SDC, HW2 AIIII37 如柏鍹 Exercise 2. 0.2 tomorrow will be. (rainy) 20.2 0.8 CSWNNY 0.4 G(cloudy) sunny cloudy rainy cloudy today it's... rainy 0.2 (a) Day 1 is sunny. $c \rightarrow c \rightarrow r$ $0.2 \times 0.4 \times 0.2 = 0.016$ # (b) import random def main(): print("HW2-1") future_day = 10 for day in range(future_day): weight = weight_list[weather_list[day]] random_result = random.choices(weather,weights = weight) weather_list.append(random_result[0]) 力輸出 print(weather_list) __name__ = "__main__": main() gogochiou@gogochiouMSI:~/NTHU_course/NYCU_SDC/H HW2-1 ['sunny', 'cloudy', 'sunny', 'cloudy', 'cloudy', 'cloudy', 'cloudy', 'cloudy', 'rainy', 'cloudy', (C) main(): print("HW2-2") future_day = 1000000 s_count = 0 c_count = 0 r_count = 0 for day in range(future_day): weight = weight_list[weather_list[day]] random_result = random.choices(weather, weights = weight) weather_list.append(random_result[0]) if random_result[0] = 'sunny': s_count+=1 elif random_result[0] = 'cloudy': c_count+=1 elif random_result[0] = 'rainy': r_count+=1 0.643924 # print(weather_list) 0.284938 print(s_count/future_day) 0.071138 print(c_count/future_day) print(r_count/future_day) stationary distribution (d)(Xt 显當天 3种天氣的机率), χ_{t} . $\chi_{t+1} = T \chi_t$ -0.8 0.4 0.2 0,2 0.4 0.6 Τ where 0 0.2 0.2 $Ax = \lambda \cdot x$, $\lambda = 1$ Think as 0.8 x_1 + 0.4 x_2 + 0.2 x_3 = $\alpha_1 = \frac{4}{14} = 0.642857$ χ , – $\chi_2 = \frac{1}{7} = 0.285714$ $0.2 \chi_1 + 0.4 \chi_2 + 0.6 \chi_3 =$ χ2 -

 $0.2 \chi_2 + 0.2 \chi_3 =$

1

 $\chi_1 + \chi_2 + \chi_3 =$

 χ_3

 $\chi_3 = \frac{1}{14} = 0.071429$

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        tomorrow will be...

        sunny
        cloudy
        rainy

        today it's...
        cloudy
        4
        4
        2

        rainy
        2
        6
        2
        0
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(e) $\Sigma P_{(x)} \log_2 \frac{1}{P_{(x)}} = 1.1981 + (by matlab colculate)$

 $P(x_t) = \sum P(x_t | x_{t-1}) \cdot P(x_{t-1})$ (日午) 今 sunny -> sunny : 0.8 $P(x_t = sunny) = 0.642867$ by cloudy: 0.2 × 0.642857 = 0.45. $P(x_t = c|oudy) = 0.285714$ 0.285714 rainy : Plxt = rainy) = 0.071429. (日午) (日乍) Cloudy → sunny : 0.4x 0.285714 = 0.177775. rainy \Rightarrow sunny $\frac{0.2 \times 0.071429}{0.642857} = 0.022$ cloudy cloudy: 0.6 × 0.071479 = 0.15 +,0 : $\sqrt{\frac{0.2 \times 0.285714}{10.2000}} = 0.8.$ rainy : 0.1 0.071429

 yesterday

 sunny
 cloudy
 rainy

 sunny
 0.8
 0.179775
 0.022

 today
 cloudy
 0.45
 0.4
 0.15

 rainy
 0
 0.8
 0.2

(g) Markov assumption 階 設只要 Xt-1 的狀態,就能知道 Xt, 世就是 Xt-2 ~ Xi 的 state 並不需要。所以 system still satisfy Markov assumption,只是 Gtate transition function 在不同時序會 不同,或我們可以把 State 擴展到 不同季節的 Sunny, Cloudy, Cainy, 世就是 4x3 = 12 個 State #

Exercise 3. (星exa.)

		our sensor tells us					tomorrow will be		be	
		sunny	cloudy	rainy				sunny	cloudy	rai
the actual weather is	sunny	.6	.4	0		today it's	sunny	.8	.2	0
	cloudy	.3	.7	0			cloudy	.4	.4	.2
	rainy	0	0	1			rainy	.2	.6	.2

(a) 因 sensor th rainy, 一定是 rainy , 又 by morkov assumption.

P(Xz = sunny | Zz = rainy , Zz = sunny)

P(Zs = Sunny | Xs = Sunny) · P(Xs = Sunny | Z4 = rainy)

 $P(z_5 = sunny \mid x_6 = sunny \mid x_5 = sunny \mid x_5 = rainy) + P(z_5 = sunny \mid x_6 = cloudy) P(x_5 = cloudy) P(x_5 = rainy) + P(z_5 = sunny \mid x_5 = rainy) P(x_5 = rainy) P(x$

$$= \frac{0.6 \times 0.2 + 0.8 \times 0.6 + 0}{0.6 \times 0.2 + 0.8 \times 0.6 + 0} = 0.4$$

 $\Sigma P(2_2|x_2') P(x_2'|x_1) = 0.6 \times 0.8 + 0.8 \times 0.2 = 0.54$

Day 2 is sunny:
$$\frac{0.6 \times 0.8}{0.54} = \frac{8}{9}$$
 $\frac{1}{9}$
 $\frac{1}{9}$
 $\frac{1}{9}$
 $\frac{1}{9}$
 $\frac{1}{9}$
 $\frac{1}{9}$

