

Module – 3.6

EXAMPLES OF STATIC INVERSE PROBLEMS

S. Lakshmivarahan

School of Computer Science

University of Oklahoma

Norman, Ok – 73069, USA

varahan@ou.edu

THREE EXAMPLES

- Recovery of vertical temperature profile of the atmosphere from satellite radiance measurement – linear problem
- 1-D Spatial linear and 2-D bilinear interpolation – linear problem
- A nonlinear least squares problem

VERTICAL TEMPERATURE PROFILE

- Problem is to retrieve the vertical temperature profile of the atmosphere from satellite radiance measurements

PROBLEM 1: SATELLITE RADIANCE – A MODEL

- Energy R_f received by a satellite at a frequency, f is related to the vertical temperature profile, $T(p)$ at the pressure level p of the atmosphere through a formula given by

$$R_f = \exp[-\gamma_f] + \int_0^1 T(p)W(p, \gamma_f)dp \quad \rightarrow (1)$$

where $W(p, \gamma_f)$ is the weight function given by

$$W(p, \gamma_f) = p\gamma_f \exp(\gamma_f p) \quad \rightarrow (2)$$

and γ_f is a constant that depends on f

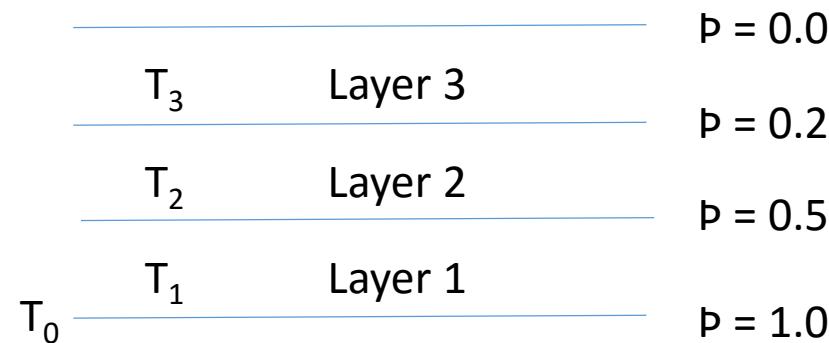
INPUT DATA

- The values of f and γ_f relevant to the problem are given by

| i | 1 | 2 | 3 | 4 | 5 |
|----------------|-----------------|-----------------|-----------------|-----------------|-----------------|
| f_i | 0.9 | 1.0 | 1.1 | 1.2 | 1.3 |
| γ_{f_i} | $\frac{1}{0.9}$ | $\frac{1}{0.7}$ | $\frac{1}{0.5}$ | $\frac{1}{0.3}$ | $\frac{1}{0.2}$ |

A DISCRETE MODEL

- The problem is to recover the function $T(p)$ from a set of discrete measurements of R_{fi} , $1 \leq i \leq 5$ – an underdetermined system
- We discretize the atmosphere by considering it as a 3-layered system



- T_0 is the temperature of the earth's surface
- T_i is the average temperature of the layer i , $1 \leq i \leq 3$
- Layers are bounded by isobaric surfaces at $p = 1.0, 0.5, 0.2$, and 0.0

A DISCRETE RELATION

- Discretizing (1) for the frequency f_i , $1 \leq i \leq 5$ using the 3-layer approximation:

$$\begin{aligned} Z_i &= R_{f_i} - \exp[-\gamma_{f_i}] \\ &= T_1 \int_{0.5}^{1.0} p \gamma_{f_i} \exp[-p \gamma_{f_i}] dp = T_1 a_{i1} \\ &\quad + T_2 \int_{0.2}^{0.5} p \gamma_{f_i} \exp[-p \gamma_{f_i}] dp + T_2 a_{i2} \\ &\quad + T_3 \int_{0.0}^{0.2} p \gamma_{f_i} \exp[-p \gamma_{f_i}] dp + T_3 a_{i3} \end{aligned}$$

where the constant a_{i1}, a_{i2}, a_{i3} are the numerical values of the respective integrals obtained using the input data and by integration by parts

A LINEAR MODEL

- By collating the five linear relations between Z_i and T_1, T_2, T_3 , for each frequency f_i , $1 \leq i \leq 5$, we get a linear model:

$$\begin{bmatrix} Z_1 \\ Z_2 \\ Z_3 \\ Z_4 \\ Z_5 \end{bmatrix} = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \\ a_{41} & a_{42} & a_{43} \\ a_{51} & a_{52} & a_{53} \end{bmatrix} \begin{bmatrix} T_1 \\ T_2 \\ T_3 \end{bmatrix} \quad \rightarrow (3)$$

$$\text{Or } Z = Hx, Z \in \mathbb{R}^5, H \in \mathbb{R}^{5 \times 3}, T \in \mathbb{R}^3 \quad \rightarrow (4)$$

