

Exercise Sheet 4

1 Sparse Coding

a) f is convex \rightarrow setting gradient to 0 will yield minimum

$$f = \frac{1}{N} \sum_{i=1}^N \|x_i - W s_i\|^2 + \lambda \frac{1}{N} \sum_{i=1}^N \|s_i\|_1 + \eta \|W\|_F^2$$

$$\frac{\partial f}{\partial W} = \frac{\partial}{\partial W} \frac{1}{N} \sum_{i=1}^N \|x_i - W s_i\|^2 + \eta \|W\|_F^2 \stackrel{!}{=} 0$$

$$\Leftrightarrow \frac{\partial}{\partial W} \frac{1}{N} \sum_{i=1}^N (x_i - W s_i)^T (x_i - W s_i) + \eta \|W\|_F^2 = 0$$

$$\Leftrightarrow \frac{1}{N} \sum_{i=1}^N -2(x_i - W s_i) s_i^T + 2\eta W = 0$$

$$\Leftrightarrow \frac{1}{N} \sum_{i=1}^N -(x_i - W s_i) s_i^T + \eta W = 0$$

$$\Leftrightarrow \frac{1}{N} \sum_{i=1}^N W(\eta \mathbb{1} + s_i s_i^T) = \frac{1}{N} \sum_{i=1}^N x_i s_i^T$$

$$\Leftrightarrow W(\sum_{ss} + \eta \mathbb{1}) = \sum_{xs} \Leftrightarrow W = \sum_{xs} (\sum_{ss} + \eta \mathbb{1})^{-1}$$

A) $\frac{\partial f}{\partial s_i} \stackrel{!}{=} 0 \Rightarrow \frac{\partial}{\partial s_i} \frac{1}{N} \sum_{i=1}^N \|x_i - W s_i\|^2 + \lambda \frac{1}{N} \sum_{i=1}^N q_i^T s_i \stackrel{!}{=} 0$

$$\Leftrightarrow \frac{1}{N} \sum_{i=1}^N \frac{\partial}{\partial s_i} (x_i - W s_i)^T (x_i - W s_i) + \lambda \frac{1}{N} \sum_{i=1}^N q_i^T s_i = 0$$

$$\Leftrightarrow \frac{2}{N} W^T (x_i - W s_i) = \frac{\lambda}{N} q_i^T$$

$$\Leftrightarrow W^T x_i - W^T W s_i = \frac{\lambda}{2} q_i^T$$

$$\Leftrightarrow s_i = (W^T W)^{-1} (W^T x_i - \frac{\lambda}{2} q_i^T)$$

2 Auto-Encoders

$$\begin{aligned} \text{a)} \quad \argmin_{\alpha, u} \frac{1}{N} \sum_{i=1}^N \|X_i - \hat{X}_i\|^2 &= \argmin_{\alpha, u} \frac{1}{N} \sum_{i=1}^N \|X_i - u(u^T X_i)\|^2 \\ &= \argmin_{\alpha, u} \frac{1}{N} \sum_{i=1}^N \|X_i - \alpha u (\alpha u^T X_i)\|^2 \\ &= \argmin_{\alpha, u} \frac{1}{N} \sum_{i=1}^N \|X_i - \alpha^2 u u^T X_i\|^2 \\ &= \argmin_{\alpha, u} \frac{1}{N} \sum_{i=1}^N (X_i^T - \alpha^2 u u^T X_i)^T (X_i - \alpha^2 u u^T X_i) \\ &= \argmin_{\alpha, u} \frac{1}{N} \sum_{i=1}^N (\alpha^4 - 2\alpha^2) (u^T X_i; X_i^T u) \\ &= \argmin_{\alpha, u} (\alpha^4 - 2\alpha^2) (u^T (\frac{1}{N} \sum_{i=1}^N X_i X_i^T) u) \\ &= \argmin_{\alpha, u} (\alpha^4 - 2\alpha^2) (u^T S u) = \argmax_{\alpha, u} u^T S u \end{aligned}$$