

# روش های عددی در ژئومکانیک

Numerical methods in geotechnic

**Ill-posed problems and Regularization Method**

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دانشگاه صنعتی خواجه نصیرالدین طوسی

## Well-Posedness

Definition due to Hadamard, 1915: Given mapping

$$A: X \rightarrow Y$$

$$Ax = y$$

**is well-posed provided:**

- (Existence) For each  $y \in Y$ ,  $\exists x \in X$  such that  $Ax = y$ ;
- (Uniqueness)
- (Stability)  $A^{-1}$  is continuous

Equation is ill-posed if it is **not well-posed**.

**Then** an arbitrarily small perturbation of the data can cause an arbitrarily large perturbation of the solution.

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**Examples of Ill-Posed problems**

$$Ax = b \Rightarrow x = (A)^{-1} b$$

$$\begin{pmatrix} 1.00000000000001 & 3 \\ 2 & 4.00000000000001 \end{pmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 7 \\ 10 \end{bmatrix} \Rightarrow \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 1.0000 \\ 2.0000 \end{bmatrix}$$

$$\det(A) = -1$$

$$\begin{pmatrix} 1.00000000000001 & 3 \\ 2 & 6.00000000000001 \end{pmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 7 \\ 14 \end{bmatrix} \quad \det(A) = 7.5495e-15$$

$$x = \text{inv}(A) * b = 1.0e+14 * \begin{bmatrix} 7.9475 & -3.9738 \\ -2.6492 & 1.3246 \end{bmatrix} \begin{bmatrix} 7 \\ 14 \end{bmatrix} \Rightarrow \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0.6250 \\ 2.0938 \end{bmatrix}$$

$$x = \text{pinv}(A) * b = \begin{bmatrix} 0.0200 & 0.0400 \\ 0.0600 & 0.1200 \end{bmatrix} \begin{bmatrix} 7 \\ 14 \end{bmatrix} \Rightarrow \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0.7000 \\ 2.1000 \end{bmatrix}$$

$\begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = ?$  جواب دقیق؟

خطاها معمولاً در اندازه گیری وارد مساله می شوند.

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**Examples of Ill-Posed problems**

- Differentiation**

$$f_1(x) = f(x) \quad f_2(x) = f(x) + \varepsilon \sin(\omega x) \implies |f_1(x) - f_2(x)| \leq \varepsilon$$

$$\implies \text{Edge-detection is ill-posed!} \quad |f_1'(x) - f_2'(x)| \leq \varepsilon \omega$$
- Image restoration**

g:blurred image      k:response function      f:unknown object

$$g(x) = (k * f)(x)$$

k:band-limited: invisible objects exist

Existence if and only if  $\int |\hat{g}(\omega) / \hat{k}(\omega)|^2 d\omega < \infty$
- Fredholm integral equation of the first kind**

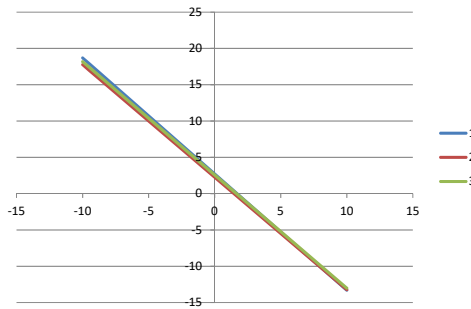
$$\int_a^b K(x, y) \varphi(x) dx = f(y) \quad , \quad c \leq y \leq d \implies \varphi(y) = ?$$

*f* is noisy data and is not the exact data

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Examples of Ill-Posed problems

$$\begin{bmatrix} 0.16 & 0.10 \\ 0.17 & 0.11 \\ 2.02 & 1.29 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0.261 \\ 0.278 \\ 3.313 \end{bmatrix} \quad \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = ?$$



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Examples of Ill-Posed problems

$$\begin{bmatrix} 0.16 & 0.10 \\ 0.17 & 0.11 \\ 2.02 & 1.29 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0.261 \\ 0.278 \\ 3.313 \end{bmatrix}$$

$$\begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = ?$$

جواب دقیق؟

خطاها معمولاً در اندازه گیری وارد مساله می شوند.

