

2.1

(a)首先化简(1)可得

$$\begin{aligned} p_i(t+1) &= p_i(t) \times \frac{\alpha\gamma}{S_i(t)} \\ &= \alpha\gamma \times \frac{p_i(t)q_i(t)}{s_i(t)} \\ &= \alpha\gamma \frac{\left(\sigma + \sum_{j \neq i} G_{ij}p_j(t)\right) p_i(t)}{G_{ii}p_i(t)} \\ &= \alpha\gamma \sum_{j \neq i} \frac{G_{ij}}{G_{ii}} p_j(t) + \frac{\alpha\gamma\sigma}{G_{ii}} \\ &= \alpha\gamma \left(\sum_{j=1}^n \frac{G_{ij}}{G_{ii}} p_j(t) \right) - \alpha\gamma p_i(t) + \frac{\alpha\gamma\sigma}{G_{ii}} \end{aligned}$$

记

$$\Lambda = \text{diag}(G_{11}, \dots, G_{nn})$$
$$1_n = \begin{bmatrix} 1 \\ \vdots \\ 1 \end{bmatrix}$$

所以

$$\begin{aligned} p(t+1) &= \alpha\gamma (\Lambda^{-1}G - I_n) p(t) + \alpha\gamma\sigma\Lambda^{-1}1_n \\ &\triangleq Ap(t) + b \end{aligned}$$

其中

$$\begin{aligned} A &= \alpha\gamma (\Lambda^{-1}G - I_n) \\ b &= \alpha\gamma\sigma\Lambda^{-1} \end{aligned}$$

(b)首先给出 $s_i(t), q_i(t)$ 的矩阵形式

$$\begin{aligned} s(t) &= \Lambda^{-1}p(t) \\ q(t) &= \sigma I_n + (G - \Lambda)p(t) \end{aligned}$$

对应代码如下：

```
import numpy as np
import matplotlib.pyplot as plt
plt.rcParams['font.sans-serif']=['SimHei'] #用来正常显示中文标签
plt.rcParams['axes.unicode_minus']=False #用来正常显示负号
```

```

G = np.array([[1, 0.2, 0.1], [0.1, 2, 0.1], [0.3, 0.1, 3]])
gamma = 3
alpha = 1.2
sigma = 0.01
p = np.random.rand(3).reshape(-1, 1)
N = 300

