

# Optimization

## Math Lecture 5

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# Main Goal

Find  $x$  such that  $Ax \approx b$ .

## Three possibilities

**There doesn't exist any  $x$  that satisfies**

**There exists exactly 1  $x$  that satisfies**

**There exists infinitely many  $x$  that satisfies**

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# One Solution

If there exists a solution, then we seek the  $x$  that satisfies  $Ax = b$ .

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# No Solutions

If there doesn't exist any  $x$  such that  $Ax = b$  then we seek the smallest  $x$  that minimizes

$$\|Ax - b\|_2$$

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# Infinite Solutions

If there exist infinite solutions, then we seek the smallest  $x$  that satisfies

$$Ax = b$$

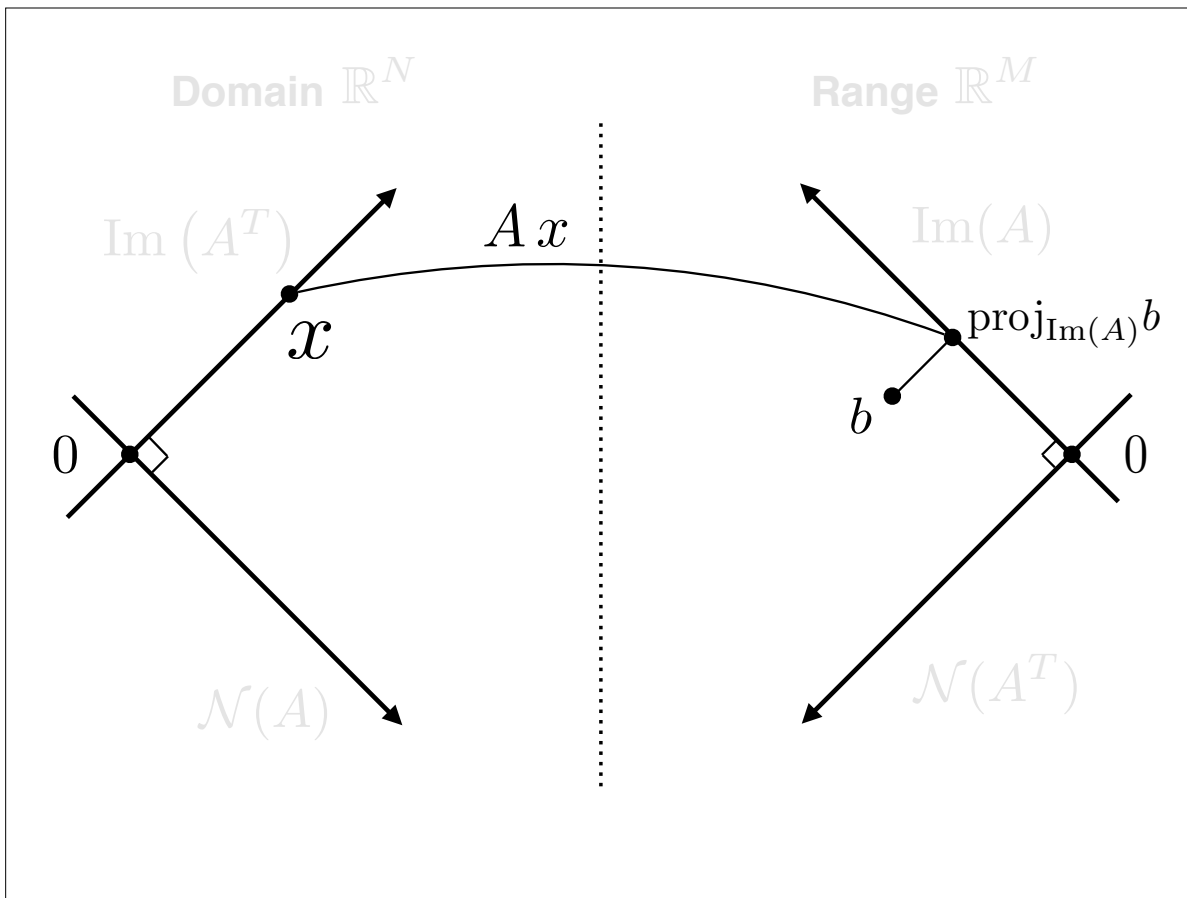
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# Pseudo-Inverse

The Pseudo-Inverse of  $A$  is the matrix  $A^\dagger$  such that  $A^\dagger b$  is the solution to the following optimization problem (for any  $b$ ).

$$\begin{aligned} & \text{minimize} && \|x\|_2 \\ & \text{subject to} && Ax = \text{proj}_{\text{Im}(A)} b \end{aligned}$$

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**Most common situation is that there are no solutions and  $A$  is tall and skinny with linearly independent columns**

$$Ax = b$$

**Left multiply by  $A^T$**

$$A^T A x = A^T b$$

**The expression above is called the “Normal Equations”**

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$$A^T A x = A^T b$$

**Since  $A$  has linearly independent columns,  $A^T A$  is invertible.**

$$x = (A^T A)^{-1} A^T b$$

**Now we see how to determine the pseudo-inverse of  $A$  for this situation**

$$A^\dagger = (A^T A)^{-1} A^T$$

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**Mathematical computer programs have the pseudo-inverse solution built in.**

**In python**

```
x = numpy.linalg.solve( A, b )
```

**In Matlab**

```
x = A \ b;
```

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# Regularization

If there are infinite solutions, we must do more to make the solution unique.

This is accomplished through Regularization

$$\text{minimize } \|Ax - b\|_2 + \gamma R(x)$$

$R$  is the regularization function.

This is also often done even if there are no solutions to make the problem better behaved.

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# Tikhonov Regularization

$$\text{minimize } \|Ax - b\|_2 + \gamma \|\Gamma x\|_2$$

$\Gamma$  is called the Tikhonov matrix

$\gamma$  is the regularization parameter.

Trades off importance of data matching and regularization

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# Tikhonov Regularization

$$\text{minimize } \|Ax - b\|_2 + \gamma \|\Gamma x\|_2$$

**Example Tikhonov matrices:**

**$D$  -  $Dx$  is a vector of all the horizontal and vertical differences. Used if  $x$  is expected to be smooth**

**$I$  - identity matrix is used when  $x$  is expected to be small**

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# Tikhonov Regularization

$$\text{minimize } \|Ax - b\|_2 + \gamma \|\Gamma x\|_2$$

**Can be combined into a minimization of a single term.**

**You will do this for homework.**

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