جواب دقیق مساله

$$b = \begin{bmatrix} 0.16 & 0.10 \\ 0.17 & 0.11 \\ 2.02 & 1.29 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \end{bmatrix} + \begin{bmatrix} +0.001 \\ -0.002 \\ +0.003 \end{bmatrix}$$

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**Examples of Ill-Posed problems**

روش اول: جوابها با حل دو معادله

$$\begin{pmatrix} 0.16 & 0.10 \\ 0.17 & 0.11 \end{pmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0.261 \\ 0.278 \end{bmatrix} \Rightarrow \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 1.5167 \\ 0.1833 \end{bmatrix}$$

$$2.02 * 1.5167 + 1.29(0.1833) = 3.313 + (-0.0128)$$

$$\begin{pmatrix} 0.16 & 0.10 \\ 2.02 & 1.29 \end{pmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0.261 \\ 3.313 \end{bmatrix} \Rightarrow \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 1.2250 \\ 0.6500 \end{bmatrix}$$

$$\begin{pmatrix} 0.17 & 0.11 \\ 2.02 & 1.29 \end{pmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0.278 \\ 3.313 \end{bmatrix} \Rightarrow \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 2.0034 \\ -0.5690 \end{bmatrix}$$

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**Examples of Ill-Posed problems**

روش دوم: جوابها به روش حداقل مربعات خطا

$$\min \|Ax - b\|^2$$

$$\begin{bmatrix} 0.16 & 0.10 \\ 0.17 & 0.11 \\ 2.02 & 1.29 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0.261 \\ 0.278 \\ 3.313 \end{bmatrix}$$

سه رقم اعشار

$$f(x_1, x_2) = [0.261 - (0.16x_1 + 0.1x_2)]^2 + [0.278 - (0.17x_1 + 0.11x_2)]^2 + [3.313 - (2.02x_1 + 1.29x_2)]^2$$

$$\min f \begin{cases} \frac{\partial f}{\partial x_1} = 0 \Rightarrow 8.270x_1 + 5.281x_2 = 13.563 \\ \frac{\partial f}{\partial x_2} = 0 \Rightarrow 5.281x_1 + 3.372x_2 = 8.661 \end{cases}$$

$$\Rightarrow \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 1.7077 \\ -0.1059 \end{bmatrix}$$

$$b = \begin{bmatrix} 0.16 & 0.10 \\ 0.17 & 0.11 \\ 2.02 & 1.29 \end{bmatrix} \begin{bmatrix} 1.7077 \\ -0.1059 \end{bmatrix} + \begin{bmatrix} 0.0016 \\ 0.0007 \\ -0.0002 \end{bmatrix}$$

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**Examples of Ill-Posed problems**

روش دوم: جواب‌ها به روش حداقل مربعات خطا

$$\min \|Ax - b\|^2$$

$$\begin{bmatrix} 0.16 & 0.10 \\ 0.17 & 0.11 \\ 2.02 & 1.29 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0.261 \\ 0.278 \\ 3.313 \end{bmatrix}$$

تعداد اعشار مورد استفاده در برنامه متلب

$$f(x_1, x_2) = [0.261 - (0.16x_1 + 0.1x_2)]^2 + [0.278 - (0.17x_1 + 0.11x_2)]^2 + [3.313 - (2.02x_1 + 1.29x_2)]^2$$

$$\min f \begin{cases} \frac{\partial f}{\partial x_1} = 0 \Rightarrow 8.2698x_1 + 5.281x_2 = 13.5626 \\ \frac{\partial f}{\partial x_2} = 0 \Rightarrow 5.281x_1 + 3.3724x_2 = 8.6609 \end{cases} \Rightarrow \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 1.4615 \\ 0.2796 \end{bmatrix}$$

$$b = \begin{bmatrix} 0.16 & 0.10 \\ 0.17 & 0.11 \\ 2.02 & 1.29 \end{bmatrix} \begin{bmatrix} 1.4615 \\ 0.2796 \end{bmatrix} + \begin{bmatrix} 0.0008 \\ 0.0012 \\ -0.0002 \end{bmatrix}$$

$$\text{Lscov}(A, b) \Rightarrow \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 1.4615 \\ 0.2796 \end{bmatrix}$$

