

1) A.) Assume  $m_\alpha = (K^T K + \alpha I)^{-1} K^T d$

Let  $U$  and  $V$  be the matrices containing the singular vectors of  $K$ , and let  $S$  be the diagonal matrix containing the singular values of  $K$ , then:

$$\begin{aligned}
 m_\alpha &= ((USV)^T (USV) + \alpha I)^{-1} (USV)^T d \\
 &= (V^T \underbrace{U^T U}_{I} S V^T + \alpha I)^{-1} V^T U^T d \\
 &= (V^T S V^T + \alpha I)^{-1} V^T U^T d = \frac{U^T d}{(V^T S V^T + \alpha I)} \\
 &= \frac{S^T}{(S^T S V^T + V^T \alpha I)} U^T d = \frac{S^T}{V^T (S^T S + \alpha I)} U^T d \quad \begin{matrix} S^T = S \\ V^T = V^{-1} \end{matrix} \\
 &= V \frac{S}{(S^T S + \alpha I)} U^T d = \boxed{\sum_{i=1}^N \frac{s_i}{s_i^2 + \alpha} (U^T d)_i v_i}
 \end{aligned}$$

For  $K^T K + \alpha I$  we have:

$$\begin{aligned}
 (USV)^T (USV) + \alpha I &= V^T U^T U S V^T + \alpha I \\
 &= V^T S V^T + \alpha I
 \end{aligned}$$

$V^T S V^T$  is the eigen decomposition of the matrix  $K^T K$ , since  $S_N$  is strictly positive, this implies that  $K^T K$  is positive definite.  $\alpha I$  is also positive definite by construction since  $\alpha > 0$ .

Therefore  $K^T K + \alpha I$  is positive definite, and also invertible.



B.)  $\min_m \frac{1}{2} \|Km - d\|^2 + \frac{\alpha}{2} \|m\|^2$

$$J(m) = \frac{1}{2} (Km - d)^T (Km - d) + \frac{\alpha}{2} m^T m$$

We consider:

$$\epsilon \in \mathbb{R}, \tilde{m} \in \mathbb{R}^n$$

$$\begin{aligned} f(\epsilon) &= \frac{1}{2} (K(m + \epsilon \tilde{m}) - d)^T (K(m + \epsilon \tilde{m}) - d) + \frac{\alpha}{2} (m + \epsilon \tilde{m})^T (m + \epsilon \tilde{m}) \\ &= \frac{1}{2} \left[ (K(m + \epsilon \tilde{m}))^T K(m + \epsilon \tilde{m}) - d^T K(m + \epsilon \tilde{m}) - (K(m + \epsilon \tilde{m}))^T d + d^T d \right] + \frac{\alpha}{2} (m + \epsilon \tilde{m})^T (m + \epsilon \tilde{m}) \\ &= \frac{1}{2} (m + \epsilon \tilde{m})^T K^T K (m + \epsilon \tilde{m}) - \frac{1}{2} d^T K(m + \epsilon \tilde{m}) - \frac{1}{2} (m + \epsilon \tilde{m})^T K^T d + \frac{1}{2} d^T d + \frac{\alpha}{2} (m + \epsilon \tilde{m})^T (m + \epsilon \tilde{m}) \\ &= \frac{1}{2} (m + \epsilon \tilde{m})^T K^T K (m + \epsilon \tilde{m}) - (m + \epsilon \tilde{m})^T K^T d + \frac{1}{2} d^T d + \frac{\alpha}{2} (m + \epsilon \tilde{m})^T (m + \epsilon \tilde{m}) \\ &= \frac{1}{2} \left[ m^T K^T K m + m^T K^T K \epsilon \tilde{m} + \epsilon \tilde{m}^T K^T K m + \epsilon \tilde{m}^T K^T K \epsilon \tilde{m} \right] - m^T K^T d - \epsilon \tilde{m}^T K^T d + \frac{1}{2} d^T d \\ &\quad + \frac{\alpha}{2} \left[ m^T m + m^T \epsilon \tilde{m} + \epsilon \tilde{m}^T m + \epsilon \tilde{m}^T \epsilon \tilde{m} \right] \\ &= \frac{1}{2} m^T K^T K m + \epsilon \tilde{m}^T K^T K m + \frac{1}{2} \epsilon^2 \tilde{m}^T K^T K \tilde{m} - m^T K^T d - \epsilon \tilde{m}^T K^T d + \frac{1}{2} d^T d + \frac{\alpha}{2} m^T m \\ &\quad + \alpha \epsilon \tilde{m}^T m + \frac{\alpha}{2} \epsilon^2 \tilde{m}^T \tilde{m} \end{aligned}$$

$$\frac{d f(\epsilon)}{d \epsilon} = \frac{d J(m + \epsilon \tilde{m})}{d \epsilon} = \frac{d J}{d m_i} \quad \text{where } \tilde{m} = e_i, \tilde{m} = \begin{bmatrix} 0 \\ \vdots \\ 1 \\ \vdots \\ 0 \end{bmatrix} \leftarrow \begin{matrix} \text{ith} \\ \text{element} \end{matrix}$$

$$= 0 + \tilde{m}^T K^T K m + \epsilon \tilde{m}^T K^T K \tilde{m} - 0 - \tilde{m}^T K^T d + 0 + 0 + \alpha \tilde{m}^T m + \alpha \epsilon \tilde{m}^T \tilde{m}$$

$$\left. \frac{d f(\epsilon)}{d \epsilon} \right|_{\epsilon=0} = \tilde{m}^T K^T K m + 0 - \tilde{m}^T K^T d + \alpha \tilde{m}^T m + 0 = \tilde{m}^T (K^T K m - K^T d + \alpha m)$$

↑  
for one element of m

$$\frac{d J}{d m} = K^T K m - K^T d + \alpha I m$$

← Add for all elements

Now we find min of  $J(m)$

$$\frac{d J}{d m} = 0 = K^T K m_\alpha - K^T d + \alpha m_\alpha$$

$$\Rightarrow K^T K m_\alpha + \alpha I m_\alpha = K^T d$$

$$(K^T K + \alpha I) m_\alpha = K^T d$$

$$m_\alpha = (K^T K + \alpha I)^{-1} K^T d$$