

Contents

Preface	vii
1 Characteristics of Inverse Problems	1
1.1 Preliminaries	1
1.2 The least squares estimator	6
1.3 The statistical properties of \mathbf{x}_{LS} and ill-posedness	8
1.4 An illustrative example	11
Exercises	12
2 Regularization by Spectral Filtering	15
2.1 Spectral filtering methods	15
2.2 Regularization parameter selection methods	19
2.3 Periodic and data-driven boundary conditions	25
Exercises	29
3 Two-Dimensional Test Cases	33
3.1 Two-dimensional image deblurring	33
3.2 Computed tomography	42
3.3 The preconditioned conjugate gradient iteration	45
Exercises	47
4 Bayes' Law, Markov Random Field Priors, and MAP Estimation	53
4.1 Bayes' law and regularization	53
4.2 Choosing $p(\mathbf{x} \delta)$: Gaussian Markov random fields	54
4.3 Choosing $p(\mathbf{x} \delta)$: Laplace Markov random fields	65
4.4 The infinite-dimensional limit	68
Exercises	72
5 Markov Chain Monte Carlo Methods for Linear Inverse Problems	77
5.1 Sampling from high-dimensional Gaussian random vectors	77
5.2 Hierarchical modeling of λ and δ and sampling from $p(\mathbf{x}, \lambda, \delta \mathbf{b})$	81
5.3 Alternative MCMC methods for sampling from $p(\mathbf{x}, \lambda, \delta \mathbf{b})$	89
Exercises	99
6 Markov Chain Monte Carlo Methods for Nonlinear Inverse Problems	105
6.1 A general setup for nonlinear inverse problems	105
6.2 Levenburg–Marquardt nonlinear least squares optimization	106

6.3	Randomize-then-optimize as a proposal for Metropolis–Hastings	109
6.4	Nonlinear test cases	112
6.5	Hierarchical modeling of λ and δ and sampling from $p(\mathbf{x}, \lambda, \delta \mathbf{b})$	118
	Exercises	122
	Bibliography	125
	Index	131