Proposition 55.1. The limit of the matrix $(X^{\top}X + KI_n)^{-1}X^{\top}$ when K > 0 goes to zero is the pseudo-inverse X^+ of X.

Proof. To show this let $X = V \Sigma U^{\top}$ be a SVD of X. Then

$$(X^{\top}X + KI_n) = U\Sigma^{\top}V^{\top}V\Sigma U^{\top} + KI_n = U(\Sigma^{\top}\Sigma + KI_n)U^{\top},$$

SO

$$(X^{\top}X + KI_n)^{-1}X^{\top} = U(\Sigma^{\top}\Sigma + KI_n)^{-1}U^{\top}U\Sigma^{\top}V^{\top} = U(\Sigma^{\top}\Sigma + KI_n)^{-1}\Sigma^{\top}V^{\top}.$$

The diagonal entries in the matrix $(\Sigma^{\top}\Sigma + KI_n)^{-1}\Sigma^{\top}$ are

$$\frac{\sigma_i}{\sigma_i^2 + K}$$
, if $\sigma_i > 0$,

and zero if $\sigma_i = 0$. All nondiagonal entries are zero. When $\sigma_i > 0$ and K > 0 goes to 0,

$$\lim_{K \to 0} \frac{\sigma_i}{\sigma_i^2 + K} = \sigma_i^{-1},$$

so

$$\lim_{K \to 0} (\Sigma^{\top} \Sigma + K I_n)^{-1} \Sigma^{\top} = \Sigma^{+},$$

which implies that

$$\lim_{K \to 0} (X^{\top} X + K I_n)^{-1} X^{\top} = X^+.$$

The dual function of the first formulation of our problem is a constant function (with value the minimum of J) so it is not useful, but the second formulation of our problem yields an interesting dual problem. The Lagrangian is

$$L(\xi, w, \lambda) = \xi^{\mathsf{T}} \xi + K w^{\mathsf{T}} w + (y - X w - \xi)^{\mathsf{T}} \lambda$$

= $\xi^{\mathsf{T}} \xi + K w^{\mathsf{T}} w - w^{\mathsf{T}} X^{\mathsf{T}} \lambda - \xi^{\mathsf{T}} \lambda + \lambda^{\mathsf{T}} y$

with $\lambda, \xi, y \in \mathbb{R}^m$. The Lagrangian $L(\xi, w, \lambda)$, as a function of ξ and w with λ held fixed, is obviously convex, in fact strictly convex.

To derive the dual function $G(\lambda)$ we minimize $L(\xi, w, \lambda)$ with respect to ξ and w. Since $L(\xi, w, \lambda)$ is (strictly) convex as a function of ξ and w, by Theorem 40.13(4), it has a minimum iff its gradient $\nabla L_{\xi,w}$ is zero (in fact, by Theorem 40.13(2), a unique minimum since the function is strictly convex). Since

$$\nabla L_{\xi,w} = \begin{pmatrix} 2\xi - \lambda \\ 2Kw - X^{\top}\lambda \end{pmatrix},$$

we get

$$\lambda = 2\xi$$

$$w = \frac{1}{2K} X^{\top} \lambda = X^{\top} \frac{\xi}{K}.$$