

# Computer Vision INM460 / IN3060

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**Mathematics Primer** 

with examples in Matlab



#### Overview of this session

- Mathematics of computer vision
  - Linear algebra
    - Vectors
    - Matrices
    - Homogenous coordinates
  - Calculus
    - Derivatives
    - Integrals

#### **Scalars**

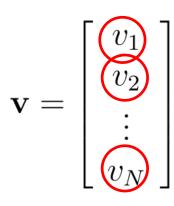
- A scalar is simply a number
- Mathematically, it is denoted with a non-bold font, for example
  - x = 6.0
  - y = 5.0
- Scalars can be added, subtracted, multiplied, and divided (except divide by 0):
  - $\cdot x + y = 11.0$
  - $\cdot x y = 1.0$
  - x \* y = 30.0
  - x / y = 1.2

 $\begin{array}{cccc}
x & = 6 \\
y & = 5 \\
x & + y \\
x & - y \\
x & * y \\
x & / y
\end{array}$ 



#### **Vectors**

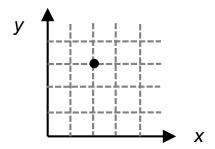
- Notation: using a bold font
- Used to represent, in *N* (typically, 2, 3, or 4) dimensions:
  - Position of a point in space (from the origin)
  - Direction
  - Colour: E.g., red, green, blue
- An N dimensional vector can be written as



components (or elements)

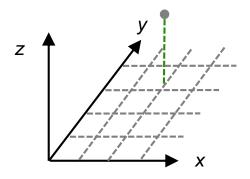
## **Vector examples**

• Represent the point (2, 3) in 2D using a vector



$$\mathbf{p}_1 = \left[ \begin{array}{c} 2\\3 \end{array} \right]$$

Represent the point (2, 3, 3) in 3D using a vector



$$\mathbf{p}_2 = \left[ \begin{array}{c} 2\\3\\3 \end{array} \right]$$

#### Column vectors and row vectors

• A *column vector* is a vector that has all components in a vertical column

$$\mathbf{v} = \left| egin{array}{c} v_1 \\ v_2 \\ dots \\ v_N \end{array} 
ight|$$

- A row vector has components in a horizontal row.
- A transpose changes column vector to a row vector, and vice-versa.

$$\mathbf{v}^T = [v_1, v_2, \cdots v_N]$$

#### Vector arithmetic

Vectors can be added and subtracted, to form a new vector

$$\mathbf{p} + \mathbf{q} = \begin{bmatrix} p_1 + q_1 \\ p_2 + q_2 \\ \vdots \\ p_N + q_N \end{bmatrix} \qquad \mathbf{p} - \mathbf{q} = \begin{bmatrix} p_1 - q_1 \\ p_2 - q_2 \\ \vdots \\ p_N - q_N \end{bmatrix}$$

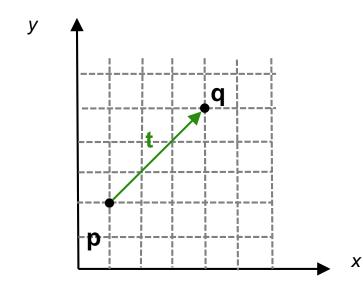
For example, if  $\mathbf{p} = [2, 2, 2]^T$  and  $\mathbf{q} = [1, 0, 1]^T$ , determine  $\mathbf{p} - \mathbf{q}$ .

$$\mathbf{p} - \mathbf{q} = \begin{bmatrix} 2 \\ 2 \\ 2 \end{bmatrix} - \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} =$$

#### Vector addition and subtraction

- Often used for translation:
  - Example: move the point  $\mathbf{p} = [1, 2]^T$  by adding a displacement  $\mathbf{t} = [3, 3]^T$  to form the point  $\mathbf{q}$ .

$$\mathbf{q} = \mathbf{p} + \mathbf{t} = \begin{bmatrix} 1 \\ 2 \end{bmatrix} + \begin{bmatrix} 3 \\ 3 \end{bmatrix} = \begin{bmatrix} 4 \\ 5 \end{bmatrix}$$

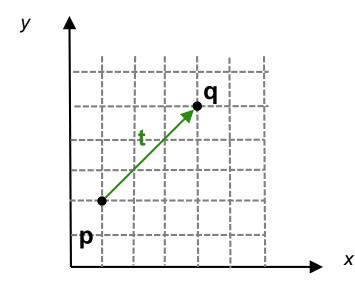




#### Vector addition and subtraction

- Also commonly used to find a vector between points:
  - Example: find the vector t between points p and q and originating from p.

$$\mathbf{t} = \mathbf{q} - \mathbf{p} = \begin{bmatrix} 4 \\ 5 \end{bmatrix} - \begin{bmatrix} 1 \\ 2 \end{bmatrix} = \begin{bmatrix} 3 \\ 3 \end{bmatrix}$$



$$p = [1, 2]'$$
  
 $q = [4, 5]'$   
 $t = q - p$ 



## Scalar multiplication

Scalar multiplication: multiplies each component of the vector by a scalar.

$$a\mathbf{v} = \mathbf{v}a = \begin{bmatrix} av_1 \\ av_2 \\ \vdots \\ av_N \end{bmatrix}$$

```
v = [1, 2, 3]';
a = 4;
a*v
```

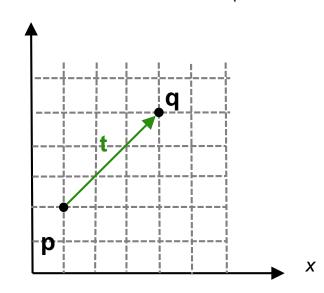


## **Vector length**

- A vector has a length (or *magnitude*) given by the expression:  $||\mathbf{v}|| = \sqrt{\sum_{i=1}^n v_i^2}$
- Note that the vector length is a scalar.
- What is the length of t?

