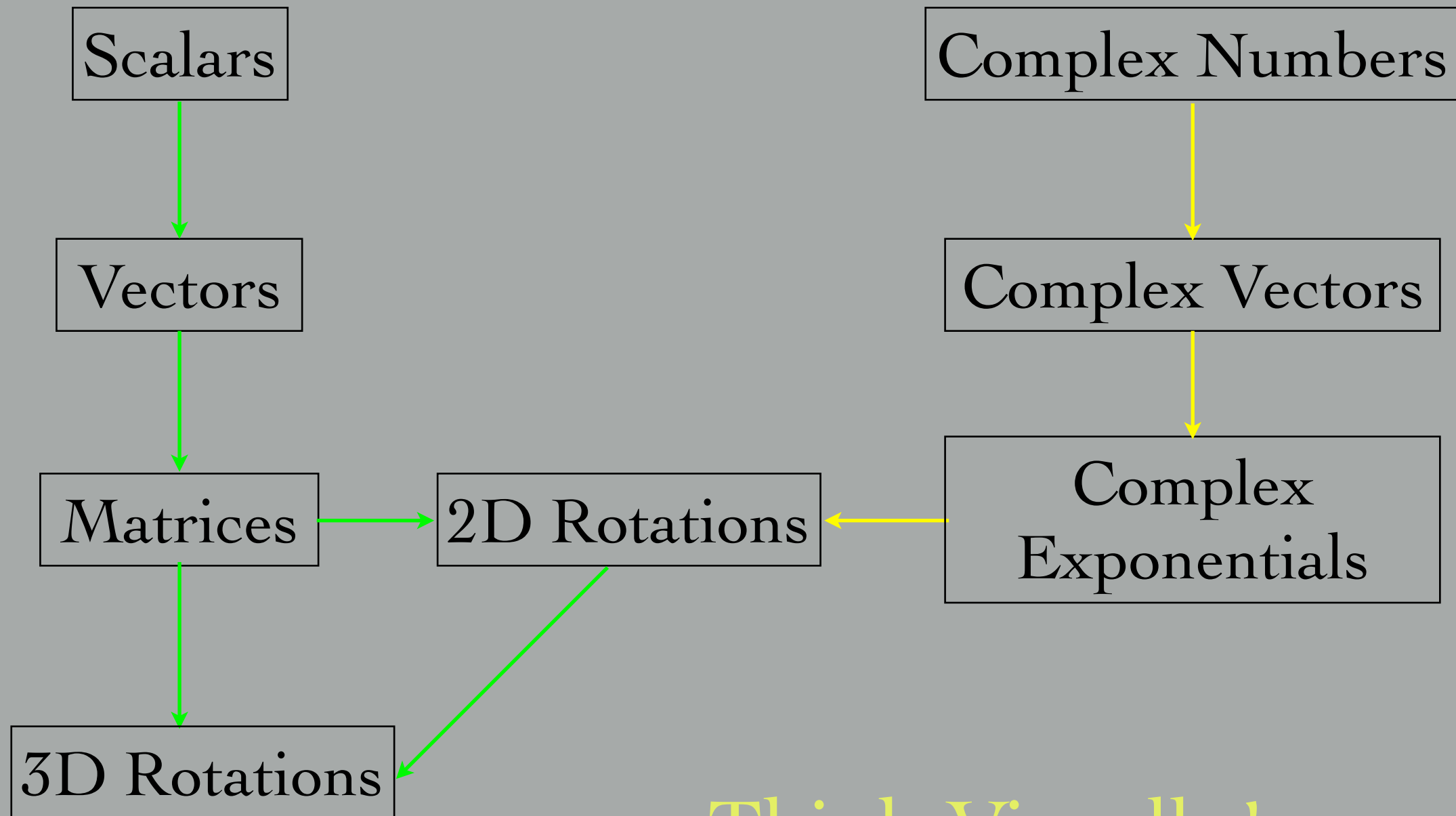


Lecture 3

Vectors, Matrices, and Complex Numbers

Lecture Summary



Think Visually!