```

```

def f(G, gamma, alpha, sigma, p, N):
    #数据维度
    n = G.shape[0]

    Lambda = np.diag(G).reshape(-1, 1)
    Res = np.array([])
    for i in range(N):
        s = Lambda * p
        q = sigma + G.dot(p) - Lambda * p
        S = s / q
        if i == 0:
            Res = np.copy(S)
        else:
            Res = np.c_[Res, S]
        #更新
        p = alpha * gamma * ((G / Lambda).dot(p) - p) + \
            alpha * gamma * sigma / Lambda

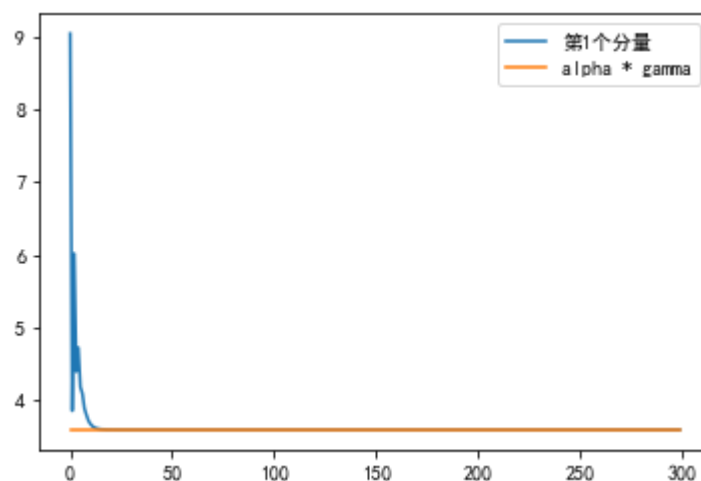
    target = np.ones(N) * alpha * gamma
    for i in range(n):
        si = Res[i, :]
        label = "第" + str(i+1) + "个分量"
        plt.plot(si, label=label)
        plt.plot(target, label="alpha * gamma")
        plt.legend()
        plt.show()

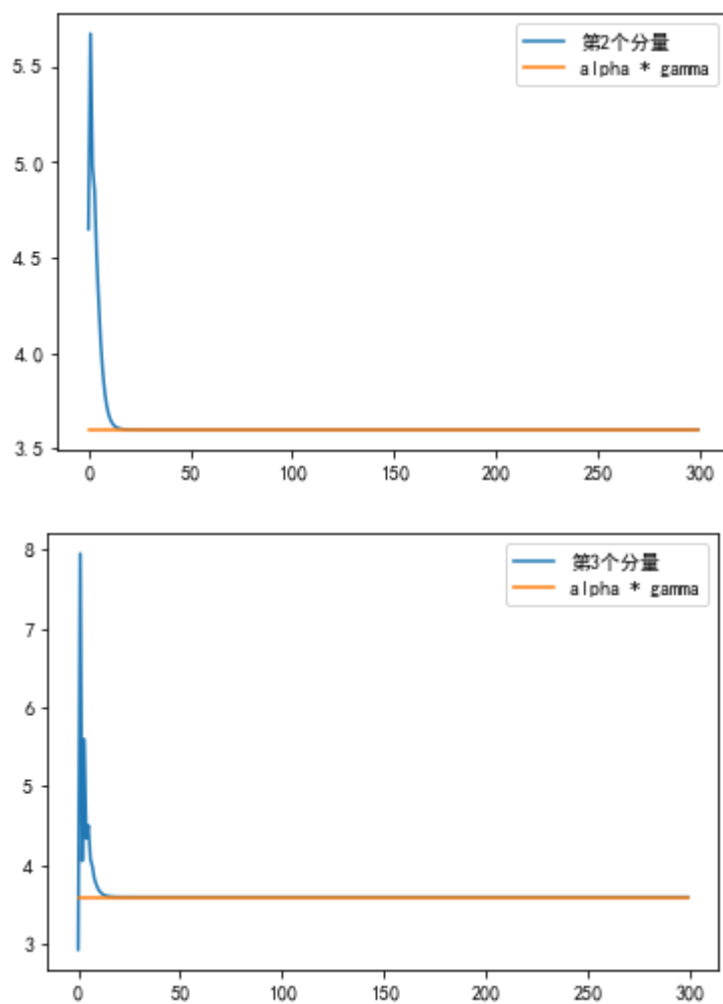
```

```

f(G, gamma, alpha, sigma, p, N)

```





利用该算法，最后每个分量都会收敛到 $\alpha\gamma$ 。

2.2

因为

$$M\ddot{q} + D\dot{q} + Kq = f$$

并且 M 可逆，所以

$$\ddot{q} = M^{-1}f - M^{-1}Kq - M^{-1}D\dot{q}$$

回顾动力系统的形式：

$$\dot{x} = Ax + Bu, \quad y = Cx + Du$$

结合题目可得

$$\begin{aligned}
\dot{x} &= \begin{bmatrix} \dot{q} \\ \ddot{q} \end{bmatrix} \\
&= \begin{bmatrix} \dot{q} \\ M^{-1}f - M^{-1}Kq - M^{-1}D\dot{q} \end{bmatrix} \\
&= \begin{bmatrix} 0 & I_k \\ -M^{-1}K & -M^{-1}D \end{bmatrix} \begin{bmatrix} q \\ \dot{q} \end{bmatrix} + \begin{bmatrix} 0 \\ M^{-1} \end{bmatrix} f \\
&= \begin{bmatrix} 0 & I_k \\ -M^{-1}K & -M^{-1}D \end{bmatrix} x + \begin{bmatrix} 0 \\ M^{-1} \end{bmatrix} u \\
y &= q \\
&= [I_k \quad 0] \begin{bmatrix} q \\ \dot{q} \end{bmatrix} \\
&= [I_k \quad 0] x
\end{aligned}$$

