

name	problem	$f$	$g$	$\mathcal{A}(\mathbf{x})$	method (complexity)
Robust Regression	$\min \ \mathbf{Ax} - \mathbf{b}\ _1$	$\ \mathbf{Ax} - \mathbf{b}\ _1$ $\ \mathbf{x} - \mathbf{b}\ _1$	0 0	$\mathbf{x}$ $\mathbf{Ax}$	
$\ell_1$ -Regularized Robust Regression	$\min \ \mathbf{Ax} - \mathbf{b}\ _1 + \ \mathbf{x}\ _1$	$\ \mathbf{Ax} - \mathbf{b}\ _1 + \ \mathbf{x}\ _1$ $\ \mathbf{Ax} - \mathbf{b}\ _1$ $\ \mathbf{x} - \mathbf{b}\ _1$	0 $\ \mathbf{x}\ _1$ $\ \mathbf{x}\ _1$	$\mathbf{x}$ $\mathbf{x}$ $\mathbf{Ax}$	
$\ell_2$ -Regularized Robust Regression	$\min \ \mathbf{Ax} - \mathbf{b}\ _1 + \ \mathbf{x}\ _2^2$	$\ \mathbf{Ax} - \mathbf{b}\ _1$ $\ \mathbf{x} - \mathbf{b}\ _1$	$\ \mathbf{x}\ _2^2$ $\ \mathbf{x}\ _2^2$	$\mathbf{x}$ $\mathbf{Ax}$	
$R$ -Regularized Robust Regression	$\min \ \mathbf{Ax} - \mathbf{b}\ _1 + \sum \frac{1}{x_i}$ s.t. $\mathbf{x} > \mathbf{0}$	$\ \mathbf{Ax} - \mathbf{b}\ _1$ $\ \mathbf{x} - \mathbf{b}\ _1$	$\sum \frac{1}{x_i}$ $\sum \frac{1}{x_i}$	$\mathbf{x}$ $\mathbf{Ax}$	
$\ell_1$ -Reg. LS	$\min \ \mathbf{Ax} - \mathbf{b}\ _2^2 + \ \mathbf{x}\ _1$	$\ \mathbf{Ax} - \mathbf{b}\ _2^2$	$\ \mathbf{x}\ _1$	$\mathbf{x}$	
$\ell_1$ -Con. LS	$\min\{\ \mathbf{Ax} - \mathbf{b}\ _2^2 : \ \mathbf{x}\ _1 \leq \alpha\}$	$\ \mathbf{Ax} - \mathbf{b}\ _2^2$	$\delta_{B_1[0, \alpha]}$	$\mathbf{x}$	

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Convex Feasibility ( $C \cap D \neq \emptyset$ )	$\min \max\{d_C(\mathbf{x}), d_D(\mathbf{x})\}$ $\min \delta_C(\mathbf{x}) + \delta_D(\mathbf{x})$	$\max\{d_C(\mathbf{x}), d_D(\mathbf{x})\}$ $\delta_C(\mathbf{x})$	$0$ $\delta_D(\mathbf{x})$	$\mathbf{x}$ $\mathbf{x}$	
Projection on Convex Sets	$\min \ \mathbf{x} - \mathbf{a}\ ^2$ s.t. $\mathbf{x} \in \cap_{i=1}^m C_i$	$\sum_{i=1}^m \delta_{C_i}(\mathbf{x}_i)$ $\sum_{i=1}^m \delta_{C_i}(\mathbf{x})$	$\ \mathbf{x} - \mathbf{a}\ ^2$ $\ \mathbf{x} - \mathbf{a}\ ^2$	$(\mathbf{x}, \dots, \mathbf{x})$ $\mathbf{x}$	
Basis Pursuit	$\min\{\ \mathbf{x}\ _1 : \mathbf{Ax} = \mathbf{b}\}$	$\ \mathbf{x}\ _1$ $\delta_{\{\mathbf{b}\}}(\mathbf{x})$	$\delta_{\{\mathbf{x}:\mathbf{Ax}=\mathbf{b}\}}$ $\ \mathbf{x}\ _1$	$\mathbf{x}$ $\mathbf{Ax}$	
Poisson Inv.	$\min_{\mathbf{x} \geq 0} \sum b_i \ln \frac{b_i}{(\mathbf{Ax})_i} + (\mathbf{Ax})_i$	$-\sum (b_i \ln x_i + x_i)$	$0$	$\mathbf{Ax}$	
Matrix Compl.	$\min \ \mathcal{M}(\mathbf{X}) - \mathbf{b}\ _F^2 + \ \mathbf{X}\ _*$ s.t. $\mathbf{X} \in \mathbb{R}^{n \times d}$	$\ \mathcal{M}(\mathbf{x}) - \mathbf{b}\ _F^2$ $\ \mathbf{X} - \mathbf{b}\ _F^2$	$\ \mathbf{X}\ _*$ $\ \mathbf{X}\ _*$	$\mathbf{x}$ $\mathcal{M}(\mathbf{X})$	
TV Denoising	$\min\{0.5\ \mathbf{x} - \mathbf{d}\ _2^2 + \ \mathbf{Dx}\ _1\}$	$0.5\ \mathbf{x}\ _1$	$0.5\ \mathbf{x} - \mathbf{d}\ _2^2$	$\mathbf{Dx}$	
TV Deblurring	$\min\{\ \mathbf{Cx} - \mathbf{b}\ _2^2 + \ \mathbf{Dx}\ _1\}$	$\ \mathbf{x} - \mathbf{b}\ _2^2 + \ \mathbf{y}\ _1$	$0$	$(\mathbf{Cx}, \mathbf{Dx})$	