Lecture Summary

Most of this chapter is just
packaging and bookkeeping

Vectors



ball has a *direction* and a *velocity*

Cartesian coordinate system



first, we need a coordinate system

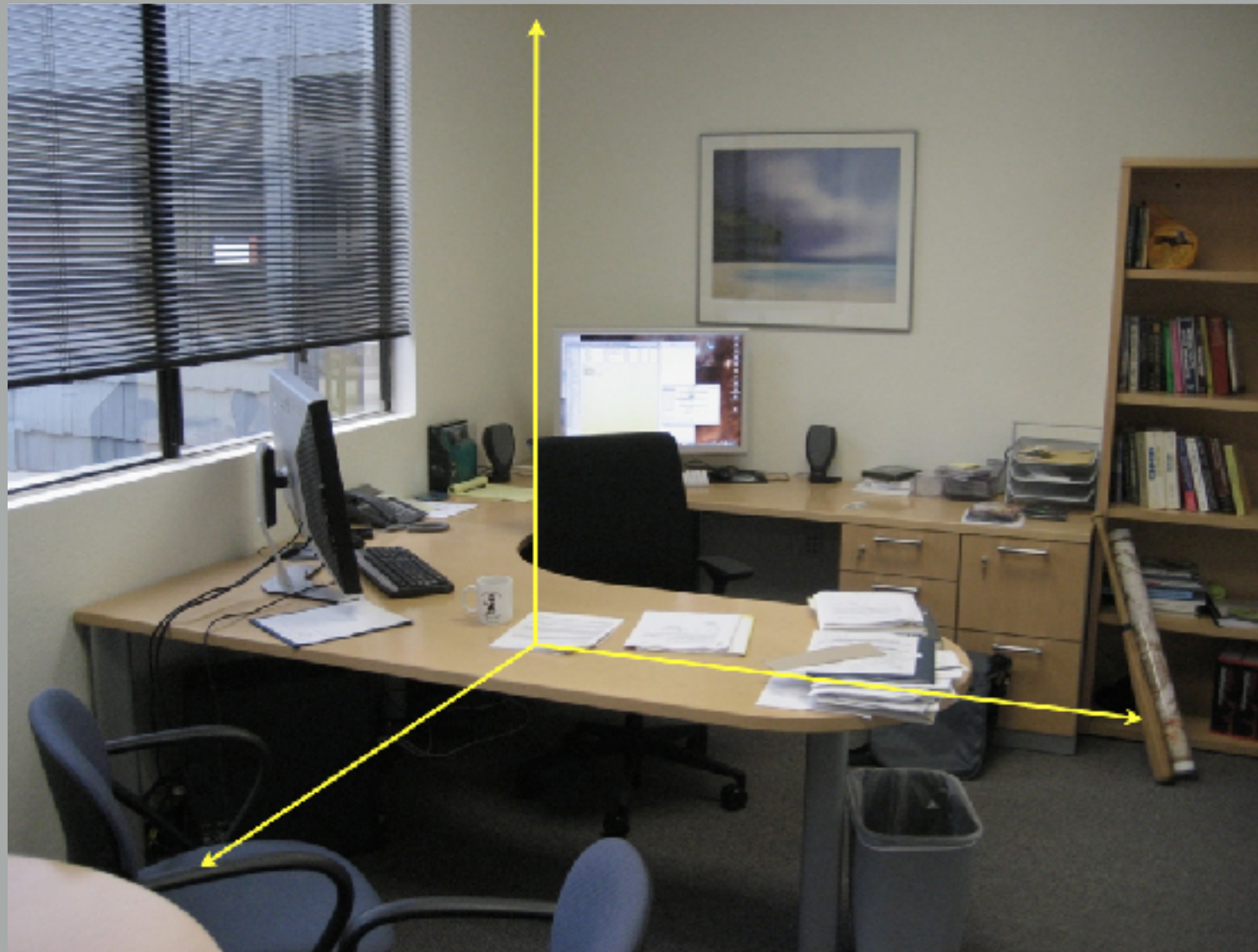
Cartesian coordinate system



