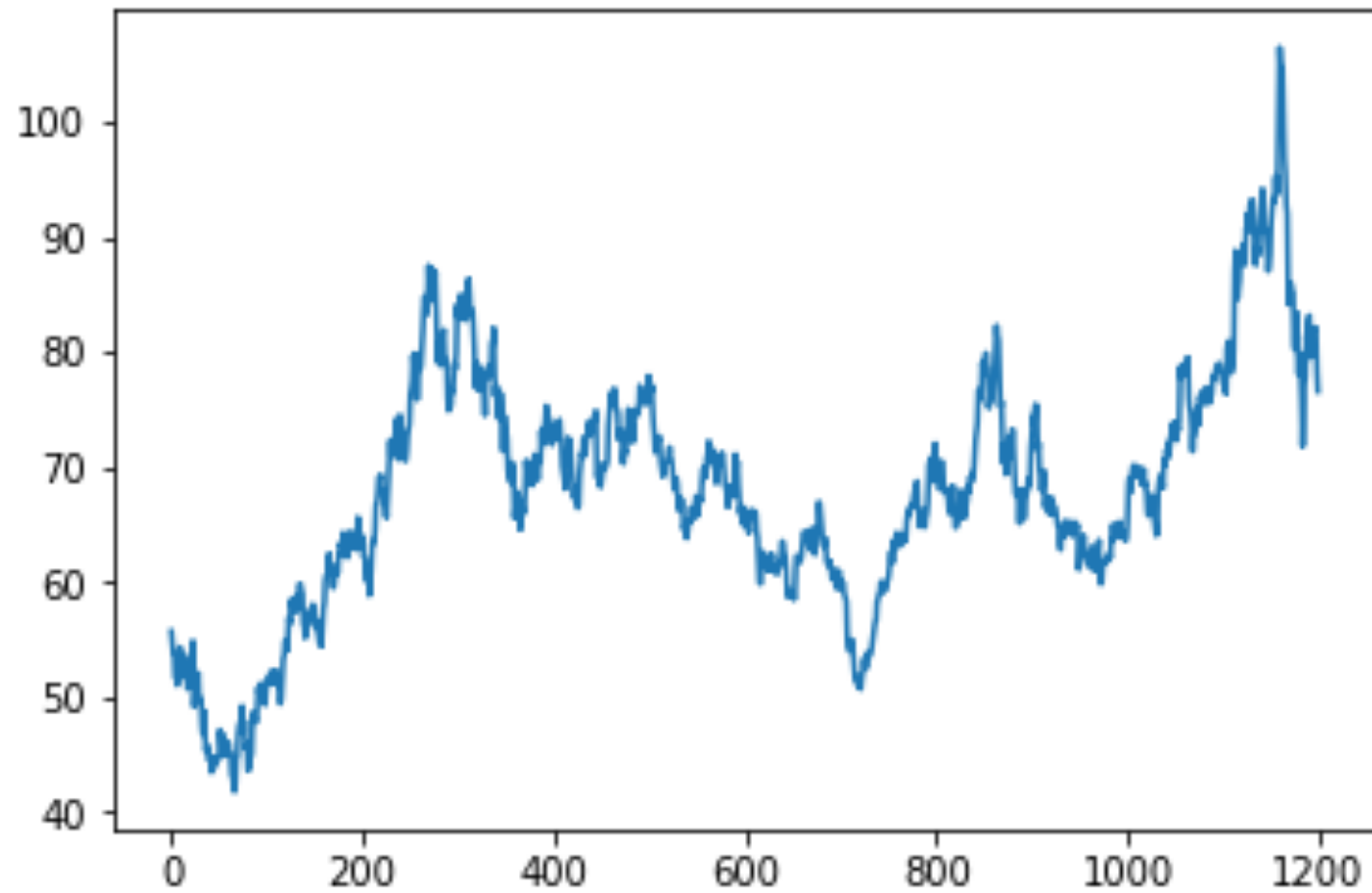
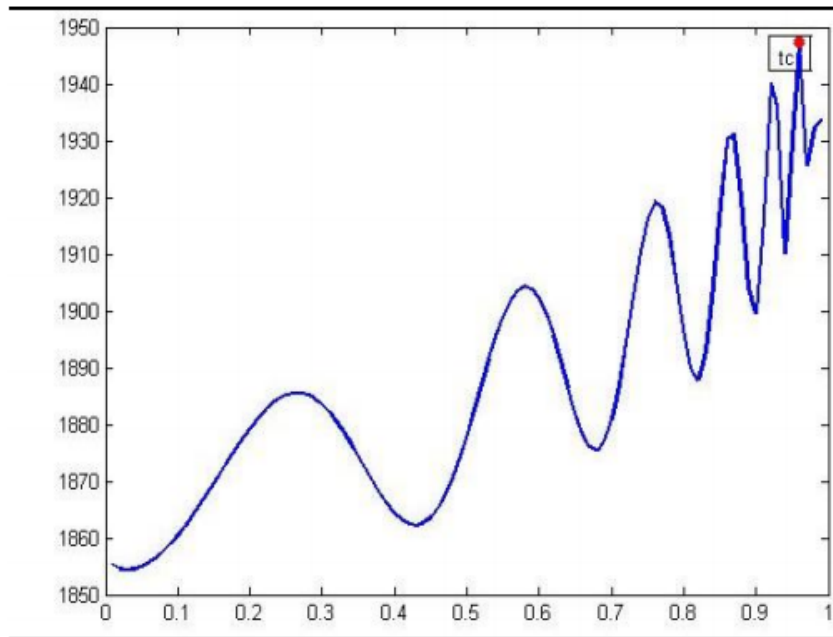


