time series with N points, you'll get $N/2+1$ independent \hat{P}_n

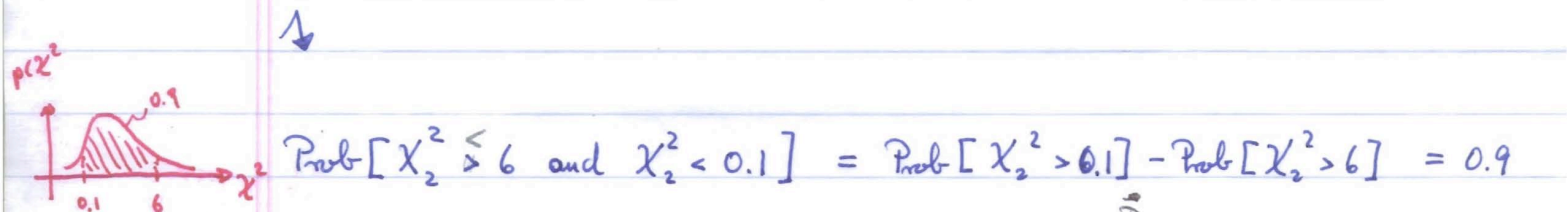
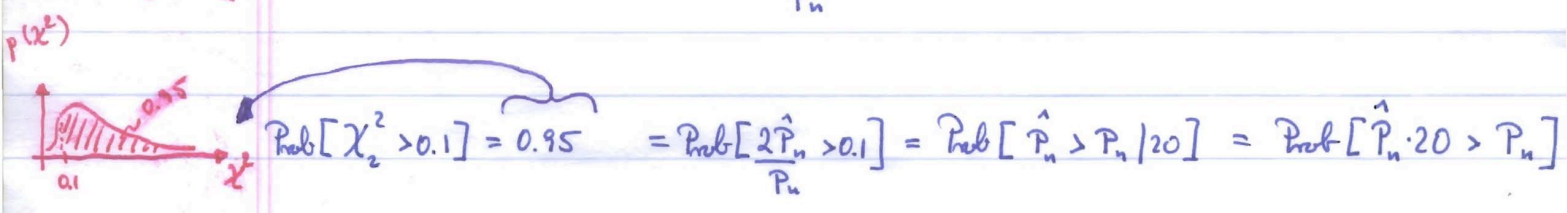
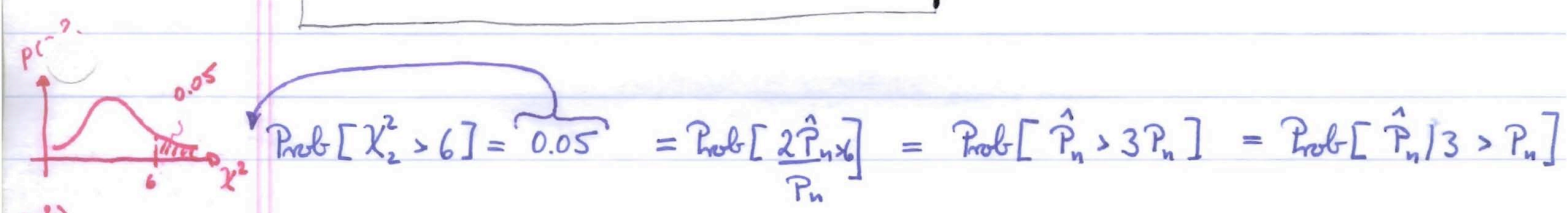
Given a time series with $2N$ points, you'll get $2N/2+1 = N+1$ independent \hat{P}_n

→ Increasing the length of the time series increases the frequency resolution ($\Delta f = 1/T = 1/N\Delta t$) and nothing else

How good are estimates \hat{P}_n ?

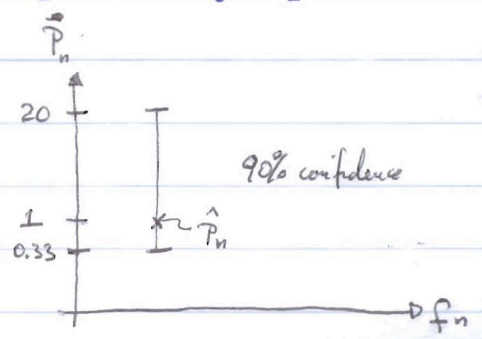
$$\frac{\hat{P}_n}{\sigma \sigma_x^2 / 2} = \frac{2 \hat{P}_n}{P_n} = \chi^2_2 \quad P_n = \Delta t \sigma_x^2 \quad E[\hat{P}_n] = P_n$$

from p.68 from p.68



↓

$$\text{Prob} \left[\frac{\hat{P}_n}{3} \leq P_n \leq 20 \hat{P}_n \right] = 0.9$$