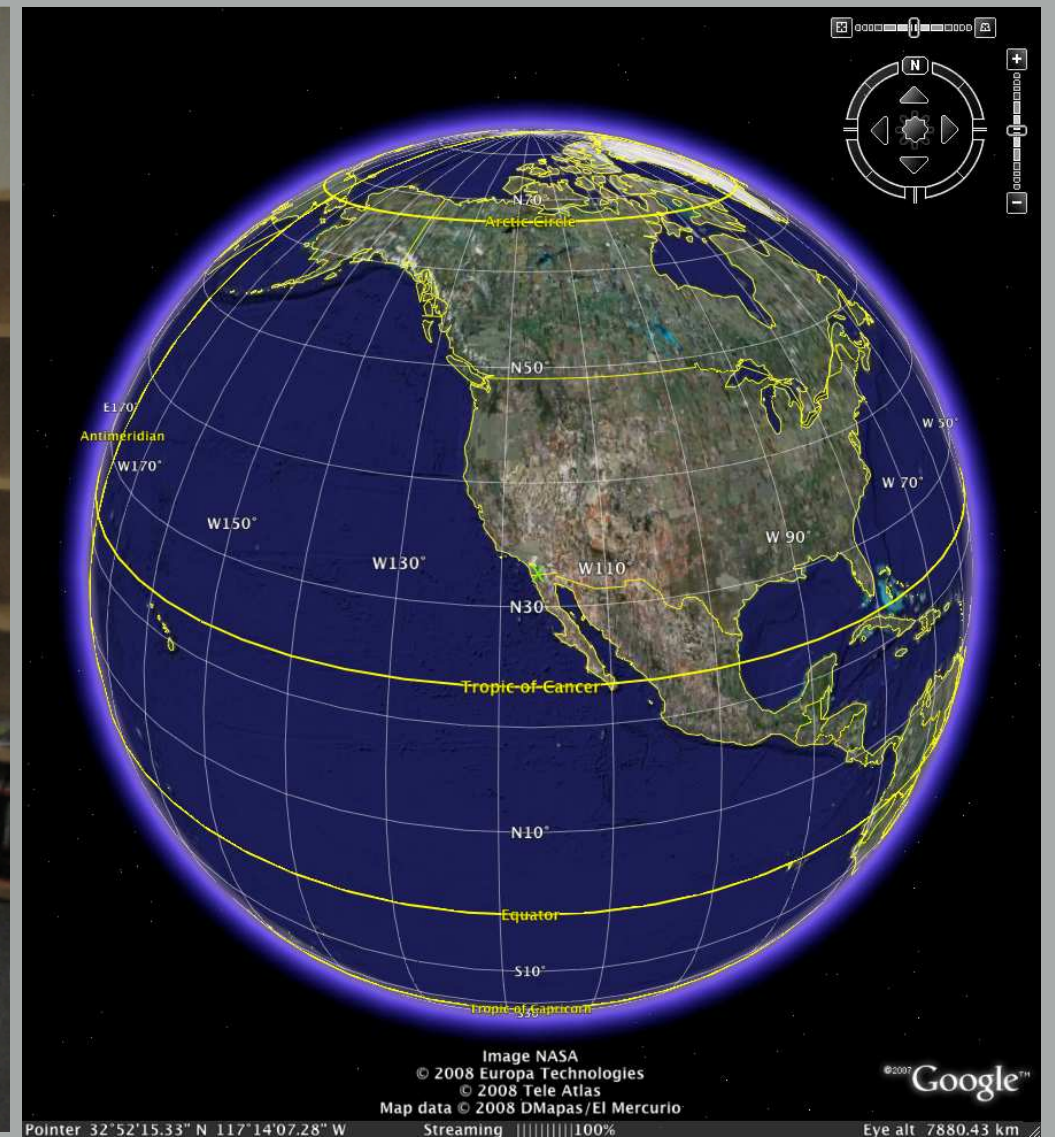
Vector in 3D



Coordinate Systems

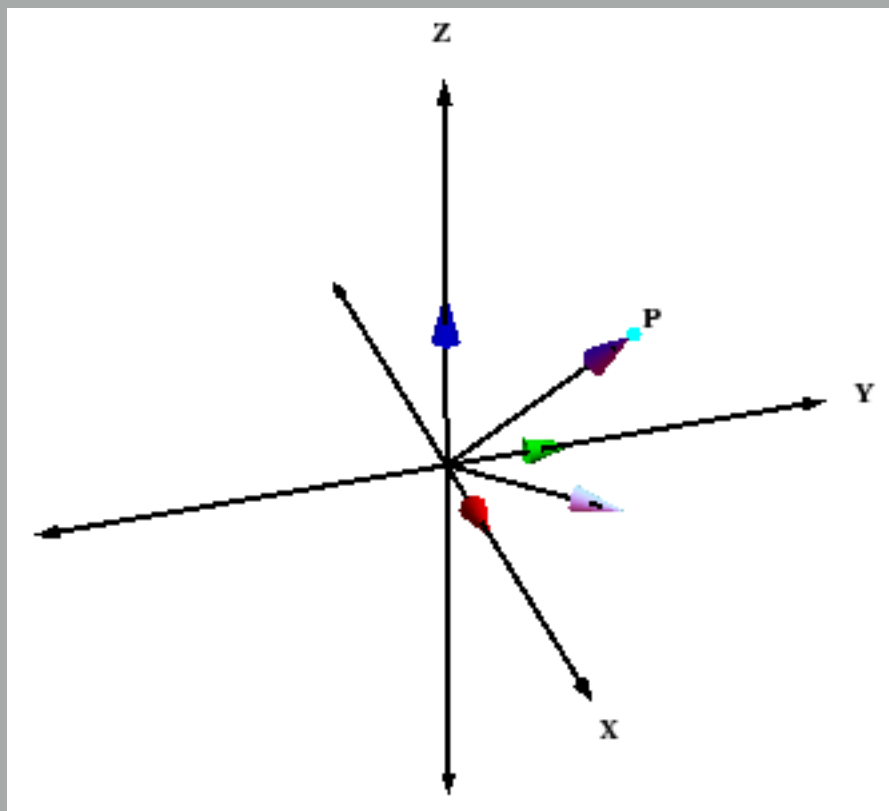