دستور متلب برای حل به روش حداقل مربعات خطا

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**Examples of Ill-Posed problems**

روش سوم: معکوس ماتریس ضرایب

$$\begin{bmatrix} 0.16 & 0.10 \\ 0.17 & 0.11 \\ 2.02 & 1.29 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0.261 \\ 0.278 \\ 3.313 \end{bmatrix}$$

ماتریس مستطیلی به روش Moore-Penrose pseudoinverse شبه معکوس می‌شود

دستور متلب برای شبه معکوس  $A^+ : \text{pinv}(A)$

$$A = \begin{bmatrix} 0.16 & 0.10 \\ 0.17 & 0.11 \\ 2.02 & 1.29 \end{bmatrix} \rightarrow A^+ = \begin{bmatrix} 204.1237 & -135.1226 & -4.3015 \\ -319.5876 & 211.6601 & 7.5009 \end{bmatrix}$$

$$Ax = b \rightarrow x = A^+b \Rightarrow \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 1.4615 \\ 0.2796 \end{bmatrix}$$

توجه  $A^+A = I$  but  $AA^+ \neq I$

left inverse  $A^+A = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$   $AA^+ = \begin{bmatrix} 0.7010 & -0.4536 & 0.0619 \\ -0.4536 & 0.3118 & 0.0938 \\ 0.0619 & 0.0938 & 0.9872 \end{bmatrix}$

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**Examples of Ill-Posed problems**

روش چهارم: استفاده از ترنسپوز ماتریس  $A$

$$\begin{bmatrix} 0.16 & 0.10 \\ 0.17 & 0.11 \\ 2.02 & 1.29 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0.261 \\ 0.278 \\ 3.313 \end{bmatrix}$$

ماتریس مستطیلی به روش Moore-Penrose pseudoinverse شبیه معکوس می‌شود

$$Ax = b \rightarrow A^T Ax = A^T b \Rightarrow x = (A^T A)^{-1} A^T b$$

$$A = \begin{bmatrix} 0.16 & 0.10 \\ 0.17 & 0.11 \\ 2.02 & 1.29 \end{bmatrix} \rightarrow A^T = \begin{bmatrix} 0.16 & 0.17 & 2.02 \\ 0.10 & 0.11 & 1.29 \end{bmatrix} \Rightarrow A^T A = \begin{bmatrix} 4.1349 & 2.6405 \\ 2.6405 & 1.6862 \end{bmatrix}$$

$\det(A^T A) = 2.8130e-05$

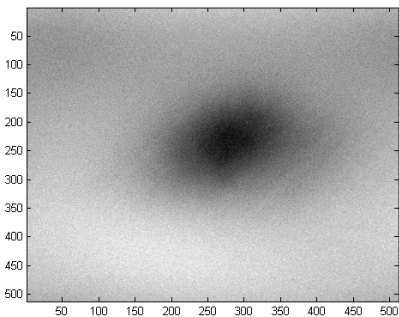
$$x = (A^T A)^{-1} A^T b = \begin{bmatrix} 59940 & -93870 \\ -93870 & 146990 \end{bmatrix} \begin{bmatrix} 6.7813 \\ 4.3304 \end{bmatrix} \Rightarrow \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 1.4615 \\ 0.2796 \end{bmatrix}$$

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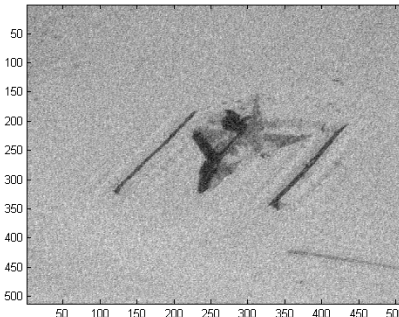
**Regularization Method**

**Primary difficulty with the discrete ill-posed problems:**

- They are essentially underdetermined due to the cluster of small singular values of  $A$ .



