

1. (a).

$$X_1 = A^0 B u_0$$

$$X_2 = A X_1 + B u_1 = A^2 B u_0 + B u_1$$

$$\vdots$$

$$X_k = \sum_{j=0}^{k-1} A^{k-1-j} B u_j$$

$$Y_k = C X_k + D u_k = C \sum_{j=0}^{k-1} A^{k-1-j} B u_j + D u_k$$

$$\text{Let } K_k = \begin{cases} D, & k=0 \\ C A^{k-1} B, & k \geq 1 \end{cases}$$

$$\therefore Y_k = \sum_{l=0}^k K_l u_{k-l}$$

(b).

$$i. \quad X_{k+1} = 0.8 X_k + u_k, \quad Y_k = 1.5 X_k + \delta u_k.$$

$$K_0 = \delta, \quad K_1 = 1.5 \cdot 1 = 1.5, \quad K_2 = 1.5 \cdot 0.8 \cdot 1 = 1.2$$

$$K_3 = 1.5 \cdot 0.8^2 = 0.96, \quad K_4 = 1.5 \cdot 0.8^3 = 0.768$$

$$ii. \quad K_0 = D = 0, \quad K_1 = 1, \quad K_2 = 0.7, \quad K_3 = C A^2 B = [1, 0] \begin{bmatrix} 0.51 & 0.13 \\ 0.26 & 0.38 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix} = 0.51$$

(c).

$$\text{naive: path: } y_0 \rightarrow y_1 \rightarrow \dots \rightarrow y_L : O(L^2)$$

every  $y_k$  need  $O(k)$  computation.

use FFT, parallelize computation. , length:  $O(\log L)$

$$\therefore O(L \log L)$$

(d).

$$\text{naive: } O(n^3 L), \quad \text{divide and conquer: } O(n^{\log_3 7} L)$$

(e).

$$K_k = \begin{cases} D & k=0 \\ C \text{diag}(\lambda_1^{k-1}, \dots, \lambda_n^{k-1}) B, & k \geq 1 \end{cases}$$

$$\therefore O(nL)$$

(f).

$$A = I_n + P P^T$$

$$S_k = P^T A^k P, \quad A^k B = B + P P^T A^{k-1} B = B + P \cdot \frac{P^T A^k B - P^T B}{S_k - I P P^T}$$

$$O(nL).$$

3/a. (i).  $XX^T = \begin{bmatrix} 0.3583 & 0.0015 \\ 0.0015 & 0.0010 \end{bmatrix}$

$$\|W^{(\alpha)}\|_F^2 = 1^2 + 0^2 + 1^2 + 0^2 = 2$$

$$\|W^{(\beta)}\|_F^2 = 1^2 + 0^2 + 0^2 + 0^2 = 1$$

$$\begin{matrix} \alpha & 0 & 2 \\ \beta & 0.001 & 1 \end{matrix}$$

(ii).  $\mathcal{L}_2(W^{(\alpha)}) = 2\lambda$ ,  $\mathcal{L}_2(W^{(\beta)}) = 0.001 + \lambda$   
 $\therefore 2\lambda > \lambda + 0.001$   
 $\therefore \lambda > 0.001$

(b). (i).  $\hat{W} = U \begin{bmatrix} \frac{\sigma_1^2}{\sigma_1^2 + \lambda} & & \\ & \ddots & \\ & & \frac{\sigma_m^2}{\sigma_m^2 + \lambda} \end{bmatrix} U^T$   
(ii).

(c).  $g_i(\lambda) = \frac{\sigma_i^2}{\sigma_i^2 + \lambda}$

1. preserve first 3 directions at at least 80%.

$$g_i(\lambda) \geq 80\%, \quad i=1,2,3.$$

$$\frac{16}{16+\lambda} \geq 0.8 \Rightarrow \lambda \leq 4$$

2. Attenuate remaining 5 directions to at most 50%.

$$g_i(\lambda) \leq 0.5$$

$$g_4(\lambda) \leq 0.5 \Rightarrow \lambda \geq 1.$$

$$\therefore 1 \leq \lambda \leq 4.$$

4/a.

$$m = \frac{1}{n} \sum_{i=1}^n x_i$$

$$m' = \frac{1}{n+1} \sum_{i=1}^{n+1} x_i$$

$$\therefore m' = \frac{nm + x_{n+1}}{n+1}$$

(b).  $A^T = [x_1, x_2, \dots, x_n]$ ,  $y = \begin{bmatrix} y_1 \\ \vdots \\ y_n \end{bmatrix}$

$$\therefore A^T A = \sum_{i=1}^n x_i x_i^T \Rightarrow A^T A + \lambda I = \lambda I + \sum_{i=1}^n x_i x_i^T$$

$$A^T y = \sum_{i=1}^n y_i x_i$$

compute:

(c).  $K^T V, K^T K: O(nd^2)$  ~~compute~~ Invert  $M(d \times d): O(d^3)$   
each query  $\cdot \xi^T (M^T P): O(d^2)$  compute  $M^T P: O(d^3)$

Total:  $O(nd^2)$ .

(d).  $W^* = (K^T K + \lambda I)^{-1} K^T V$

output =  $\xi^T W^*$

$\lambda$ , keys, values, queries are all differentiable.

(e).  $O(d^2)$

(f).

(g).  $0 = \xi^T W^* = \xi^T (K^T K + \lambda I)^{-1} K^T V$

Let:

$$a = (K^T K + \lambda I)^{-1} K^T \xi, \quad \alpha \in \mathbb{R}^n.$$

$$0 = \sum_{i=1}^n \alpha_i V_i$$

$$\therefore W_i = \alpha_i = [(K^T K + \lambda I)^{-1} K^T \xi]_i$$

$$W = (K^T K + \lambda I)^{-1} K^T \xi.$$

5. (a). Grok.

(b). Nobody

(c). 3-4 h