$$\mathbf{t} = \begin{bmatrix} 3 \\ 3 \end{bmatrix}$$

$$||\mathbf{t}|| = \sqrt{3^2 + 3^2} = \sqrt{18} = 3\sqrt{2}$$



```
In Matlab (Two ways)
```

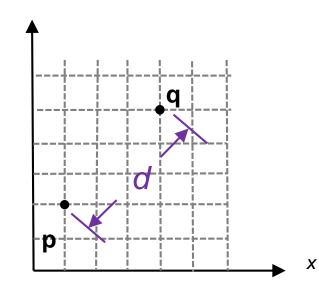
```
p = [1, 2]';
q = [4, 5]';
t = q - p
lengthT = sqrt(sum(t.^2))
lengthT = norm(t)
```



#### Distance between two points

- The distance between two points is simply the length of the vector between the two points, as described on the previous slide!
- It can be expressed as

$$d = \sqrt{\sum_{i} (p_i - q_i)^2}$$



In Matlab (Two ways)

```
p = [1, 2]';
q = [4, 5]';
d = norm(p-q)
d = pdist2(p', q')
```



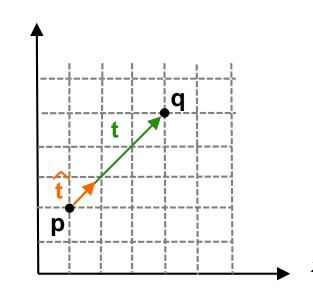
#### **Vector normalisation**

- Normalisation scales a vector so that it has unit length. That is, the length
  of the vector is one after normalisation.
- Any vector with at least one non-zero component can be normalised.

$$rac{\mathbf{v}}{||\mathbf{v}||}$$

• Determine  $\hat{\mathbf{t}}$ , by normalising  $\mathbf{t}$ 

$$\hat{\mathbf{t}} = \frac{\mathbf{t}}{||\mathbf{t}||} = \frac{1}{3\sqrt{2}} \begin{bmatrix} 3\\3 \end{bmatrix} = \begin{bmatrix} \frac{1}{\sqrt{2}}\\ \frac{1}{\sqrt{2}} \end{bmatrix}$$



```
p = [1, 2]';
q = [4, 5]';
t = q - p;
that = t / norm(t)
```

#### **Element-wise product**

- The element-wise product of two vectors is the product of the matching elements of each vector. The result is a vector.
- In this module, we will use Matlab syntax of .\* to represent the element-wise product.
  - Example: What is the element-wise product of vectors  $\mathbf{p} = [1, 2, 3]^T$  and  $\mathbf{q} = [1, 2, 3]^T$  $[4, 5, 6]^{\mathsf{T}}$ ?

$$\begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} \cdot * \begin{bmatrix} 4 \\ 5 \\ 6 \end{bmatrix} = \begin{bmatrix} 4 \\ 10 \\ 18 \end{bmatrix}$$

## **Dot product**

- Dot product
  - Definition

$$\mathbf{p} \cdot \mathbf{q} = \sum_{i=1}^{N} p_i q_i$$

- Multiplies the ith component of each vector together, then takes the sum.
- Note that the dot product is a scalar.
- Example: What is the dot product of vectors  $\mathbf{p} = [3, 2, 1]^T$  and  $\mathbf{q} = [1, 0, -1]^T$ ?

Answer: 
$$(3)(1) + (2)(0) + (1)(-1) = 2$$

$$p = [3, 2, 1]';$$
  
 $q = [1, 0, -1]';$   
 $dot(p, q)$ 

#### Homogenous coordinates

- In computer vision (and computer graphics), often homogeneous coordinates are used to represent points. This is useful when dealing with transformations (like rotation or translation).
- Homogeneous coordinates add another dimension (or element) to the vector representing a point. For example, a 2D point [x, y]<sup>T</sup> is represented in homogeneous coordinates as

$$\begin{bmatrix} x \\ y \end{bmatrix} \qquad \qquad \qquad \qquad \begin{bmatrix} x \\ y \\ w \end{bmatrix}$$

- Two sets of homogenous coordinates represent the same point if one is a multiple of another. For example,  $[1, 2, 1]^T = [2, 4, 2]^T$
- Dividing by w (when non-zero) puts the point in Cartesian coordinates.
   [x, y, w]T = [x/w, y/w, 1]T



#### **Matrices**

- Of fundamental importance in computer vision
- A matrix F is a rectangular array of numbers (elements) that has N rows and M columns (matrix with size N x M). For example, here is a 3 x 4 matrix:

$$\mathbf{F} = \begin{bmatrix} F_{11} & F_{12} & F_{13} & F_{14} \\ F_{21} & F_{22} & F_{23} & F_{24} \\ F_{31} & F_{32} & F_{33} & F_{34} \end{bmatrix}$$
 index element

- Matrices for which N = M are called square.
- Matrices are typically used to represent transformations (e.g., between coordinate systems, rotation, scale, translation, etc.)
- In computer vision, matrices are often size 3 x 3 or 3 x 4.