**Original image**



**Regularized image**

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## Regularization Method

**Most of inverse problems end in Ill-posed linear systems:**

$$KO \approx G = G_{true} + \delta, \quad G_{true} = KO_{true}$$

$\delta$  : noise

**Which is directly obtained from Fredholm integral:**

$$G(x) = \int_a^b K(x,t)O(t)dt$$

**Having Kernel function  $K(x,t)$  and  $G(x)$ ,  $O(t)$  is evaluated:**

$$O(x) = F_{\omega}^{-1} \left[ \frac{F_x [G(x)](\omega)}{F_x [K(x)](\omega)} \right] = \int_{-\infty}^{+\infty} \frac{F_x [G(x)](\omega)}{F_x [K(x)](\omega)} e^{2\pi i \omega x} d\omega$$

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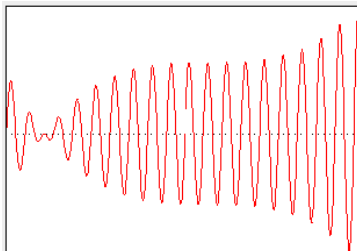
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## Regularization Method

**According to Riemann-Lebesgue lemma, Fourier transform of sinusoidal functions which their values fluctuates around zero tends to zero in infinity.**

Let  $f : \mathbb{R} \rightarrow \mathbb{C}$

$$\int_{-\infty}^{+\infty} f(x)e^{izx} dx \rightarrow 0 \text{ as } z \rightarrow \pm\infty$$



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## Regularization Method

However, obtaining  $O(x)$  would be difficult using  $G(x)$ .

**X** has a high frequency  $\rightarrow \|KO\| \ll \|O\|$

### Tikhonov Regularization Method:

$$Ax = b$$

$$\min \left\{ \|Ax - b\|_2^2 + \lambda \|Lx\|_2^2 \right\} \quad \lambda = \frac{\|Ax - b\|_2^2}{\|Lx\|_2^2}$$

$$\hat{x} = (A^T A + \lambda L^T L)^{-1} A^T b$$

$L$  Tikhonov matrix

$L = I$  In many cases, this matrix is identity matrix (standard form)

$L = 0$  reduces to the unregularized least squares solution

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## Regularization

### Morozov's discrepancy principle

Ask for the norm of the residual to be equal to the norm of the noise vector (if it is known)

Choose  $\lambda_0$  and  $0 < d < 1$

Set  $j = 0$  and find  $x_0 = (A^T A + \lambda_0 L^T L)^{-1} A^T b$  for  $\lambda_0$

While  $\|Ax - b\| > \delta$

$$j = j + 1, \lambda_j = d \lambda_{j-1}$$

$$x_j = (A^T A + \lambda_j L^T L)^{-1} A^T b$$

While not  $\|Ax - b\| = \delta$

$$\lambda_{\max} = \lambda_{j-1}, \lambda_{\min} = \lambda_j$$

$$\lambda = (\lambda_{\max} + \lambda_{\min}) / 2$$

$$x_\lambda = (A^T A + \lambda L^T L)^{-1} A^T b$$

If  $\|Ax_\lambda - b\| > \delta$  then  $\lambda_{\max} = \lambda$  else  $\lambda_{\min} = \lambda$

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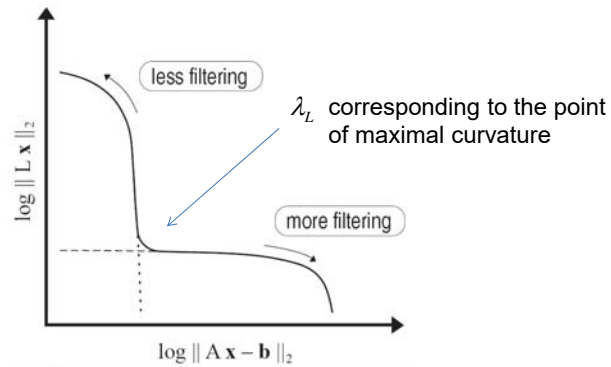


### Regularization

plot in log-log scale the curve  $(\|Lx_\lambda\|, \|Ax_\lambda - b\|)$  obtained by varying the value of  $\lambda \in [0, \infty)$