2.3

回顾离散时间的线性动力系统形式：

$$x(t+1) = A(t)x(t) + B(t)u(t), \quad y(t) = C(t)x(t) + D(t)u(t)$$

MA模型：

记

$$\begin{aligned}
x(k) &= \begin{bmatrix} u(k-1) \\ \vdots \\ u(k-r) \end{bmatrix} \\
u(k) &= u(k) \\
C &= [a_1, \dots, a_r] \in \mathbb{R}^{1 \times r} \\
D &= a_0
\end{aligned}$$

那么

$$y(k) = Cx(k) + Du(k)$$

AR模型：

记

$$\begin{aligned}
 x(k) &= \begin{bmatrix} y(k-1) \\ \vdots \\ y(k-p) \end{bmatrix} \\
 u(k) &= \begin{bmatrix} u(k) \\ 0 \\ \vdots \\ 0 \end{bmatrix} \in \mathbb{R}^p \\
 b &= [b_1, \dots, b_p] \in \mathbb{R}^{1 \times p} \\
 A &= \begin{bmatrix} b \\ I_{p-1} & 0 \end{bmatrix} \\
 B &= 1
 \end{aligned}$$

那么

$$\begin{aligned}
 x(k+1) &= \begin{bmatrix} y(k) \\ y(k-1) \\ \vdots \\ y(k-p+1) \end{bmatrix} \\
 &= \begin{bmatrix} u(k) + b_1 y(k-1) + \dots + b_p y(k-p) \\ y(k-1) \\ \vdots \\ y(k-p+1) \end{bmatrix} \\
 &= \begin{bmatrix} b_1 y(k-1) + \dots + b_p y(k-p) \\ y(k-1) \\ \vdots \\ y(k-p+1) \end{bmatrix} + \begin{bmatrix} u(k) \\ 0 \\ \vdots \\ 0 \end{bmatrix} \\
 &= \begin{bmatrix} b \\ I_{p-1} & 0 \end{bmatrix} x(k) + u(k) \\
 &= Ax(k) + Bu(k)
 \end{aligned}$$

ARMA模型:

记

$$x(k) = \begin{bmatrix} y(k-1) \\ \vdots \\ y(k-p) \\ u(k-1) \\ \vdots \\ u(k-r) \end{bmatrix} \in \mathbb{R}^{p+r}$$

$$u(k) = u(k)$$

$$a = [a_1, \dots, a_r] \in \mathbb{R}^{1 \times r}$$

$$b = [b_1, \dots, b_p] \in \mathbb{R}^{1 \times p}$$

$$A_1 = [a, b] \in \mathbb{R}^{1 \times p}$$

$$A_2 = [I_{p-1}, 0] \in \mathbb{R}^{(p-1) \times p}$$

$$A_3 = 0 \in \mathbb{R}^{1 \times p}$$

$$A_4 = [0, I_{r-1}, 0] \in \mathbb{R}^{(r-1) \times p}$$

那么

$$\begin{aligned}
x(k+1) &= \begin{bmatrix} y(k) \\ y(k-1) \\ \vdots \\ y(k-p+1) \\ u(k) \\ u(k-1) \\ \vdots \\ u(k-r+1) \end{bmatrix} \\
&= \begin{bmatrix} b_1 y(k-1) + \cdots + b_p y(k-p) + a_0 u(k) + \cdots + a_r u(k-r) \\ y(k-1) \\ \vdots \\ y(k-p+1) \\ u(k) \\ u(k-1) \\ \vdots \\ u(k-r+1) \end{bmatrix} \\
&= \begin{bmatrix} A_1 \\ A_2 \\ A_3 \\ A_4 \end{bmatrix} x(k) + \begin{bmatrix} a_0 \\ 0 \\ \vdots \\ 0 \\ 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix} u(k)
\end{aligned}$$