#### **Matrices**

A vector is simply a matrix with N or M equal to 1. For example, this is column vector is a 3 x 1 matrix:

$$\mathbf{p} = \left[ \begin{array}{c} p_x \\ p_y \\ p_z \end{array} \right]$$

Two matrices are equal if and only if all elements are equal. Note this requires the matrices have the same size. For example,

$$\mathbf{A} = \begin{bmatrix} 1 & 3 \\ 3 & 8 \end{bmatrix}$$

$$\mathbf{A} = \begin{bmatrix} 1 & 3 \\ 3 & 8 \end{bmatrix} \qquad \qquad \mathbf{B} = \begin{bmatrix} 8 & 3 \\ 3 & 8 \end{bmatrix}$$

$$\mathbf{A} 
eq \mathbf{B}$$

## **Matrix diagonal**

- The matrix diagonal is the set of matrix elements along the diagonal of the (typically square) matrix.
- A diagonal matrix has
  - non-zero elements along the diagonal
  - zero off diagonal
- An example of a diagonal matrix:

$$\mathbf{A} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 2 & 0 & 0 \\ 0 & 0 & 3 & 0 \\ 0 & 0 & 0 & 4 \end{bmatrix}$$

In Matlab (Two ways)

```
A = diag([1, 2, 3, 4])

A = [1, 0, 0, 0; 0 2 0 0; 0 0 3 0; 0 0 0 4]
```

## **Matrix transpose**

• Transpose: switches rows with columns  $F_{ij}^T = F_{ji}$ 

$$\mathbf{F} = \begin{bmatrix} F_{11} & F_{12} & F_{13} & F_{14} \\ F_{21} & F_{22} & F_{23} & F_{24} \\ F_{31} & F_{32} & F_{33} & F_{34} \end{bmatrix}$$

$$\mathbf{F}^T = \left[ \begin{array}{cccc} F_{11} & F_{21} & F_{31} \\ F_{12} & F_{22} & F_{32} \\ F_{13} & F_{23} & F_{33} \\ F_{14} & F_{24} & F_{34} \end{array} \right]$$

Addition, subtraction: performed element-wise. Requires matrices to have the same size.

$$\mathbf{F} + \mathbf{G} = \begin{bmatrix} F_{11} + G_{11} & F_{12} + G_{12} & \cdots & F_{1M} + G_{1M} \\ F_{21} + G_{21} & F_{22} + G_{22} & \cdots & F_{2M} + G_{2M} \\ \vdots & \vdots & \ddots & \vdots \\ F_{N1} + G_{N1} & F_{N2} + G_{N2} & \cdots & F_{NM} + G_{NM} \end{bmatrix}$$

In Matlab

F = [1, 2; 3, 4];
F'
G = [1, 1; 1, 1];
F + G



## Scalar / matrix multiplication

Scalar / matrix multiplication multiplies each element with scalar

$$a\mathbf{F} = \mathbf{F}a = \begin{bmatrix} aF_{11} & aF_{12} & \cdots & aF_{1M} \\ aF_{21} & aF_{22} & \cdots & aF_{2M} \\ \vdots & \vdots & \ddots & \vdots \\ aF_{N1} & aF_{N2} & \cdots & aF_{NM} \end{bmatrix}$$

```
F = [1, 2; 3, 4]

a = 2;
```

## **Element-wise product**

• There is also an element-wise product for matrices, which simply multiplies the matching elements of each matrix.

$$\mathbf{A} = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \quad \mathbf{B} = \begin{bmatrix} 5 & 6 \\ 7 & 8 \end{bmatrix} ?$$

$$\mathbf{A}. * \mathbf{B} = \begin{bmatrix} 5 & 12 \\ 21 & 32 \end{bmatrix}$$

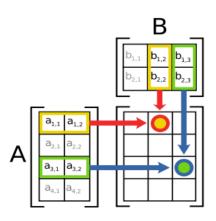
$$A = [1, 2; 3, 4];$$
  
 $q = [5, 6; 7, 8]';$   
 $A .* B$ 



## **Matrix multiplication**

- Two matrices can be multiplied together **only** when the number of columns in the matrix on the left equals the number of rows in the matrix on the right.
- For example, suppose
  - matrix A is N x M
  - matrix B is L x R
  - $\Rightarrow$  The matrix product **AB** is only defined in M = L. The resulting matrix will be of size N x R.
- It is performed by multiplying each row of A by each column of B

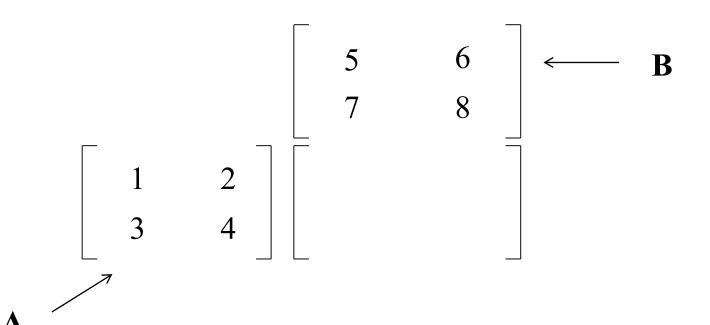
$$[\mathbf{AB}]_{ij} = \sum_{k=1}^{M} A_{ik} B_{kj}$$





Determine AB

$$\mathbf{A} = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \qquad \mathbf{B} = \begin{bmatrix} 5 & 6 \\ 7 & 8 \end{bmatrix}$$





$$\mathbf{A} = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \qquad \mathbf{B} = \begin{bmatrix} 5 & 6 \\ 7 & 8 \end{bmatrix}$$

$$\begin{bmatrix}
5 & 6 \\
7 & 8
\end{bmatrix}$$

$$\begin{bmatrix}
1 & 2 \\
3 & 4
\end{bmatrix}$$

$$\begin{bmatrix}
(1)(5)+(2)(7) \\
\hline
\end{array}$$



$$\mathbf{A} = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \quad \mathbf{B} = \begin{bmatrix} 5 & 6 \\ 7 & 8 \end{bmatrix}$$

$$\mathbf{B} = \begin{bmatrix} 5 & 6 \\ 7 & 8 \end{bmatrix}$$

$$\begin{bmatrix} 5 & 6 \\ 7 & 8 \end{bmatrix}$$

$$\begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}$$

$$\begin{bmatrix} (1)(5)+(2)(7) & (1)(6)+(2)(8) \\ \end{bmatrix}$$



$$\mathbf{A} = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \quad \mathbf{B} = \begin{bmatrix} 5 & 6 \\ 7 & 8 \end{bmatrix}$$



$$\mathbf{A} = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \qquad \mathbf{B} = \begin{bmatrix} 5 & 6 \\ 7 & 8 \end{bmatrix}$$