In most cases this curve is shaped as an "L"

Lawson and Hanson proposed to choose the value  $\lambda_L$  corresponding to the "corner" of the L-curve



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### Regularization Method

$$Ax = b$$

روش چهارم: روش منظم سازی تیخونف

$$\begin{bmatrix} 0.16 & 0.10 \\ 0.17 & 0.11 \\ 2.02 & 1.29 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0.261 \\ 0.278 \\ 3.313 \end{bmatrix}$$

$$\min \left\{ \|Ax - b\|_2^2 + \lambda \|Lx\|_2^2 \right\}, L = I_n$$

$$\begin{cases} 0.16x_1 + 0.1x_2 = 0.261 \\ 0.17x_1 + 0.11x_2 = 0.278 \\ 2.02x_1 + 1.29x_2 = 3.313 \end{cases}$$

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### Regularization Method

$$f(x_1, x_2) = [0.261 - (0.16x_1 + 0.1x_2)]^2 + [0.278 - (0.17x_1 + 0.11x_2)]^2 + [3.313 - (2.02x_1 + 1.29x_2)]^2 + \lambda(x_1 + x_2)^2$$

$$\min f \begin{cases} \frac{\partial f}{\partial x_1} = 0 \Rightarrow \\ \frac{\partial f}{\partial x_2} = 0 \Rightarrow \end{cases} \begin{cases} 8.2698x_1 + 5.281x_2 + 2\lambda(x_1 + x_2) = 13.5626 \\ 5.281x_1 + 3.3724x_2 + 2\lambda(x_1 + x_2) = 8.6609 \end{cases}$$

$\lambda = 0.01$

 $\begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 4.5280 \\ -4.5225 \end{bmatrix}$

$\lambda = 0.001$

 $\begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 4.4448 \\ -4.3922 \end{bmatrix}$

$\lambda = 0.0001$

 $\begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 3.8948 \\ -3.5309 \end{bmatrix}$

$\lambda = 0.00001$

 $\begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 2.9627 \\ -2.0713 \end{bmatrix}$

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### Regularization Method

$\lambda = 0.01$

$\begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 4.5280 \\ -4.5225 \end{bmatrix}$

$b = \begin{bmatrix} 0.16 & 0.10 \\ 0.17 & 0.11 \\ 2.02 & 1.29 \end{bmatrix} \begin{bmatrix} 4.5280 \\ -4.5225 \end{bmatrix} + \begin{bmatrix} 0.0112 \\ -0.0057 \\ -0.0004 \end{bmatrix}$

$\begin{matrix} \log(\|Lx\|) & 1.8563 \\ \log(\|Ax - b\|) & -4.3733 \end{matrix}$

$\lambda = 0.001$

$\begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 4.4448 \\ -4.3922 \end{bmatrix}$

$b = \begin{bmatrix} 0.16 & 0.10 \\ 0.17 & 0.11 \\ 2.02 & 1.29 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \begin{bmatrix} 0.0109 \\ -0.0055 \\ -0.0004 \end{bmatrix}$

$\begin{matrix} \log(\|Lx\|) & 1.8324 \\ \log(\|Ax - b\|) & -4.4005 \end{matrix}$

$\lambda = 0.0001$

$\begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 3.8948 \\ -3.5309 \end{bmatrix}$

$b = \begin{bmatrix} 0.16 & 0.10 \\ 0.17 & 0.11 \\ 2.02 & 1.29 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \begin{bmatrix} 0.0091 \\ -0.0043 \\ -0.0004 \end{bmatrix}$

$\begin{matrix} \log(\|Lx\|) & 1.6596 \\ \log(\|Ax - b\|) & -4.6007 \end{matrix}$

