

1. A problem of the form *find* x , when $m = Ax + \epsilon$ is given is called an inverse problem if it is *not* well-posed in the sense of Hadamard.
 - (a) Explain Hadamard's definition of a well-posed problem.
 - (b) What are the crucial properties of the matrices U , D and V in the singular value decomposition $A = UDV^T$?
 - (c) Give the formula for truncated SVD solution.
2. Consider the Tikhonov solution $x_\delta = \arg \min_x \{\|Ax - m\|^2 + \delta\|L(x - x^*)\|^2\}$.
 - (a) What is the role of the term $\|Ax - m\|^2$?
 - (b) What are the roles of L and x^* ?
 - (c) Write the problem in stacked form.
3. In total variation regularization we take a difference matrix L and consider the minimization problem

$$\arg \min_{x \in \mathbb{R}^n} \left(\|Ax - m\|_2^2 + \delta \|Lx\|_1 \right). \quad (1)$$

- (a) What is the goal of total variation regularization?
- (b) Show that (1) is equivalent to the problem

$$\arg \min_x \left(\|Ax\|_2^2 - 2m^T Ax + \beta \mathbf{1}^T V_+ + \beta \mathbf{1}^T V_- \right) \quad (2)$$

with suitable definition of β , $\mathbf{1}$, V_+ and V_- . Hint: write $Lx = V_+ - V_-$.

- (c) Convert (2) to quadratic standard form $\arg \min_y \{\frac{1}{2}y^T Hy + f^T y\}$ with appropriate definition of y and equality and inequality restraints.
4. Consider the measurement model $m = Ax + \epsilon$, where the size of A is $n \times n$.
 - (a) Assume that each component of the noise vector ϵ is Cauchy distributed: $\epsilon_j \sim \pi^{-1}(1 + \epsilon_j^2)^{-1}$. Write down the likelihood distribution $p(m|x)$.
 - (b) Assume that we know *a priori* that the components of the vector x are non-negative and less than 1. Define a prior distribution expressing this information.
 - (c) Write the posterior distribution $p(x|m)$ using Bayes formula and the prior and likelihood distributions in (a) and (b). (Normalization constant is not needed.)