A TWIN EXPERIMENT – COMPUTER PROJECT: GENERATE OBSERVATION

- Set $T_1 = 0.9, T_2 = 0.85, T_3 = 0.875$, set $\bar{x} = (T_1, T_2, T_3)^T$
- Evaluate a_{i1}, a_{i2}, a_{i3} for $1 \leq i \leq 5$ using the input data
- This gives the matrix H
- Compute $\bar{Z} \in R^5$ using (3) as $\bar{Z} = H\bar{x}$
- Now generate an observation noise vector $V \in R^5$ such that $V \sim N(0, \sigma^2 I_5)$ where I_5 is the identity matrix of order 5 and σ^2 is the common variance of the radiance measurement
- Let $Z = \bar{Z} + V$, be the noisy observation

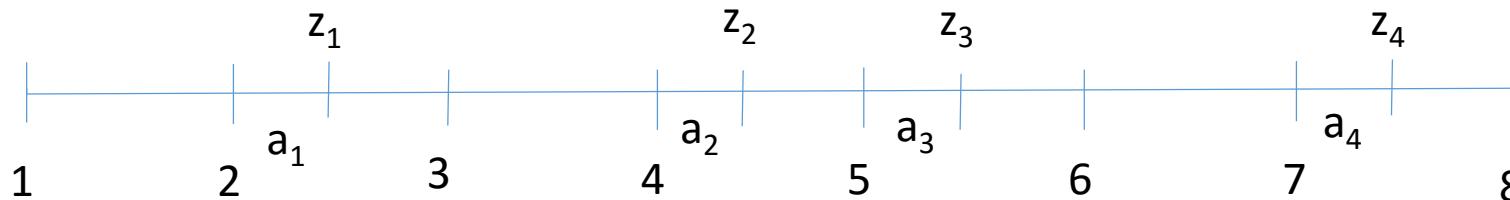
A TWIN EXPERIMENT – RECOVER T FROM NOISY OBSERVATION

- Using this noisy observation vector Z , now solve the overdetermined linear least squares problem $Z = Hx$ and recover x
- Compute the $\|x - x_{LS}\|_2$ and plot it as a function of σ^2 by repeatedly solving the problem for $\sigma^2 = 0.0, 0.1, 0.4, 0.8, 1.0, 1.2$
- Comment on your result

PROBLEM 2: SPATIAL INTERPOLATION – 1-D

- Consider a uniform spatial computational grid in 1-D with n points:

$$n = 8$$



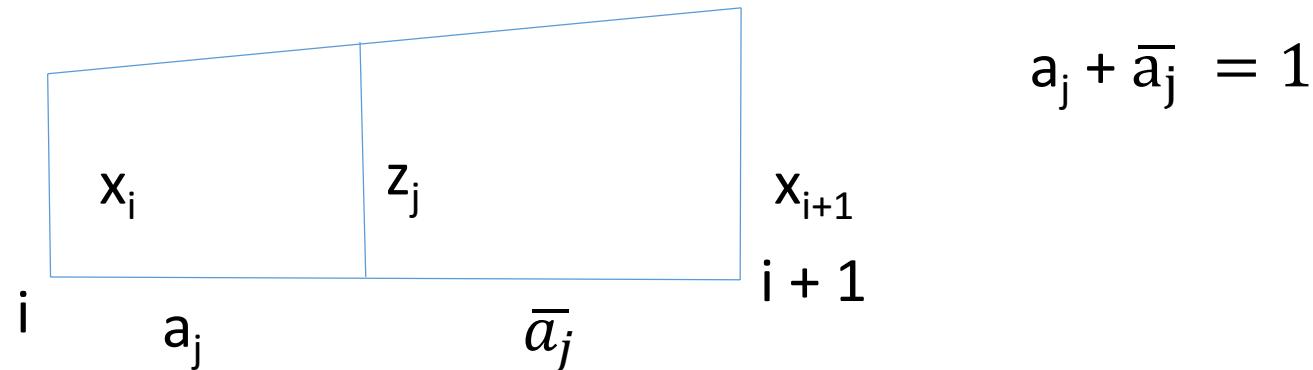
- The grid interval is assumed to be unity
- Let $x = (x_1, x_2, x_3, \dots, x_n)^T \in R^n$ be the unknown state vector
- Let z_1, z_2, \dots, z_m be the m observations of a scalar field variable such as, say temperature, concentration of a pollutant, to name a few, where $m < n$