Multiplication example

$$\mathbf{A} = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \qquad \mathbf{B} = \begin{bmatrix} 5 & 6 \\ 7 & 8 \end{bmatrix}$$

$$\begin{bmatrix} 5 & 6 \\ 7 & 8 \end{bmatrix}$$

$$\begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \begin{bmatrix} (1)(5)+(2)(7) & (1)(6)+(2)(8) \\ (3)(5)+(4)(7) & (3)(6)+(4)(8) \end{bmatrix} = \begin{bmatrix} 19 & 22 \\ 43 & 50 \end{bmatrix}$$

$$A = [1, 2; 3, 4]$$
  
 $B = [5, 6; 7, 8];$   
 $A*B$ 



## (Double) Exercise

$$\mathbf{A} = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}$$

• Let 
$$\mathbf{A} = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}$$
 and  $\mathbf{B} = \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix}$ 

Left side of room: Determine AB

Right side of room: Determine **BA** 

$$A = [1, 2; 3, 4]$$
  
 $B = [0, 1; 1, 1];$   
 $A*B$ 

$$A = [1, 2; 3, 4]$$
  
 $B = [0, 1; 1, 1];$   
 $B*A$ 

#### Matrix multiplication not commutative

• One point to be aware of is that matrix multiplication generally does not commute. That is,

$$AB \neq BA$$

For example,

$$\begin{bmatrix} a & b \\ c & d \end{bmatrix} \begin{bmatrix} e & f \\ g & h \end{bmatrix} = \begin{bmatrix} ae + bg & \dots \\ \dots & \dots \end{bmatrix} \quad \begin{bmatrix} e & f \\ g & h \end{bmatrix} \begin{bmatrix} a & b \\ c & d \end{bmatrix} = \begin{bmatrix} ea + fc & \dots \\ \dots & \dots \end{bmatrix}$$



#### Pre- and post-multiplication

- There are terms for describing the order of matrix multiplication
  - When a matrix appears on the left, it is pre-multiplied.
  - When a matrix appears on the right, it is post-multiplied.
- For example, in the matrix product AB,
  - A is pre-multiplied to B
  - B is post-multiplied to A

## **Identity matrix**

- In scalar multiplication, any number multiplied with 1 results in the original number.
- Similarly, there is a special square matrix, called the *identity matrix*, which when multiplied by another matrix **A**, results in **A**.
- The identity matrix has the form

For a square matrix A, AI = IA = A

$$\mathbf{I} = \begin{bmatrix} 1 & 0 & \cdots & 0 \\ 0 & 1 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & 1 \end{bmatrix}$$

In Matlab  $A = [1, 2; 3, 4] \\
I = eye(2) \\
A*I \\
T*A$ 

#### **Inverse matrix**

- Also in scalar multiplication, multiplication (1/a)\*a = 1; that is, multiplication of a number by its inverse gives identity. Similarly, multiplication a matrix by its inverse gives the identity matrix.
- The inverse of a matrix A is denoted A<sup>-1</sup>. So AA<sup>-1</sup>=I, and A<sup>-1</sup>A=I
- If the inverse exists, the matrix is said to be *invertible*. Note: not all matrices are invertible!
- The inverse matrix can be computed for square matrices (number of rows equals number of columns) only.

$$\mathbf{A} = \begin{bmatrix} 9 & -2 \\ -3 & -4 \end{bmatrix} \quad \mathbf{A}^{-1} = \begin{bmatrix} \frac{2}{21} & \frac{-1}{21} \\ \frac{-1}{14} & \frac{-3}{14} \end{bmatrix} \quad \mathbf{A}\mathbf{A}^{-1} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

$$A = [9, -2; -3, -4];$$
  
 $inv(A)$   
 $A*inv(A)$ 

## System of linear equations

- In a practical context, matrices are used all the time to represent a <u>system of linear equations</u>. In computer vision, this comes up frequently, for example, in
  - Transformations (e.g., warping, rotation, projection)
  - Estimation (least squares, optical flow)
  - (more)
- Example:

$$2x + 3y = 6$$
$$4x + 9y = 15$$

Can be written as

$$\left[\begin{array}{cc} 2 & 3 \\ 4 & 9 \end{array}\right] \left[\begin{array}{c} x \\ y \end{array}\right] = \left[\begin{array}{c} 6 \\ 15 \end{array}\right]$$

## Solving a system of linear equations

This has the form of A x = b, where A is a 2x2 matrix, and x and b are 2x1 matrices.

$$\begin{bmatrix} 2 & 3 \\ 4 & 9 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} 6 \\ 15 \end{bmatrix}$$

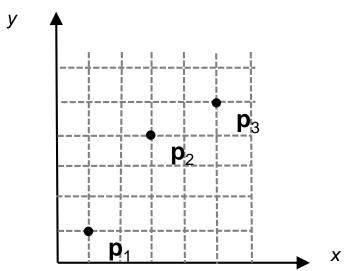
$$\uparrow \qquad \uparrow \qquad \uparrow \qquad b$$

• Using the inverse of **A**, there is a simple solution, namely  $\mathbf{x} = \mathbf{A}^{-1} \mathbf{b}$ 

```
In Matlab
A = [2, 3; 4, 9];
b = [6, 15]';
x = inv(A)*b
```

# Overdetermined system

- In the previous example, there were 2 equations and 2 unknowns (x and y).
   When the number of equations (N) is equal to the number of unknowns (M), we say the system is determined.
- If N < M, the system is <u>undetermined</u>; there is no unique solution.
- If N > M, the system is <u>overdetermined</u>; and we normally will look for the best fitting solution (for example, minimising the error in the least squares sense).
- Example: Find the best fitting line through the points  $\mathbf{p}_1 = [1, 1]^T$ ,  $\mathbf{p}_2 = [3, 4]^T$  and  $\mathbf{p}_3 = [5, 5]^T$ .



