

## Exercise 2: Principle Component Analysis

1.  $f(GR) = \text{tr}(Q^T G^T S G Q) = \text{tr}(Q^T Q \cdot G^T S G) = \text{tr}(G^T S G)$ .  $V$  orthogonal matrix  $Q \in \mathbb{R}^{k \times k}$

2. Denote the SVD of  $\tilde{X}$  by  $\tilde{X} = U \Sigma V^T$ .  $\Sigma = \text{diag}(\lambda_1, \dots, \lambda_d)$ .  $\lambda_1 \geq \dots \geq \lambda_d \geq 0$

thus,  $S = \frac{1}{n-1} V \Sigma^2 U^T$ .  $y = U^T g$ . so,  $\|y\| = 1$

therefore,  $f(g) = \frac{1}{n-1} g^T U \Sigma^2 U^T g = \frac{1}{n-1} \sum_{i=1}^d y_i^2 \lambda_i^2 \leq \frac{1}{n-1} \lambda_1^2$ .

note that, the equivalence condition is  $U^T g = e_1$ ,

i.e  $g = U e_1$ . is the first principal component vector of the data.

3. Note that, the columns of  $U$ ,  $\{u_i = U e_i\}_{i=1}^d$ , are the orthonormal basis of  $\mathbb{R}^d$ .

then,  $\forall g \in \mathbb{R}^d$ . with  $\|g\| = 1$ .  $g = \sum_{i=1}^d a_i u_i$ . and  $\sum_{i=1}^d a_i^2 = 1$ .

since  $\langle g_1, g \rangle = 0$ , we have  $\langle u_1, g \rangle = a_1 = 0$ .

so,  $f(g) = \frac{1}{n-1} \sum_{i=2}^d a_i^2 \lambda_i^2 \leq \frac{1}{n-1} \lambda_2^2$ .

the equivalence condition is  $a_2 = 1$ .

i.e  $g_2 = U e_2$ . the second principal component vector of the data.

4. the first  $k$  principal component vector is  $U e_k$ .

5.  $f(g_k) = \frac{1}{n-1} \lambda_k^2$

6.  $\lambda_1 > \lambda_2 > \dots > \lambda_k$ .

## Exercise 5: Grid World with a Given Policy

1.

0.25	0.05									
0.7	0.8	0.05								
	0.15	0.25								
0.05		0.3	0.15							
		0.7	0.3							
		0.7	0.1		0.7					
			0.7	0.25	0.05					
			0.7	0.7	0.25					
			0.05		0.25					
				0.05	0.25	1				
				0.7			1			

$$1a). P(S_1 = s_i) = \frac{1}{11}$$

$$P(S_2 = s_i) = \frac{1}{7}$$

0.25		0.05
1.55		1.47
0.4		0.325
0.5		0.39
1		0.58
1.5		0.71
1		1.025
1.65		1.815
0.3		0.15
1.1		1.65
1.7		2.85

2. 1a).  $P(S_1 = s_i) = \sum_{j=1}^4 m_{ij}$ .  $\forall i=1, 2, \dots, 11$ .

$$\text{let } u = (1, \dots, 1)^T \in \mathbb{R}^n. \quad P_i = (P(S_t = s_i) \dots P(S_n = s_n))^T, \quad i=1, 2.$$

$$\text{then } p_i = \frac{1}{q} M u. \quad p_i = M p_i = \frac{1}{q} M^T u.$$

$$1b). P(S_t = s_i) = [\frac{1}{q} M^T u]_i = \frac{1}{q} e_i^T M^T u, \quad i=1, 2, \dots, 9.$$

1c).

$$3. V^T(s_1) = 0.9^4 \times 100 = 65.61$$

$$V^T(s_2) = 0.9^3 \times 100 = 72.9$$

$$V^T(s_3) = 0.9^2 \times 100 = 81$$

$$V^T(s_4) = 0.9^3 \times 100 = 72.9$$

$$V^T(s_5) = 0.9 \times 100 = 90.$$

$$V^T(s_6) = 0.9^2 \times 100 = 81$$

$$V^T(s_7) = 0.9 \times 100 = 90.$$

$$V^T(s_8) = 100.$$

$$V^T(s_9) = 0.9^3 \times 100 = 72.9.$$

$$4. P(S_t = s_i) = [M^T p_0]_i = e_i^T M^T p_0, \quad i=1, 2, \dots, 9.$$

## Exercise 7: Value Iteration and Policy Iteration

$$1. V^*(s) = \max_a Q(s, a) = \max_a r(s, a) + \gamma V^*(\delta(s, a)).$$

$$Q(s, a) = r(s, a) + \gamma \max_{a'} Q(\delta(s, a), a').$$

$$V^{k+1}(s) = \max_a Q(s, a) = \max_a r(s, a) + \gamma V^k(\delta(s, a)).$$

$$|V^*(s) - V^{k+1}(s)| = |\max_a [r(s, a) + \gamma V^*(\delta(s, a))] - \max_a [r(s, a) + \gamma V^k(\delta(s, a))]|$$

$$\leq \max_a |[r(s, a) + \gamma V^*(\delta(s, a))] - [r(s, a) + \gamma V^k(\delta(s, a))]|$$

$$= \gamma \max_a |V^*(\delta(s, a)) - V^k(\delta(s, a))| \quad \forall s.$$

$$\Rightarrow \|V^* - V^{k+1}\|_\infty \leq \gamma \|V^* - V^k\|_\infty$$

$$2. 1a) Q^T(s, a) = r(s, a_0) + \gamma r(s, a_1) + \gamma^2 r(s, a_2) + \dots$$

$$s_0 = s, \quad a_0 = 0, \quad \begin{cases} s_t = \delta(s_{t-1}, a_{t-1}), & t=1, 2, \dots \\ a_{t+1} = \pi^*(s_t) \end{cases}$$

$$= r(s_0, a_0) + \gamma Q^T(s_1, a_1)$$

$$= r(s, \alpha) + \gamma Q^{\pi}(s(s, \alpha), \pi(s)).$$

$$V^{\pi}(s) = R(s) + \gamma \sum_{s' \in S} P(s'|s, \pi(s)) V^{\pi}(s'). \quad V^{\pi} = R + \gamma P^{\pi} V^{\pi}.$$

$$1b). V^{\pi'}(s) = R(s) + \gamma \sum_{s' \in S} P(s'|s, \pi'(s)) V^{\pi'}(s').$$

$$\pi'(s) = \operatorname{argmax}_{a \in A} Q^{\pi}(s, a)$$

$$V^{\pi'}(s) = R(s) + \gamma \max_{a \in A} \sum_{s' \in S} P(s'|s, \pi'(s)) V^{\pi'}(s')$$

$$\geq R(s) + \gamma \sum_{s' \in S} P(s'|s, \pi'(s)) V^{\pi'}(s')$$

$$= V^{\pi'}(s).$$