记

$$A = \begin{bmatrix} A_1 \\ A_2 \\ A_3 \\ A_4 \end{bmatrix} \in \mathbb{R}^{(p+r) \times (p+r)}$$

$$B = \begin{bmatrix} a_0 \\ 0 \\ \vdots \\ 0 \\ 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix} \in \mathbb{R}^{p+r}$$

其中

$$B_i = \begin{cases} a_0 & i = 1 \\ 1 & i = p + 1 \\ 0 & \text{其他} \end{cases}$$

那么模型为

$$x(k+1) = Ax(k) + Bu(k)$$

2.4

定义 n 维标准单位列向量 $u_i \in \mathbb{R}^n$:

$$u_1 = \begin{bmatrix} 1 \\ 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix}, u_2 = \begin{bmatrix} 0 \\ 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix}, \dots, u_n = \begin{bmatrix} 0 \\ 0 \\ 0 \\ \vdots \\ 1 \end{bmatrix}$$

那么

$$x = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} = \sum_{i=1}^n x_i u_i$$

定义 m 维标准单位列向量 $v_i \in \mathbb{R}^m$:

$$v_1 = \begin{bmatrix} 1 \\ 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix}, v_2 = \begin{bmatrix} 0 \\ 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix}, \dots, v_m = \begin{bmatrix} 0 \\ 0 \\ 0 \\ \vdots \\ 1 \end{bmatrix}$$

假设

$$f(u_i) = \sum_{j=1}^m a_{ji} v_j, A = [a_{ji}] \in \mathbb{R}^{m \times n}$$

那么

$$\begin{aligned} f(x) &= f\left(\sum_{i=1}^n x_i u_i\right) \\ &= \sum_{i=1}^n x_i f(u_i) \\ &= \sum_{i=1}^n x_i \sum_{j=1}^m a_{ji} v_j \\ &= \sum_{j=1}^m \sum_{i=1}^n a_{ji} x_i v_j \\ &= \begin{bmatrix} \sum_{i=1}^n a_{1i} x_i \\ \dots \\ \sum_{i=1}^n a_{mi} x_i \end{bmatrix} \\ &= Ax \end{aligned}$$

如果还存在 $\tilde{A} \in \mathbb{R}^{m \times n}$, 使得

$$f(x) = \tilde{A}x$$

那么取 $x = u_i, i = 1, \dots, n$ 可得

$$\begin{aligned} f(u_i) &= Au_i \\ &= \begin{bmatrix} a_{1i} \\ \dots \\ a_{mi} \end{bmatrix} \\ &= \tilde{A}u_i \\ &= \begin{bmatrix} \tilde{a}_{1i} \\ \dots \\ \tilde{a}_{mi} \end{bmatrix} \end{aligned}$$

因此

$$A = \tilde{A}$$

2.6

由

$$\mathcal{D}p = \sum_{i=1}^{n-1} i a_i x^{i-1}$$

可得

$$\begin{bmatrix} a_1 \\ 2a_2 \\ \vdots \\ (n-1)a_{n-1} \end{bmatrix} = \begin{bmatrix} 0 & 1 & & \\ 0 & & 2 & \\ 0 & & & \ddots \\ 0 & & & & n-1 \end{bmatrix} \begin{bmatrix} a_0 \\ a_1 \\ \vdots \\ a_{n-1} \end{bmatrix}$$

因此

$$D = \begin{bmatrix} 0 & 1 & & \\ 0 & & 2 & \\ 0 & & & \ddots \\ 0 & & & & n-1 \end{bmatrix}$$

2.9

(a)不难看出

$$\begin{bmatrix} y_1 \\ y_2 \end{bmatrix} = \begin{bmatrix} 2 & 0 \\ 0.5 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