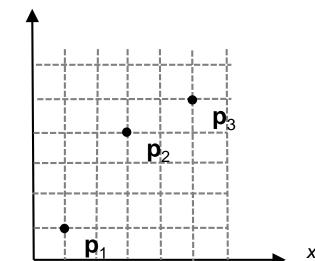


### Overdetermined system

Equation of line: y = mx + b, which is the same as xm + b = y. In our problem, we have two unknowns (m, b) and three equations (as we have three points).

$$1m + b = 1$$
$$3m + b = 4$$
$$5m + b = 5$$

$$\begin{bmatrix} 1 & 1 \\ 3 & 1 \\ 5 & 1 \end{bmatrix} \begin{bmatrix} m \\ b \end{bmatrix} = \begin{bmatrix} 1 \\ 4 \\ 5 \end{bmatrix}$$



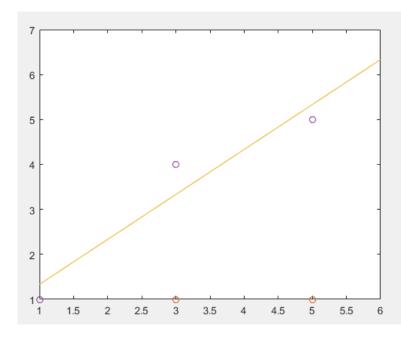


## Overdetermined system

- Note: we cannot invert **A** directly, since it is not square (i.e., it does not have an equal number of rows and columns).
- However, we can use the pseudo-inverse, (A<sup>T</sup>A)<sup>-1</sup>A<sup>T</sup>
- That is,  $\mathbf{c} = (\mathbf{A}^T \mathbf{A})^{-1} \mathbf{A}^T \mathbf{d}$ . In Matlab the pinv function implements this. This provides a least squares solution.

```
A = [1, 1; 3, 1; 5, 1];
d = [1; 4; 5];
c = pinv(A)*d

% visualise the result
m = c(1);
b = c(2);
x = [1:6];
y = m*x+b;
plot(x,y);
hold on;
scatter(A(:, 1), d);
```





### **Transformations**

- Transformations are frequently used in computer vision to transform points, that is, taking points and translating, rotating, scaling, shearing, and/or projecting them.
- Transformations are often represented using matrices, and to transform a point, it is simply a matter of matrix and vector multiplication.
- This takes the form p' = T p, where T is the transformation matrix, p is the original point, and p' is the transformed result.

# **Example (rigid body transformation)**

- One type of transformation is a rigid body transformation. This applies rotation, and translation.
- Since neither rotation or translation cause distortions to the shape, the object moves rigidly to its new location.

$$\begin{bmatrix} x' \\ y' \\ w' \end{bmatrix} = \begin{bmatrix} \cos \theta & -\sin \theta & t_x \\ \sin \theta & \cos \theta & t_y \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$$

Performing the matrix multiplication, you can see

$$x' = x \cos \theta - y \sin \theta + t_x$$
  

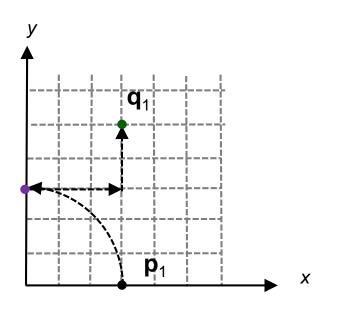
$$y' = x \sin \theta + y \cos \theta + t_y$$
  

$$w' = 1$$



# **Example (rigid body transformation)**

- Let's use this transformation to rotate by 90 degrees and translate by 3 in x and 2 in y. If we apply this transformation to a point  $\mathbf{p}_1 = [3, 0]^T$ , we would expect it to move to  $\mathbf{q}_1 = [3, 5]^T$ .
- Note: Matlab trigonometric functions like sin and cos expect angles in radians.
- Applying a transformation to all the points making up a shape will transform the shape.



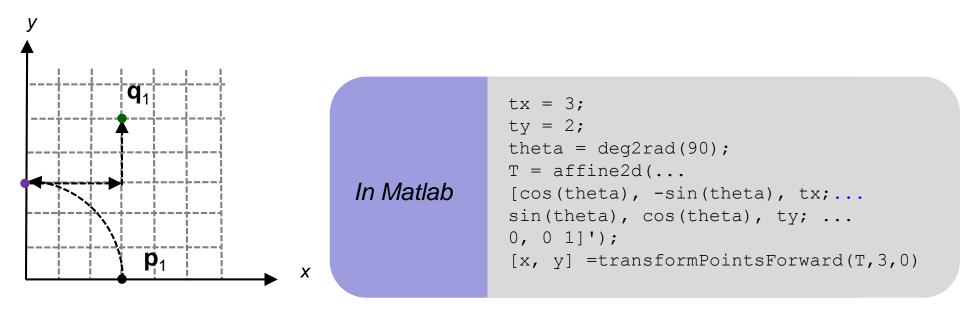
In Matlab

```
p1 = [3, 0, 1]';
tx = 3;
ty = 2;
theta = deg2rad(90);
T = [cos(theta), -sin(theta), tx;...
sin(theta), cos(theta), ty; 0, 0 1];
q1 = T*p1
```



# Using affine2d

- Matlab has an affine2d class useful for many transformations, including the rigid body transformation.
- You can create an affine2d object that includes a transformation using the affine2d function. Note this function expects the *transpose* of the matrix described earlier.
- Then you can transform points using the transformPointsForward function.





