# Energy Autocorrelation Function

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### 1 Autocovariance

 $e_t$  represents the energy of the system at time t, and by time interval k, or lag k, we get energy  $e_{t+k}$ . Now we estimated define covariance  $\hat{\gamma}_k$  with lag k under non-infinite observations N as  $cov[e_t, e_{e+k}]$ , which is [1]:

$$\hat{\gamma}_k = \frac{1}{N} \sum_{t=1}^{N-k} (e_t - \mu)(e_{t+k} - \mu)$$
 (1)

where  $\mu$  is the average energy and k=0,1,2,3,...,K. Here we set K not larger than N/1000.

### 2 Autocorrelation

Then, autocorrelation function at lag k is defined as [1]:

$$\hat{\rho}_k = \frac{\hat{\gamma}_k}{\hat{\gamma}_0} \tag{2}$$

#### 2.1 Matlab Code

# 3 Results Compared with SIF

The results are shown in Fig.1.

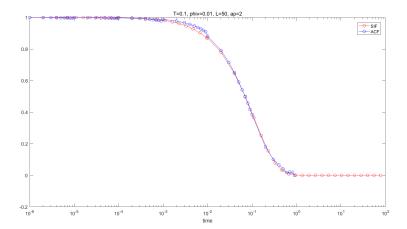


Figure 1: Results of energy autocorrelation function (ACF) and self-intermediate function (SIF) of a DPLM system with parameters shown in the title. Noted that here ACF curve is drawn by ACF results from 5 different time-scale simulations (with other parameters same), and for each result we extract their first ten time unit results, i.e., first ten lags (time = dt \* lags). For example, if time scale is from  $(10^{-6} - 10^{-1})$ , we extract a  $(10^{-6} - 10^{-5})$  result, and with five extracted results from  $(10^{-6} - 10^{-5})$   $(10^{-5} - 10^{-4})$  ...  $(10^{-1} - 10^{-0})$ , we get the blue curve.

## References

[1] G. E. P. Box, G. M. Jenkins, and G. C. Reinsel, "Time series analysis: forecasting and control," *Hoboken, New Jersey: John Wiley Sons, Inc.*, pp. 24–25, 2016.