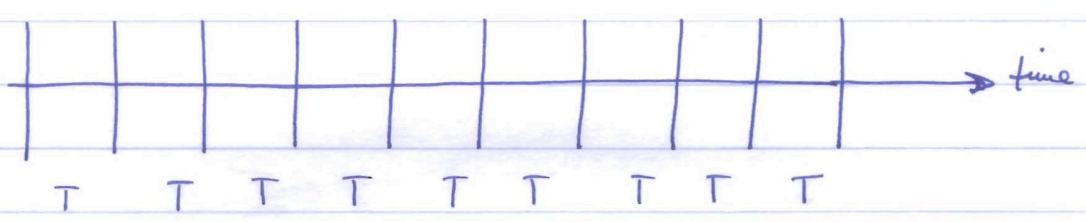
There is a 90% chance that the ~~estimate~~ ^{true} variance \bar{P}_n lies between a third and 20 times the estimated variance \hat{P}_n .

→ This "confidence" interval is ridiculous large.

→ Increase the d.o.f. somehow

How?

Do the FFT on pieces of the time series and average the estimates from each piece together.

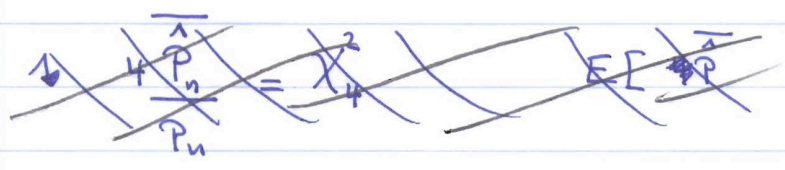


Each ^{of the "m"} segment consists of N points $\hat{P}_n = \frac{1}{m} \sum_{n=1}^m |X_{n, \omega}|^2$

For 2 segments, i.e., m=2

$$\bar{\hat{P}}_n = \frac{1}{2} (\hat{P}_{n,1} + \hat{P}_{n,2})$$

$$= \frac{\Delta t^2}{N \Delta t} \cdot \frac{1}{2} (A_{n,1}^2 + A_{n,2}^2 + B_{n,1}^2 + B_{n,2}^2)$$



for x_i from $N(0,1)$
 $\chi_n^2 = x_1^2 + x_2^2 + \dots + x_n^2$

and

$$\frac{\overline{\hat{P}_n}}{\Delta t \sigma_x^2 / 4} = \frac{A_{n,1}^2}{N\sigma_x^2/2} + \frac{A_{n,2}^2}{N\sigma_x^2/2} + \frac{B_{n,1}^2}{N\sigma_x^2/2} + \frac{B_{n,2}^2}{N\sigma_x^2/2} = \chi_4^2$$

because $E[A_{n,i}^2] = N\sigma_x^2/2$
 and thus $E\left[\frac{A_{n,i}^2}{N\sigma_x^2/2}\right] = 1$

hence $A_{n,i}$ is $N(0, \sigma_x \sqrt{N/2})$

↓
 $E[4\hat{P}/\Delta t \sigma_x^2] = E[\chi_4^2] = 4$

↓ $E[\hat{P}/\Delta t] = \sigma_x^2 = P_n / \Delta t$ or $P_n = \Delta t \sigma_x^2$

Extend this procedure to a time series consisting of m segments each $N \cdot \Delta t$ long

↓
 $\frac{2m \hat{P}_n}{P_n} = \chi_{2m}^2$ where $P_n / \Delta t = \sigma_x^2$
 and $2m$ are the degrees of freedom

~~$E\left[\frac{2m \hat{P}_n}{P_n}\right] = 2m$~~
 ~~$\frac{4m^2}{P_n^2} E[\hat{P}^2] = 2m$~~
 ~~$E[\hat{P}^2] = \frac{1}{2m} P_n$~~
 ~~$\lim_{m \rightarrow \infty} E[\hat{P}^2] = \sigma$~~
~~consistent estimate~~

$\text{Prob}[\chi_{2m}^2 > \chi_{2m, \alpha}^2] = \alpha$

$\text{Prob}[\chi_{2m}^2 > \chi_{2m, 1-\alpha}^2] = 1 - \alpha$

$\text{Prob}[\chi_{2m, \alpha}^2 \geq \chi_{2m}^2 > \chi_{2m, 1-\alpha}^2] = \text{Prob}[\chi_{2m}^2 > \chi_{2m, 1-\alpha}^2] - \text{Prob}[\chi_{2m}^2 > \chi_{2m, \alpha}^2]$
 $= 1 - 2\alpha$