Cartesian

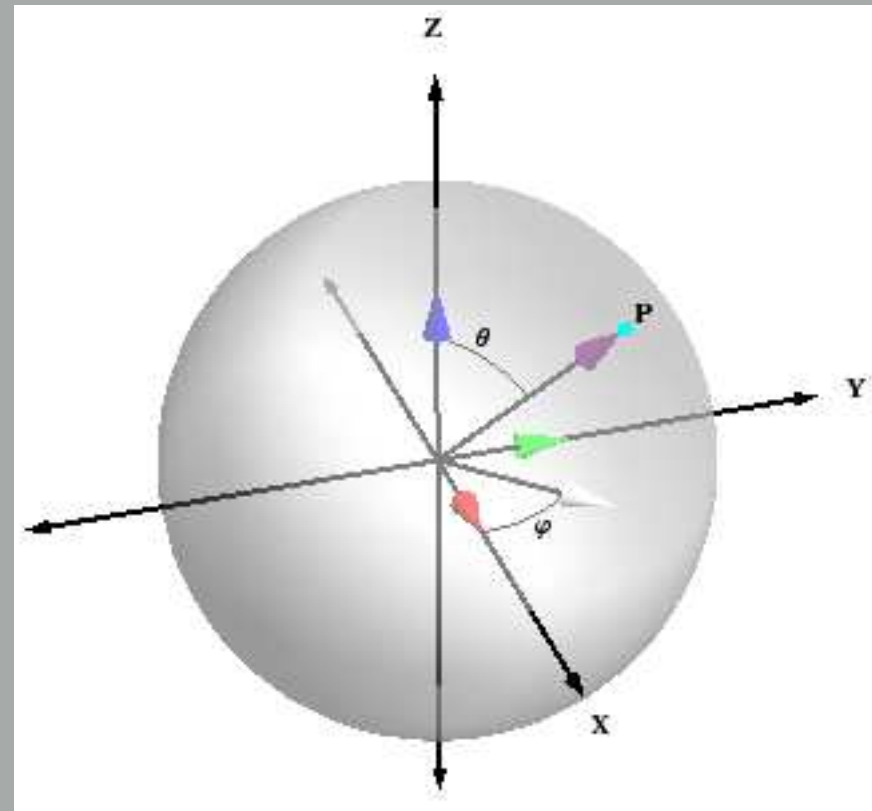


Spherical

Coordinate Systems

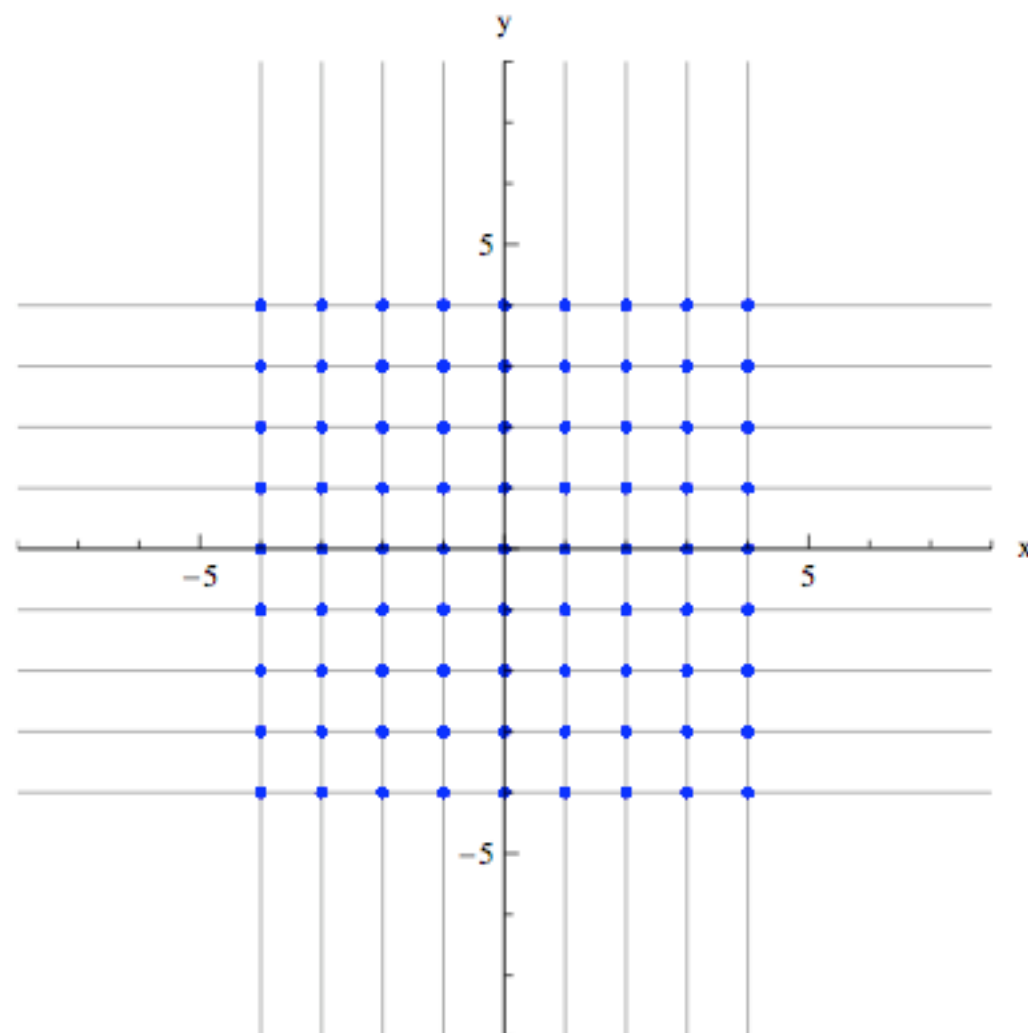