### Watch out!

- When multiplying matrices A and B together, multiplication is only possible if the number columns of A equals the number of rows of B.
  - Example: Multiplying a 3 x 3 matrix with a 3x 1 matrix (vector)

```
A = [1, 2, 3; 4, 5, 6; 7, 8, 9];
B = [3, 3, 3]';
A*B
ans =
18
45
72
```

Example: Multiplying a 3 x 3 matrix with a 1 x 3 matrix (vector)

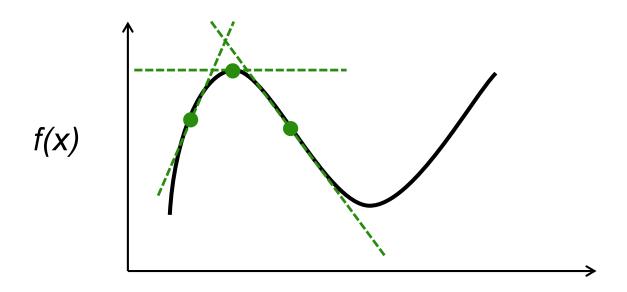


```
A = [1, 2, 3; 4, 5, 6; 7, 8, 9];
B = [3, 3, 3];
A*B
Error using *
Inner matrix dimensions must agree.
```



### **Derivatives**

- A derivative measures how much a function changes as its input changes.
- For a function f(x), the derivative (with respect to x) measures how much f changes when x changes.
- The (first) derivative provides the slope of a curve at a given point.



### **Notation for derivatives**

- There are several ways one might denote a derivative.
- A standard notation is

$$\frac{d}{dx}f(x)$$

this denotes the change in f(x) as x changes.

This is equivalent to

$$\frac{df(x)}{dx}$$

here the f(x) simply appears in the numerator.

• For notational convenience, sometimes the x in f(x) is dropped. Since the derivative is with respect to x, it is inferred that f is a function of x.

$$\frac{df}{dx}$$

• Finally, you may see a shorthand notation for a derivative,  $f_x$ 

### **Derivatives**

- The process of finding a derivative is called *differentiation*
- There are some basic rules:
  - The derivative of a constant (denoted with *c* below) is 0. This is intuitive because a constant value does not change.

$$\frac{d}{dx}c = 0$$

The power rule provides derivatives for integer powers of x

$$\frac{d}{dx}x^n = nx^{n-1}$$

What is the derivative of f(x) = 5?

$$\frac{d}{dx}5 = 0$$

What is the derivative of  $f(x) = x^2$ ?

$$\frac{d}{dx}x^2 = 2x$$

### **Derivatives**

 Constant multiple rule: If f is a differentiable function of x, and c is a constant, then

$$\frac{d}{dx}(cf) = c\frac{df}{dx}$$

Sum rule: The derivative of a sum is the sum of derivatives

$$\frac{d}{dx}(f_1 + f_2) = \frac{df_1}{dx} + \frac{df_2}{dx}$$

What is the derivative of  $f(x) = 5x^2 + 5x + 5$ ?

$$\frac{d}{dx}(5x^2 + 5x + 5) = \frac{5x^2}{dx} + \frac{5x}{dx} + \frac{d}{dx}5$$

$$= \frac{d}{dx}(5x^2) + \frac{d}{dx}(5x) + \frac{d}{dx}5$$

$$= 5\frac{d}{dx}(x^2) + 5\frac{d}{dx}(x) + \frac{d}{dx}5$$

$$= 5(2x) + 5(1) + 0$$

$$= 10x + 5$$



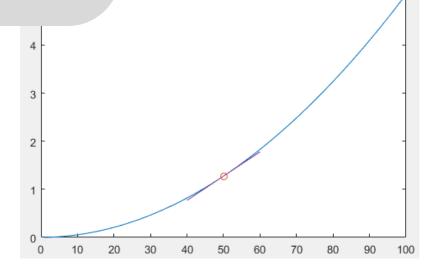
#### In Matlab

In Matlab

```
f = 5*x.^2+5*x+5;
plot(x,f);
a = 50; % A point to analyse
hold on;
scatter(x(a), f(a));

% Plot tangent at a
f_x = 10*x+5;
m = f_x(a);
xx = [x(a)-10:x(a)+10];
y = f(x(a))+m*(xx-x(a));
plot(xx, y);
```

x = [1:100];



### Second derivative

 You can differentiate a function multiple times. The second derivative is denoted as:

$$\frac{d^2}{dx^2}f$$
 ,  $\frac{df^2}{dx^2}$  , or  $f_{xx}$ 

This is simply achieved by taking the derivative of the derivative.

What is the second derivative of  $f(x) = 5x^2 + 5x + 5$ ?

$$\frac{d^{2}}{dx^{2}}(5x^{2} + 5x + 5) = \frac{d}{dx}\left(\frac{d}{dx}(5x^{2} + 5x + 5)\right)$$

$$= \frac{d}{dx}(10x + 5)$$

$$= 10$$



#### Multi-variate case

- The previous examples of derivatives were single variate, since f(x) describes f
  as function of a single variable x.
- In computer vision (and many other fields), often data has multiple dimensions.
- A good example is an image, which describes brightness (or colour) as function of two variables, x and y.

f(x, y)



f(300, 125) = 150

### **Partial derivative**

• Derivatives can be taken in the multi-variate case as well. The *partial* derivative is typically notated with a curly symbol; or the shorthand notation.

$$\frac{\partial}{\partial x}f$$
,  $\frac{\partial f}{\partial x}$  , or  $f_x$ 

 When taking a partial derivative, the derivative is taken with respect to a specific variable; the others are held constant.