所以

$$A = \begin{bmatrix} 2 & 0 \\ 0.5 & 1 \end{bmatrix}$$

(b)利用(a)不难得出

$$\begin{aligned}
 B &= A^4 \\
 &= \left(\begin{bmatrix} 2 & 0 \\ 0.5 & 1 \end{bmatrix} \begin{bmatrix} 2 & 0 \\ 0.5 & 1 \end{bmatrix} \right)^2 \\
 &= \left(\begin{bmatrix} 4 & 0 \\ 1.5 & 1 \end{bmatrix} \right)^2 \\
 &= \begin{bmatrix} 4 & 0 \\ 1.5 & 1 \end{bmatrix} \begin{bmatrix} 4 & 0 \\ 1.5 & 1 \end{bmatrix} \\
 &= \begin{bmatrix} 16 & 0 \\ 7.5 & 1 \end{bmatrix}
 \end{aligned}$$

另一方面，直接考虑该问题，我们知道 x_1 到 z_1 的路线只有一条，所以 $b_{11} = 2^4 = 16$ ； x_2 到 z_2 的路线只有一条，所以 $b_{22} = 1^4 = 1$ ； x_2 到 z_1 没有路线，所以 $b_{12} = 0$ ； x_1 到 y_2 的路线一共有4条，总权重为

$$0.5 + 2 \times 0.5 + 2^2 \times 0.5 + 2^3 \times 0.5 = 7.5$$

2.12

首先考虑 $(A^2)_{ij}$ ：

$$(A^2)_{ij} = \sum_{k=1}^n A_{ik} A_{kj} = i \text{ 到 } j \text{ 长度为 } 2 \text{ 的路径数量}$$

递推可得

$$B_{ij} = (A^k)_{ij} = i \text{ 到 } j \text{ 长度为 } k \text{ 的路径数量}$$

补充题

(a) $\forall \alpha, \beta, \alpha + \beta = 1$ ：

$$\begin{aligned}
 f(\alpha x + \beta y) &= A(\alpha x + \beta y) + b \\
 &= A(\alpha x + \beta y) + b(\alpha + \beta) \\
 &= \alpha(Ax + b) + \beta(Ax + b) \\
 &= \alpha f(x) + \beta f(y)
 \end{aligned}$$

(b) 考虑

$$g(x) = f(x) - f(0)$$

下面证明该函数为线性函数，首先证明

$$g(kx) = kg(x)$$

由定义，这等价于

$$\begin{aligned}
 f(kx) - f(0) &= k(f(x) - f(0)) \Leftrightarrow \\
 f(kx) + (k-1)f(0) &= kf(x) \Leftrightarrow \\
 \frac{1}{k}f(kx) + \frac{k-1}{k}f(0) &= f(x)
 \end{aligned}$$

最后一行由 $f(x)$ 的性质即可得到。

接着证明

$$g(x+y) = g(x) + g(y)$$

事实上, 我们有

$$\begin{aligned}
 g(x+y) &= f(x+y) - f(0) \\
 &= f\left(\frac{1}{2} \times 2x + \frac{1}{2} \times 2y\right) - \frac{1}{2}f(0) - \frac{1}{2}f(0) \\
 &= \frac{1}{2}f(2x) + \frac{1}{2}f(2y) - \frac{1}{2}f(0) - \frac{1}{2}f(0) && \text{由 } f(x) \text{ 的性质} \\
 &= \frac{1}{2}(f(2x) - f(0)) + \frac{1}{2}(f(2y) - f(0)) \\
 &= \frac{1}{2}g(2x) + \frac{1}{2}g(2y) \\
 &= \frac{1}{2} \times 2 \times g(x) + \frac{1}{2} \times 2 \times g(y) && \text{由 } g(kx) = kg(x) \\
 &= g(x) + g(y)
 \end{aligned}$$

结合以上两点, $g(x)$ 是线性函数, 所以存在唯一的 A , 使得

$$g(x) = f(x) - f(0) = Ax$$

现在记

$$b = f(0)$$

那么

$$g(x) = Ax + b$$

A 的唯一性已经说明, 现在说明 b 的唯一性即可。若存在 \tilde{b} 同样满足条件, 那么

$$\tilde{b} = g(0) = b$$

因此 b 唯一。