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D_{av} \delta, \qquad \chi_2 = \begin{bmatrix} 8/4 \\ 1/4 \\ 0 \end{bmatrix} \qquad \chi_3 = 8unny
                                               \frac{P(Z_3 \mid x_1, x_3, Z_2) P(x_3 \mid x_1, Z_2)}{\sum P(Z_3 \mid x_3') P(x_3' \mid x_1, Z_2)}
                  \rho(X_3 \mid Y_1, Z_2 \cdot 3) =
                   Z P (Z_3 | X_3') \cdot P(X_3' | X_1, Z_2) = (0.6 \quad 0.8 \quad 0) \left( \begin{bmatrix} 0.8 & 0.4 & 0.2 \\ 0.2 & 0.4 & 0.6 \\ 0 & 0.2 & 0.2 \end{bmatrix} \begin{bmatrix} \frac{Q}{q} \\ \frac{1}{q} \\ 0 \end{bmatrix} \right) 
                                                         = (0.6 \ 0.3 \ 0) \cdot \begin{pmatrix} 34/45 \\ 2/9 \\ 1/45 \end{pmatrix} = 0.52.
                                            0.6 x 84/45 = 0.8718
                  Day 3 is sunny :
                                              0.8 x */9 = 0.1282
      Day 4. => Sensor tells rainy
                        ⇒ Pay 4 is sunny '
                                                         cloudy :
-way 2.
                                                                 Z2 = Sunny , Z3 = Sunny , Z4 = rainy .
          Day 2. with day 3,4 dat
                    P(\chi_2 = 1 \mid Z_{2:4}) = \sqrt{P(Z_2 = 1 \mid \chi_2 = 1, Z_{3:4})} \times P(\chi_2 = 1 \mid Z_{3:4})
                                                   = n . P(Z==1) x==1) P(Z==1 | x==1 , Z+) > P(x==1 | Z+)
                                                    = n · P(Z==1 | X==1) · ( Z · P(Z==1 | X==i , X==1, Z+) × P(X==i | X==1, Z+)) × P(X==1 | Z+)
                                                    = 1 . P [Z==1 | X==1) . (\(\Sigma\) P (Z3=1 | X3=i) * P (X3 . i | X2=1) ) * P (Z4 | X==1) * P (X1=1)
                                                    = 1 - 0.6 x ( 0.6 x 0.8 + 0.3 x 0.2) x (\( \Sigma P(Z=3 | \chi_+ > \bar{\eta} \), \( \chi_2=1 \) \( P(\chi_+ = \bar{\eta} | \chi_2 = 1) \) \( P(\chi_+ = \bar{\eta} | \chi_2 = 1) \)
                                                                                                = ZP(Z+=3|X+=j) * ZP(X+=j|X==k) *P(X3=k|X==1)
                                                                                                 = 1 - (\sum P(X_4 = 3 | X_3 = k) + P(X_3 = k | X_2 = 1))
                                                                                                  = 1 × (8.8×0 + 8.2×8.2+ 8×0.2) = 0.04.
                     P(X=1|Z=+) = n · 0.6 x 0.54 x 0.04 x 0.8 = 0.0104 n.
                    P(x_2=2|Z_{2:4}) = \eta P(Z_2=1|X_2=2, Z_{3:4}) P(X_2=2|Z_{3:4})
                                        = n. P(Z==1/2=2) · P(Z==1 | X==2, Z+) * P(X==2 | Z+)
                                        = 1 P(Z==1|x==2) . (\(\S \) P(Z==1|x==i \) * P(x==i \x=2) ) * (\(\S \) P(x==3|x==k) * P(x==k|x==2) ) . P(x==2)
                                         = 10.3 \times (0.6 \times 0.4 + 0.6 \times 0.4 + 0 \times 0.2) \times (0 \times 0.4 + 0.2 \times 0.4 + 0.2 \times 0.2) \times 0.2
                                         = 0.0026 n.
                     P(X_2=3|Z_2=) = \eta P(Z_2=1|X_2=3) P(Z_3=1|X_2=3) \times P(X_2=3|Z_4)
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⇒ N = 10000 = 1000 = 10000 = 10000

Day 2 is sunny : $\frac{0.0104}{0.013} =$

rainy: $\frac{0}{0.013}$

0.013

0.2

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Day 8 :
   P(X_3 = 1 \mid Z_{3:4}) = \frac{P(X_3 = 1, Z_4 = 3 \mid Z_{2:3})}{P(Z_4 = 3 \mid Z_{2:3})} = \eta P(Z_4 = 3 \mid X_3 = 1, Z_{2:3}) \times P(X_3 = 1 \mid Z_{2:3}).
                                               = n (ZP(Z+=3 | X++i) P(X+=i | X3=1)) P(X3=1 | Z2,3).
                                               = 0.
    P(X_{8}:2|Z_{2:4}) = \eta(Z|P(Z_{4}:3|X_{4}:i)|X_{4}:i|X_{8}:2)) - P(X_{2}:2|Z_{2:3})
                 = n · (0.2×1) v 0.1282
    P(X_3=3|Z_{2:4}) = \eta(\Sigma P(Z_4=3|X_4=i)(X_4=i)X_3=3)) \times P(X_3=3|Z_{2:3})
                     = 0.
    Day 3 is sunny :
           cloudy:
                          1
            rainy:
                          0
                                4
Day 4 : Sunny
                 cloudy
                             0
                             1
                rainy
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(c): By (b)
$$\Rightarrow$$
 the most likely sequence is (hindsight),

02. $p3$ $p4$

Sunny \rightarrow Cloudy \rightarrow rainy.

probability: 0.8 = 80% ϕ