What is the partial derivatives 
$$\frac{\partial f}{\partial x}$$
,  $\frac{\partial f}{\partial y}$  of  $f(x, y) = x^2 + xy + y^2$ ?

$$\frac{\partial f}{\partial x} = \frac{\partial}{\partial x}x^2 + \frac{\partial}{\partial x}xy + \frac{\partial}{\partial x}y^2$$

$$= 2x + (1)y + (0)$$

$$= 2x + y$$

$$\frac{\partial f}{\partial y} = \frac{\partial}{\partial y}x^2 + \frac{\partial}{\partial y}xy + \frac{\partial}{\partial y}y^2$$

= (0) + x(1) + 2y

= x + 2y



### **Gradient**

 The gradient of a multi-variate function f is a vector, that has a derivative with respect to each of its arguments. Consider a 2D function f(x, y). The gradient is

$$\nabla f = \left[ \frac{\partial f}{\partial x}, \frac{\partial f}{\partial y} \right]^T$$

- The gradient generalises the concept of a derivative to several dimensions. It
  is a vector that points in the direction of maximal increase of f.
- Example: Imagine you're on a walk, you're at a point (x, y), and f(x, y)
  measures height of the terrain. The gradient will point in the direction of
  greatest change in height at (x, y)



http://cycloclimbing.com/tour2007/

## Laplacian

The Laplacian is the dot product of the gradient with itself. It is denoted as

$$\Delta f = \nabla^2 f = \nabla f \cdot \nabla f$$

In 2D, the Laplacian of a function f(x, y) is

$$\Delta f = \left[ \frac{\partial f}{\partial x}, \frac{\partial f}{\partial y} \right]^T \cdot \left[ \frac{\partial f}{\partial x}, \frac{\partial f}{\partial y} \right]^T$$
$$= \frac{\partial^2 f}{\partial x^2} + \frac{\partial^2 f}{\partial y^2}$$



# Mixed partial derivative

You can take a mixed derivative, that is a derivative using multiple variables.
 For example, 22 f

$$\frac{\partial^2 f}{\partial x \partial y}$$
 , or  $f_{xy}$ 

denotes the derivative of f with respect to both x and y. The order in which you take the derivatives does not matter (e.g, x first then, y; or y first, then x).

What is the mixed partial derivative  $f_{xy}$  of  $f(x, y) = x^2 + xy + y^2$ ?

$$\frac{\partial^2 f}{\partial x \partial y}(x^2 + xy + y^2) = \frac{\partial}{\partial x} \left( \frac{\partial}{\partial y} (x^2 + xy + y^2) \right)$$
$$= \frac{\partial}{\partial x} (x^2 + xy + y^2)$$
$$= \frac{\partial}{\partial x} (x^2 + xy + y^2)$$
$$= 1$$

# **Taylor series expansion**

- A function f(x) can be approximated at a point a using a Taylor series expansion.
- This expresses the function f(x) at a using derivatives.

$$f(x) = f(a) + f_x(a)(x - a) + \frac{f_{xx}(a)}{2}(x - a)^2 + \dots$$

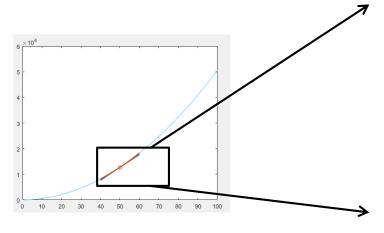
• The ... represents higher order terms, which for simplicity are often ignored as we're interested in an approximation.

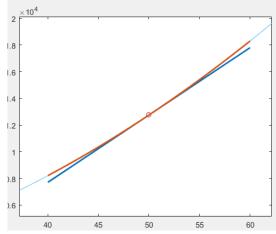


# **Taylor series expansion**

In Matlab

```
x = [1:100];
f = 5*x.^2+5*x+5;
f x = 10 * x + 5;
f xx = 10*ones(size(x));
plot(x, f); hold on;
a = 50; % A point to analyse
scatter(x(a), f(a));
x = [x(a)-10:x(a)+10];
% First order approximation at a
f first = f(a) + f x(a)*(x-a);
h = plot(x, f first);
set(h, 'LineWidth', 2);
% Second order approximation at a
f second = f(a) + f x(a)*(x-a)+0.5*f xx(a)*(x-a).^2;
h = plot(x, f second);
set(h, 'LineWidth', 2);
```





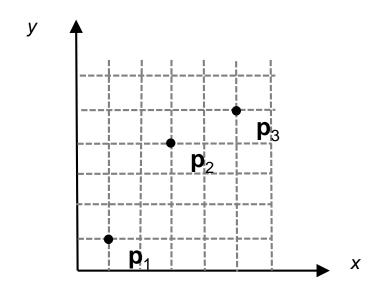


# **Optimisation**

- We've already seen an example of optimisation using least squares to solve an overdetermined system of linear equations. This trick is incredibly useful.
- Matlab has an Optimisation Toolbox that can be used to solve harder optimisation problems. For example, suppose you wanted to solve a least squares problem, but it was constrained: solutions only within a specific range are valid. For this, you can use lsqlin, which is part of Matlab.
- Or, perhaps you have a non-linear optimisation problem, where the variables you're trying to find are combined in non-linear ways. For this, you can use the <a href="lsqnonlin">lsqnonlin</a> function in Matlab.
- These functions try to find a solution that minimises an error (that you can define).