# Exercise

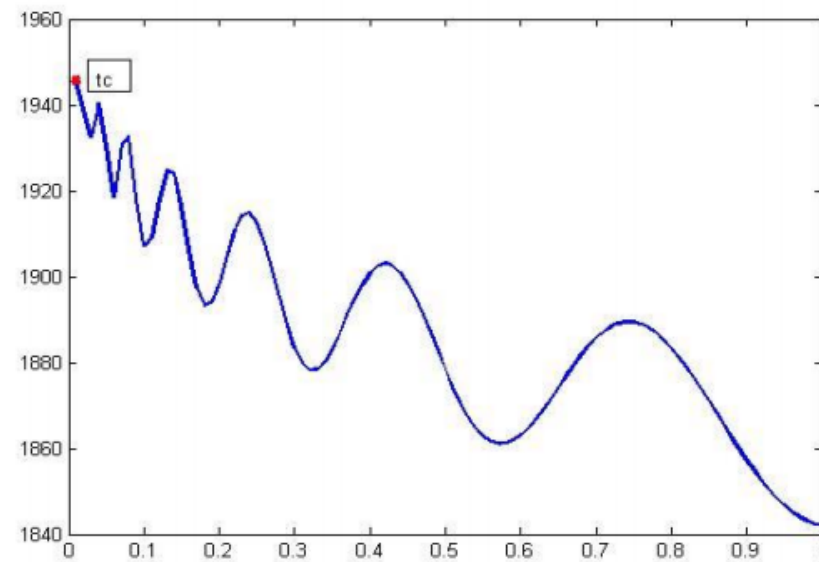
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# Financial Application (Bubble modeling)





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# log-periodic power laws (LPPL) for bubble modeling

$$\ln[p(t)] \approx A + B(t_c - t)^\beta \{1 + C \cos[\omega \ln(t_c - t) + \phi]\}, \quad (12)$$

where  $A > 0$  is the value of  $[\ln p(t_c)]$  at the critical time,  $B < 0$  is the increase in  $[\ln p(t)]$  over the time unit before the crash if  $C$  were to be close to zero,  $C \neq 0$  is the proportional magnitude of the oscillations around the exponential growth,  $0 < \beta < 1$  should be positive to ensure a finite price at the critical time  $t_c$  of the bubble and quantifies the power law acceleration of prices, and  $\omega$  is the frequency of the oscillations during the bubble, while  $0 < \phi < 2\pi$  is a phase parameter. Expression (12), which is known as the LPPL, is the fundamental equation that describes the temporal growth of prices before a crash and it has been proposed in different forms in various papers (e.g. Sornette 2003a, Lin, Ren, and Sornette 2009 and references therein). We remark that  $A$ ,  $B$ ,  $C$  and  $\phi$  are just units distributions of betas and omegas, as described in Sornette and Johansen (2001) and Johansen (2003), and do not carry any structural information.

- ▶ Two step algorithm
  - ▶ Each gene includes 4 non-linear variables  $t_c$ ,  $B$ ,  $\omega$ ,  $\Phi$
  - ▶ Use linear regression to estimate best  $A$ ,  $B$ ,  $C$
- ▶ For each parameter setting, we can measure the fitness between synthetic signals and real financial time-series data.
- ▶ Apply genetic algorithm to approximate the optimal solution by minimizing the average fitness error between time 0 and  $t_c$ .
- ▶ Homework:
  - ▶ LPPL
    - ▶ Find the optimal LPPL parameters, suppose  $t_c$  in  $[1151, 1166]$
    - ▶ Plot the synthetic signals and real time-series data with different colors in a figure