DISTRIBUTION OF THE OBSERVATIONS

- Let the j^{th} observation z_j be contained in the interval $[i, i + 1]$
- Referring to the Figure above, $m = 4$, $n = 8$, z_1 is in $[2, 3]$, z_2 is in $[4, 5]$, z_3 is in $[5, 6]$ and z_4 is in $[7, 8]$
- Problem: Given $Z \in R^m$, find $x \in R^n$ where Z and x refer to the same quantities such as temperature, concentration, etc

A LINEAR INTERPOLATION

- Consider the interval $[i, i + 1]$ containing z_j



- a_j is the distance of z_j from the left end i
- Relate x_i , x_{i+1} and z_j using a simple linear relation as:

$$\frac{z_j - x_i}{a_j} = \frac{x_{i+1} - z_j}{\bar{a}_j} \quad \rightarrow (4)$$

- That is, $z_j = \bar{a}_j x_i + a_j x_{i+1}$ $\rightarrow (5)$

A LINEAR INVERSE PROBLEM: UNDERDETERMINED CASE

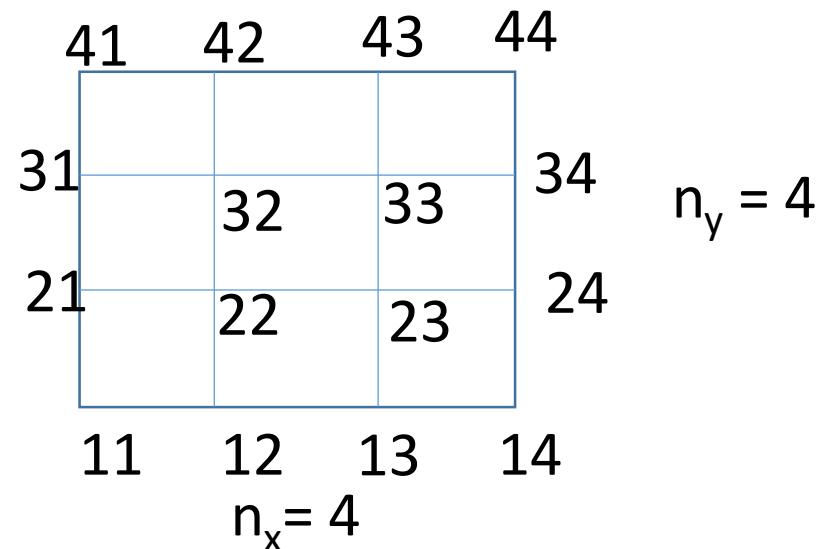
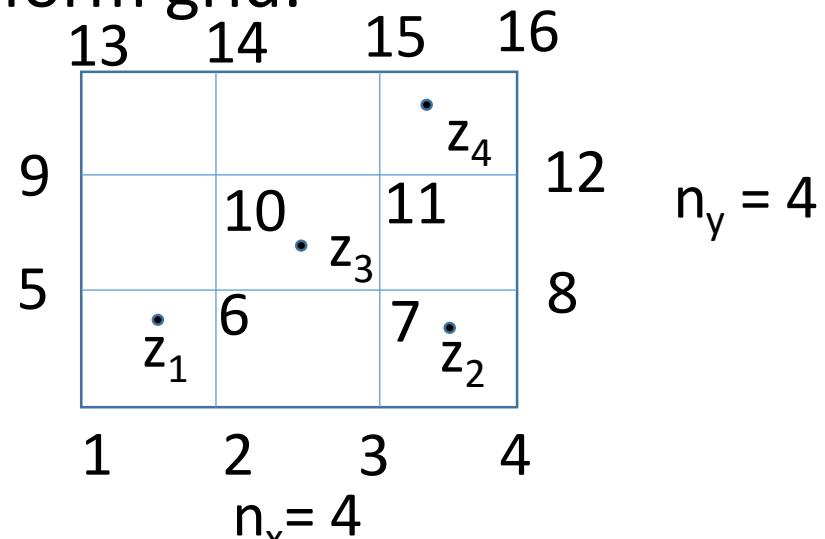
- Applying (5) to each of the $m = 4$ observations on the uniform grid with $n = 8$ points:

$$\begin{bmatrix} Z_1 \\ Z_2 \\ Z_3 \\ Z_4 \end{bmatrix} = \begin{bmatrix} 1 & 2 & 3 & 4 & 5 & 6 & 7 & 8 \\ 0 & \bar{a}_1 & a_1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & \bar{a}_2 & a_2 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & \bar{a}_3 & a_3 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & \bar{a}_4 & a_4 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \\ x_6 \\ x_7 \\ x_8 \end{bmatrix} \rightarrow (7)$$