# Nonlinear optimisation example

- Find the equation of a curve  $f(x) = a + bx + abx^2$  that best passes through the points  $\mathbf{p}_1 = [1, 1]^T$ ,  $\mathbf{p}_2 = [3, 4]^T$  and  $\mathbf{p}_3 = [5, 5]^T$ .
- Here, we can't use a linear technique, because the variables we'd like to determine, a and b, are combined in a non-linear way on the last term.





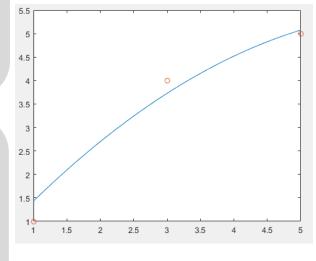
# Nonlinear optimisation example

```
In Matlab
```

```
function error = CurveError(t, p)
a = t(1);
b = t(2);
L = size(p, 1);
for i = 1:L
    x = p(i, 1);
    y = p(i, 2);
    error(i) = a + b*x +a*b*x^2 - y;
end
```

#### In Matlab

```
p = [1, 1; 3, 4; 5, 5];
t0 = [0, 0];
topt = lsqnonlin(@(t)CurveError(t, p), t0);
a = topt(1);
b = topt(2);
x = [1:.1:5];
f = a+b*x+a*b*x.^2;
plot(x, f);
hold on;
scatter(p(:, 1), p(:, 2));
```



Finds the t that minimises the CurveError function, starting with an initial guess t0 = [a, b]

# Integration

Integration is the opposite of differentiation. For example, we know

$$f(x) = x^2$$

$$\frac{df}{dx} = 2x$$

• That is, the derivative of  $x^2$  is 2x. Going the other way, the integral of 2x is

$$\int (2x)dx = x^2 + C$$

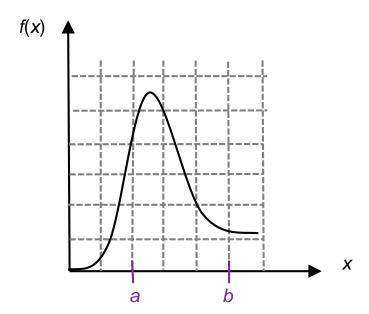
• This gives us  $x^2$  again. A constant C has been added, since any function of the form  $x^2 + C$  has a derivative 2x, since the derivative of a constant is 0.



# Indefinite vs definite integral

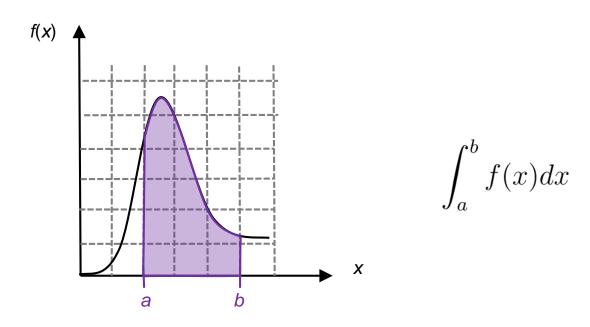
- An equation of the form  $\int f(x)dx$  is an *indefinite* integral.
- An equation of the form  $\int_a^b f(x) dx$  is a *definite* integral. The difference

is that in the case of a definite integral, we are only interested in evaluating the integral in a defined region of  $x \in [a, b]$ .



### Area under the curve

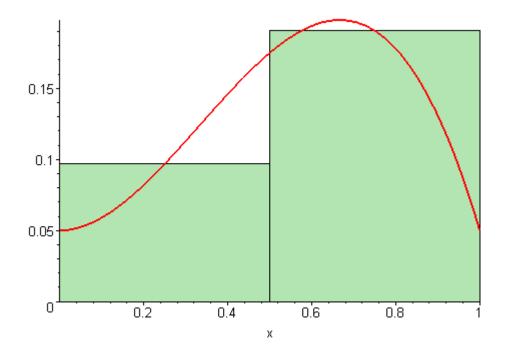
• For a signal f(x), an integral measures the area under the curve. A definite integral allows the area to be limited to a specific region.





## The integral as a sum

- Often on a computer, the integral is approximated using a sum. One can sample the signal and use rectangles to approximate the area.
- The integral is approximated by summing the areas of the rectangles.
- As the size of the rectangles gets smaller, we get a better approximation.
- ⇒ When you see an integral symbol, it may be helpful to think of it as a sum



$$\int_{a}^{b} f(x)dx \approx \sum_{x=a}^{b} f(x)\delta x$$

 $\delta x$  is the rectangle width



## The integral as a sum

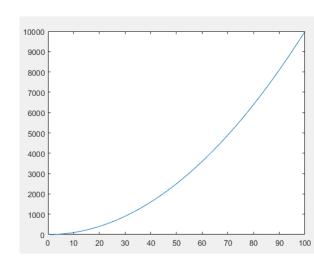
• Example: determine

$$\int_{20}^{80} x^2 dx$$

In Matlab

```
x = [1:100];
f = x.^2;
plot(x, f);

% Between 20 and 80, sample f(x) every dx
% units and multiply by box width (dx)
a = 20;
b = 80;
dx = 5;
intfx = 0;
for xx = a:dx:b-dx
    intfx = intfx + f(xx) * dx;
end
```



- When
  - $\circ$  dx = 5, intfx = 153250
  - $\circ$  dx = 1, intfx = 165010
  - Actual area: 168000



# **Double integrals**

 Integrals can be done over multiple variables. This is quite common in computer vision, as images are defined a two dimensional domain (x and y).
 An indefinite integral over a 2D domain may look like:

$$\iint f(x,y)dxdy$$

• Sometimes definite integrals are defined using a symbol like *R* to mean a particular region (of an image, for example):

$$\iint_{B} f(x,y) dx dy$$

