AR(1)_process

February 2, 2024

1 AR(1) process

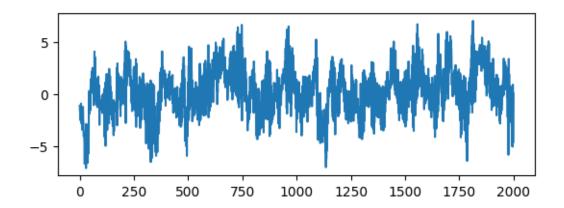
Consider a AR(1) process for r_{t+1} with $\phi_0=0.02,$ and $\phi_1=0.9$ Import Modules

```
[]: | #import the module for simulating data
     from statsmodels.tsa.arima_process import ArmaProcess
     import matplotlib.pyplot as plt
     import pandas as pd
     import numpy as np
     import math
     import matplotlib.pyplot as plt
     from datetime import datetime
     from scipy.stats import norm
     import sympy as sym
     import statsmodels.api
     from pandas import read_csv
     from matplotlib import pyplot
     from statsmodels.graphics.tsaplots import plot_acf
     from statsmodels.tsa.stattools import acf
     from sklearn.linear_model import LinearRegression
     from statsmodels.stats.diagnostic import het_white
     import statsmodels.api as sm
     from statsmodels.tsa.ar_model import AutoReg, ar_select_order
```

```
[]: # Plot 1 : AR parameter 0.2, 0.9
plt.subplot(2,1,1)
ar1 = np.array([1, -0.02, -0.9])
AR_object1 = ArmaProcess(ar1)

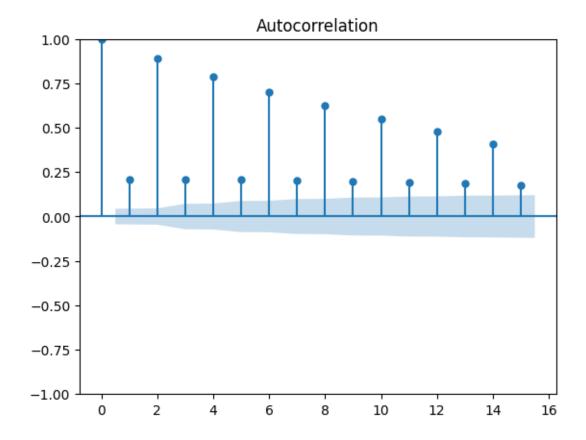
# Creating and plotting time series function of AR(1) process
simulated_data_1 = AR_object1.generate_sample(nsample = 2000)
plt.plot(simulated_data_1)
```

[]: [<matplotlib.lines.Line2D at 0x7e599d12f250>]



1. Plot the autocorrelation function for this process

[]: # Plotting the autocorrelation function for AR(1) Process with 10 lags plot_acf(simulated_data_1, lags = 15) pyplot.show()



###2. Is the process stationary?

Yes, because the plot indicates the mean and the variance of r_{t+1} is very apporximate Note that we can express the current value of an AR(1) as a function of past shocks, ϵ In particular, define $x_t = r_t - \mu$, where $\mu = E(r_t)$. Then