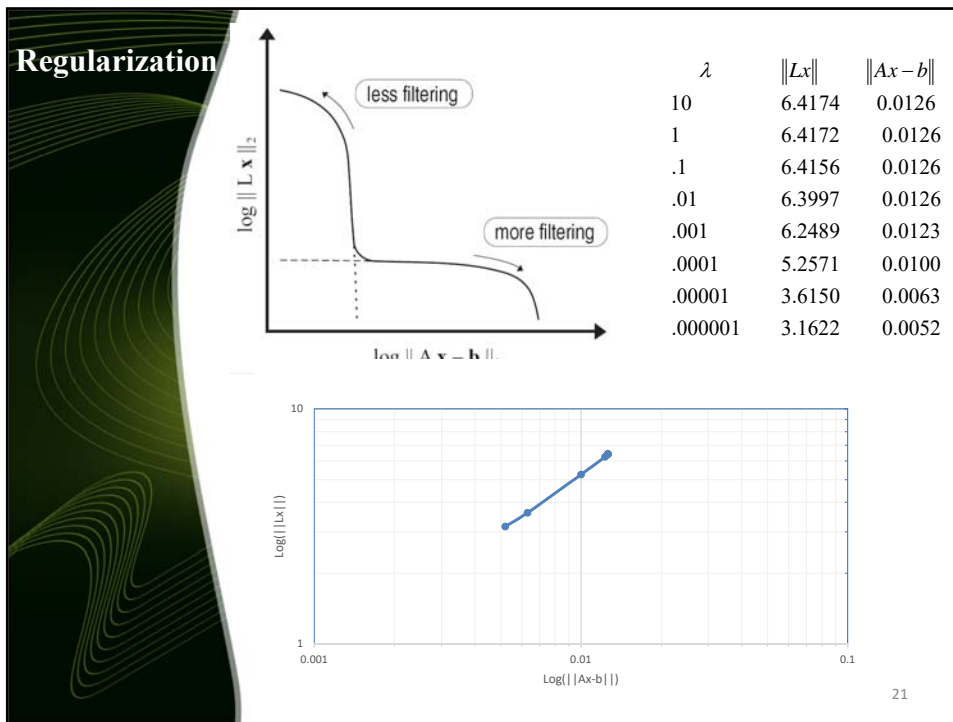
$\lambda = 0.00001$

$\begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 2.9627 \\ -2.0713 \end{bmatrix}$

$b = \begin{bmatrix} 0.16 & 0.10 \\ 0.17 & 0.11 \\ 2.02 & 1.29 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \begin{bmatrix} 0.0059 \\ -0.0022 \\ -0.0003 \end{bmatrix}$

$\begin{matrix} \log(\|Lx\|) & 1.2851 \\ \log(\|Ax - b\|) & -5.0670 \end{matrix}$

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### Examples of

- **Fredholm integral equation of the first kind**

$$\int_a^b K(x, y)\varphi(x)dx = f(y), \quad c \leq y \leq d \Rightarrow \varphi(y) = ?$$

Simpson's rule  $\sum_{j=1}^n K_{ij} \Delta x_j \varphi_j = f(x_i), \quad i = 1, \dots, n$

inverting the matrix  $A^n$  will fail  $A_{ij}^n = K_{ij} \Delta x_j$

$$\int_0^1 \frac{1}{1+100(x-y)^2} \varphi(x)dx = f(y), \quad -2 \leq y \leq 2$$

the data f is chosen such that:

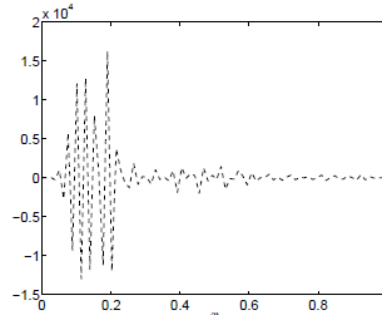
$$\varphi(x) = e^{-100(x-.25)^2} + e^{-100(x-.75)^2}$$

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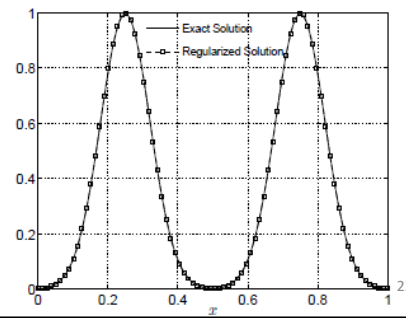
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•Fredholm integral equation of the first kind

inverting the matrix  $A^n$



exact solution and the regularized solution



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