```
In [1]:
```

```
%matplotlib inline
```

In [78]:

```
import numpy as np
import matplotlib.pyplot as plt
class Solow Swan():
    def init (self, n=0.02, # population growth rate
                         s=0.20, # savings rate
                                  # depreciation rate
                         \delta = 0.10,
                         \alpha = 0.36,
                                   # share of labor
                                  # snare of III
# technology constant
                         z=1.00,
                         k=0.01): # initial capital stock
        self.n, self.s, self.\delta, self.\alpha, self.z = n, s, \delta, \alpha, z
        self.k = k
    def steady state(self):
        # Unpack parameters
        n, s, \delta, \alpha, z = self.n, self.s, self.\delta, self.z
        # Compute and return steady-state level of capital per worker, output per worker, and
capital per output
        capital per worker steady state = ((s * z) / (\delta + n)) * (1 / (1 - \alpha))
        output per worker steady state = (z * (s / (\delta + n)) **\alpha) ** (1 / (1 - \alpha))
        capital_per_output_steady_state = \alpha / (\delta + n)
        return capital per worker steady state, output per worker steady state,
capital per output steady state
    def random process(self, t):
         # Generate and return random process of length t
        process = np.zeros(t)
        for i in range(t):
             process[i] = np.random.randn()
        return process
    def log deviation coefficients(self):
        # Unpack parameters
        n, \delta, \alpha = self.n, self.\delta, self.\alpha
        # Compute and return coefficients \rho 1 and \rho 2
        \rho 1 = (1 + \alpha * n - \delta * (1 - \alpha)) / (1 + n)
        \rho 2 = (\delta + n) / (1 + n)
        return p1, p2
    def output per worker(self, random process):
        # Unpack parameters
        \alpha, z, k = self.\alpha, self.z, self.k
        # Generate random process of length t for level of technology
        technology_process = z * np.exp(random_process)
        # Generate and return output per worker
        return technology process * k**\alpha
    def capital_per_worker(self, random_process):
        # Unpack parameters
        n, s, \delta, \alpha, z = self.n, self.s, self.\delta, self.z
        # Update capital per worker
        return (s * self.output per worker(random process) + (1 - \delta) * self.k) / (1 + n)
    def capital per worker update(self, random process):
         # Update current state of capital per worker
        self.k = self.capital per worker(random process)
    def capital per output(self, random process):
         # Generate and return capital per output
        return self.capital per worker(random process) / self.output per worker(random process)
    def capital per worker log deviation (self, t, random process):
         # Unpack parameter
        k = self.k
          Unpack steady-state level of capital per worker
```

```
\rho1, \rho2 = self.log deviation coefficients()
        # Generate addends for summation of log deviations of capital per worker from steady state
of length t
        if t == 0:
            capital per worker log deviation addend = np.log(k / capital per worker steady state)
            capital per worker log deviation addend = np.zeros(t)
            for i in range(t-1):
                \texttt{capital\_per\_worker\_log\_deviation\_addend[i]} = \rho2 \ * \rho1**i \ * \texttt{random\_process[(t-1)-i]}
        # Generate and return log deviation of output per worker from steady state
        capital per worker log deviation = np.sum(capital per worker log deviation addend)
        return capital per worker log deviation
    def output per worker log deviation(self, t, random process):
        #Unpack parameter
        \alpha = self.\alpha
        # Unpack coefficients p1 and p2
        \rho1, \rho2 = self.log deviation coefficients()
        # Generate and return log deviation of output per worker from steady state
        output per worker log deviation = random process[t] + \alpha *
self.capital per worker log deviation(t, random process)
        return output per worker log deviation
    def sequence(self, t):
        # Unpack parameters
        n, s, \delta, \alpha, z = self.n, self.s, self.\delta, self.\alpha, self.z
        # Unpack coefficients p1 and p2
        \rho1, \rho2 = self.log deviation coefficients()
        # Generate time series of length t for capital per worker, output per worker, and capital
per worker
        capital per worker path = np.zeros(t)
        output per worker path = np.zeros(t)
        capital_per_output_path = np.zeros(t)
        # Generate time series of length t for log deviations of capital per worker and output per
worker from steady state
        capital per worker log deviation path = np.zeros(t)
        output per worker log deviation path = np.zeros(t)
        # Generate random process of length t for level of technology
        random process = self.random process(t)
        technology_process = z * np.exp(random_process)
        for i in range(t):
            # Compute current level of capital per output, output per worker, and capital per outpu
            capital per worker path[i] = self.k
            output_per_worker_path[i] = self.output_per_worker(random_process[i])
            capital_per_output_path[i] = capital_per_worker_path[i] / output_per_worker_path[i]
            # Update current state of capital per worker
            self.capital_per_worker_update(random_process[i])
            # Compute current level of log deviations of capital per output, output per worker, and
capital per output from steady state
            capital per worker log deviation path[i] = self.capital per worker log deviation(i, ran
dom process)
            output per worker log deviation path[i] = self.output per worker log deviation(i, rando
m process)
        return capital_per_worker_path, output_per_worker_path, capital_per_output_path, capital_pe
r worker log deviation path, output_per_worker_log_deviation_path, random_process
                                                                                                    |
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In [79]:
solow = Solow Swan()
In [98]:
t = 150
x = list(range(t))
In [99]:
```

solow.steady_state()

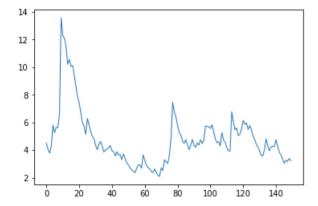
Out [99]:

In [100]:

```
k, y, ky, k_deviation, y_deviation, random_process = solow.sequence(t)
```

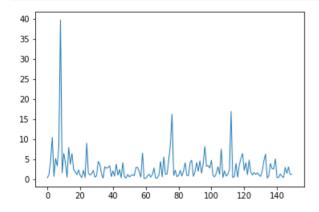
In [101]:

```
fig, ax = plt.subplots()
ax.plot(x, k, linewidth=1)
plt.show()
```



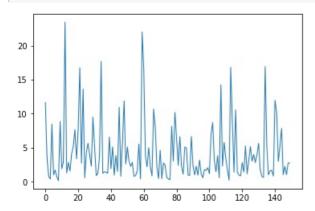
In [102]:

```
fig, ax = plt.subplots()
ax.plot(x, y, linewidth=1)
plt.show()
```



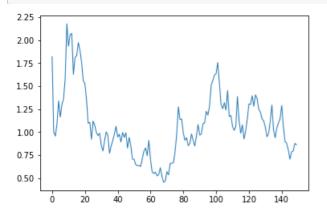
In [103]:

```
fig, ax = plt.subplots()
ax.plot(x, ky, linewidth=1)
plt.show()
```



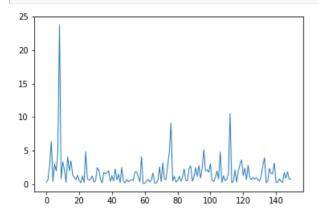
In [104]:

plt.show()



In [105]:

```
fig, ax = plt.subplots()
ax.plot(x, np.exp(y_deviation), linewidth=1)
plt.show()
```



In [106]:

```
fig, ax = plt.subplots()
ax.plot(x, random_process, linewidth=1)
plt.show()
```