- That is, 1-D interpolation matrix H is such that row sum is 1 and $Z = Hx$
- We can solve for $x_{LS} = H^T(HH^T)^{-1}Z$
- We can estimate the temperature, concentration at the computational grid from the observation

SPATIAL INTERPOLATION – 2D

- Consider 2-D version with $n = n_x n_y$ grid points arranged in an n_x by n_y uniform grid:



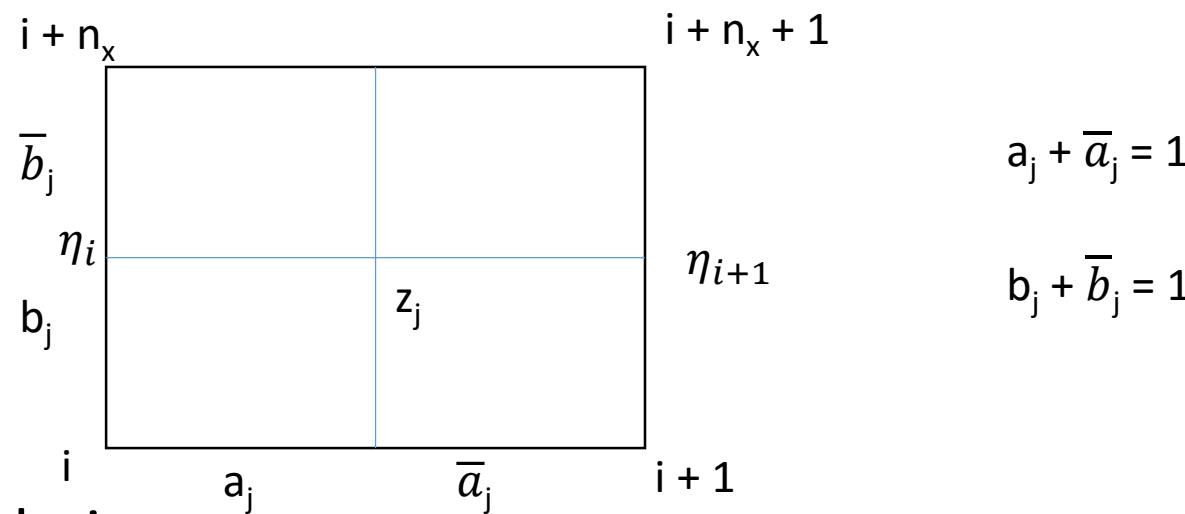
- The left numbering is row major order scheme and the right is the standard (i, j) notation
- The node label k in row major order related to the (i, j) scheme as

$$k = (i - 1)n_x + j$$

- With $n_x = 4$, the node label 7 correspond to $(2,3)$ since $7 = (2-1)*4+3$

A BILINEAR INTERPOLATION

- Let the j^{th} observation z_j be contained in the 2D-grid box whose south-east corner node has label i :



- By 1-D linear interpolation:

$$z_j = \bar{a}_j \eta_i + a_j \eta_{i+1} \quad \rightarrow (7)$$

- Again, by 1-D linear interpolation

$$\eta_i = x_i \bar{b}_j + x_{i+n_x} b_j \quad \rightarrow (8)$$

$$\eta_{i+1} = x_{i+1} \bar{b}_j + x_{i+n_x+1} b_j \quad \rightarrow (9)$$

A LINEAR INVERSE PROBLEM

- Substituting (8) – (9) in (7) and simplifying

$$z_j = \bar{a}_j \bar{b}_j x_i + a_j \bar{b}_j x_{i+1} + \bar{a}_j b_j x_{i+n_x} + a_j b_j x_{i+n_x+1} \quad \rightarrow (10)$$

- By collating the four relations for the four observation in the 2-D and:

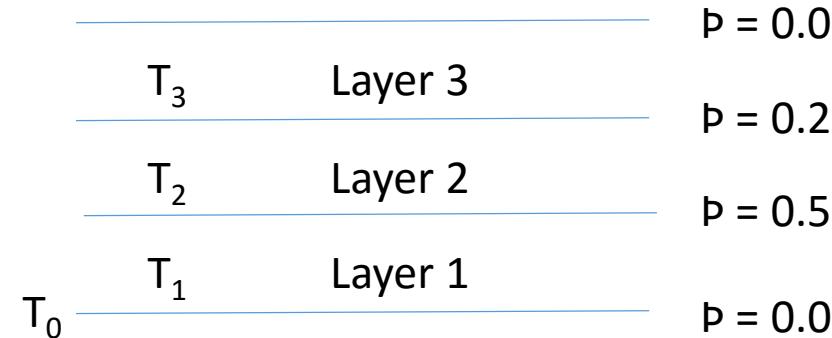
$$\begin{bmatrix} z_1 \\ z_2 \\ z_3 \\ z_4 \end{bmatrix} = \left[\begin{array}{cccc|cccc|cccc|cccc} 1 & 2 & 3 & 4 & 5 & 6 & 7 & 8 & 9 & 10 & 11 & 12 & 13 & 14 & 15 & 16 \\ * & * & 0 & 0 & * & * & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & * & * & 0 & 0 & * & * & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & * & * & 0 & 0 & * & * & 0 & 0 & 0 & 0 & 0 \end{array} \right] \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \\ \vdots \\ x_8 \\ x_9 \\ \vdots \\ x_{12} \\ x_{13} \\ \vdots \\ x_{16} \end{bmatrix}$$

* - represents non-zero element based on (10)

- The 2-D interpolation matrix is such that the row sum is 1 and $Z = Hx$
- Hence $x_{LS} = H^T(HH^T)^{-1}Z$ is the optimal estimate

PROBLEM 3: A NON LINEAR PROBLEM

- Consider a three layered atmosphere



- Let $T(p) = x_1(p - x_2)^2 + x_3, 0 \leq p \leq 1$ $\rightarrow (9)$
- Let $x = (x_1, x_2, x_3)^T \in \mathbb{R}^3$ be the unknown

RELATION BETWEEN TEMPERATURE AND RADIANCE

- The observations are measures of overlapping fractions of the area under the curve:

$$\overline{Z_{ij}} = \int_{\beta_i}^{\beta_j} T(\beta) d\beta \quad \rightarrow (10)$$

- The observations are given by

| β_i | β_j | $\overline{Z_{ij}}$ |
|-----------|-----------|---------------------|
| 0.00 | 0.25 | 0.21 |
| 0.20 | 0.50 | 0.15 |
| 0.30 | 0.70 | 0.51 |
| 0.6 | 0.80 | 0.11 |

THE MODEL EQUATIONS

- After integration:

$$\begin{aligned}\overline{Z_{ij}} &= \int_{p_i}^{p_j} [x_1(p - x_2)^2 + x_3] dp \\ &= \frac{x_1}{3}[(p_j - x_2)^3 - (p_i - x_2)^3] + x_3(p_j - p_i)\end{aligned}$$

- Referring to the Table in slide 19:

$$z_1 = 0.21 = \frac{x_1}{3}[(0.25 - x_2)^3 - x_2^3] + 0.25x_3 = h_1(x)$$

$$z_2 = 0.15 = \frac{x_1}{3}[(0.5 - x_2)^3 - (0.2 - x_2)^3] + 0.3x_3 = h_2(x)$$

$$z_3 = 0.51 = \frac{x_1}{3}[(0.7 - x_2)^3 - (0.3 - x_2)^3] + 0.4x_3 = h_3(x)$$

$$z_4 = 0.11 = \frac{x_1}{3}[(0.8 - x_2)^3 - (0.6 - x_2)^3] + 0.2x_3 = h_4(x)$$

NONLINEAR INVERSE PROBLEM

- Let $Z = h(x)$ with $Z = (z_1, z_2, z_3, z_4)^\top$
$$h(x) = (h_1(x), h_2(x), h_3(x), h_4(x))^\top$$
- Compute $r(x) = Z - h(x)$
- compute $f(x) = (Z - h(x))^\top (Z - h(x))$
- Set $\nabla_x f(x) = 0$ and solve for x
- Check if $\nabla_x^2 f(x)$ is PD

APPROXIMATIONS

- Compute the Jacobian $D_x(h)$ and $D_x^2(h, y)$
- Build first and second order approximation to $h(x)$
- Solve the minimization arising from the first and second-order approximation

EXERCISES

- 11.1) (a) Compute the solution of $\nabla_x f(x) = 0$ for the non linear problem described in slides 18-21, by using the nonlinear solvers in MATLAB
(b) Evaluate the Hessian $\nabla_x^2 f(x)$ at each of the solution obtained in (a) and find the maxima and minima of $f(x)$
- 11.2) (a) Compute the Jacobian and the Hessian of $h(x)$ described in slide 21
(b) Using these develop a first order and second order approximation to $f(x)$
(c) Starting from $x_c = (1, 1, 1)^T$, iterate twice and comment on the progress of your algorithm

REFERENCES

- This module follows closely chapters 5 through 7 of LLD (2006)