Cartesian

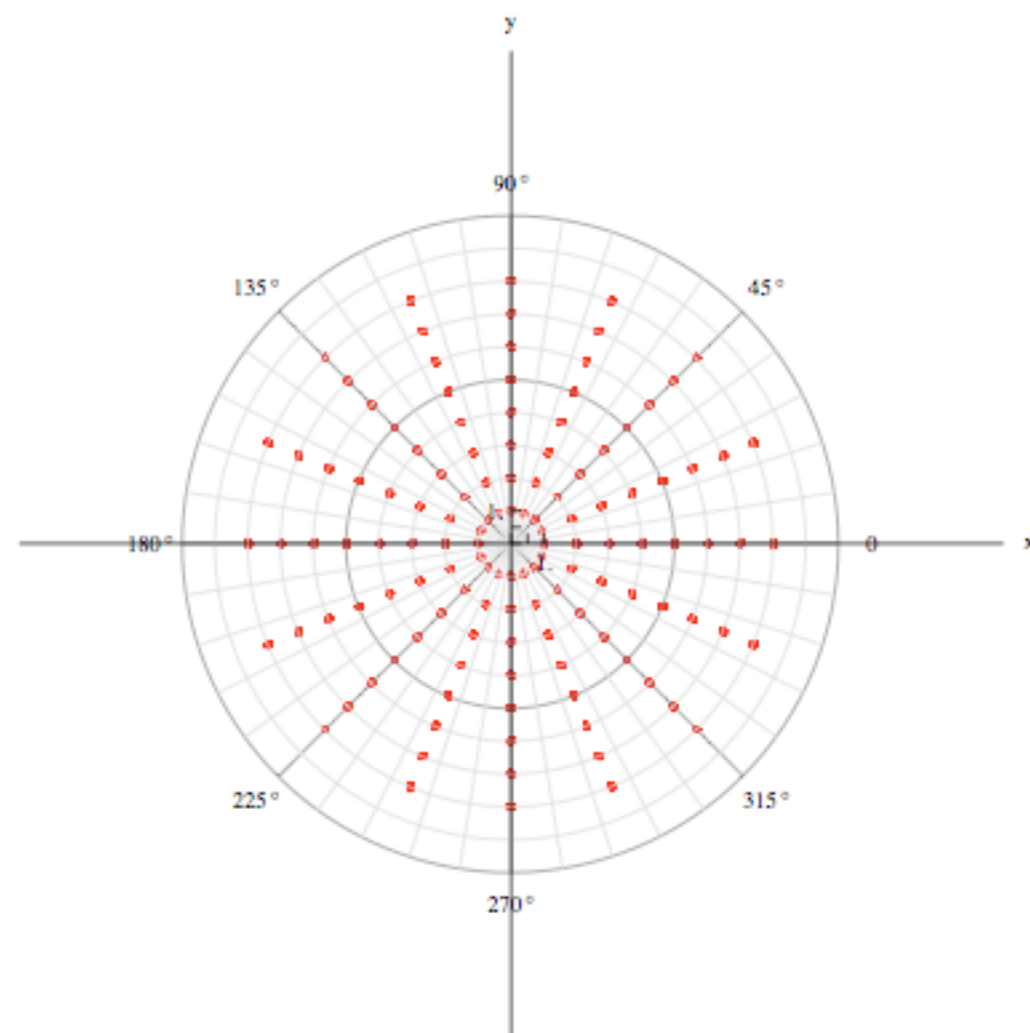


Spherical

Example: MRI



(a) Cartesian grid.



(b) Polar grid.

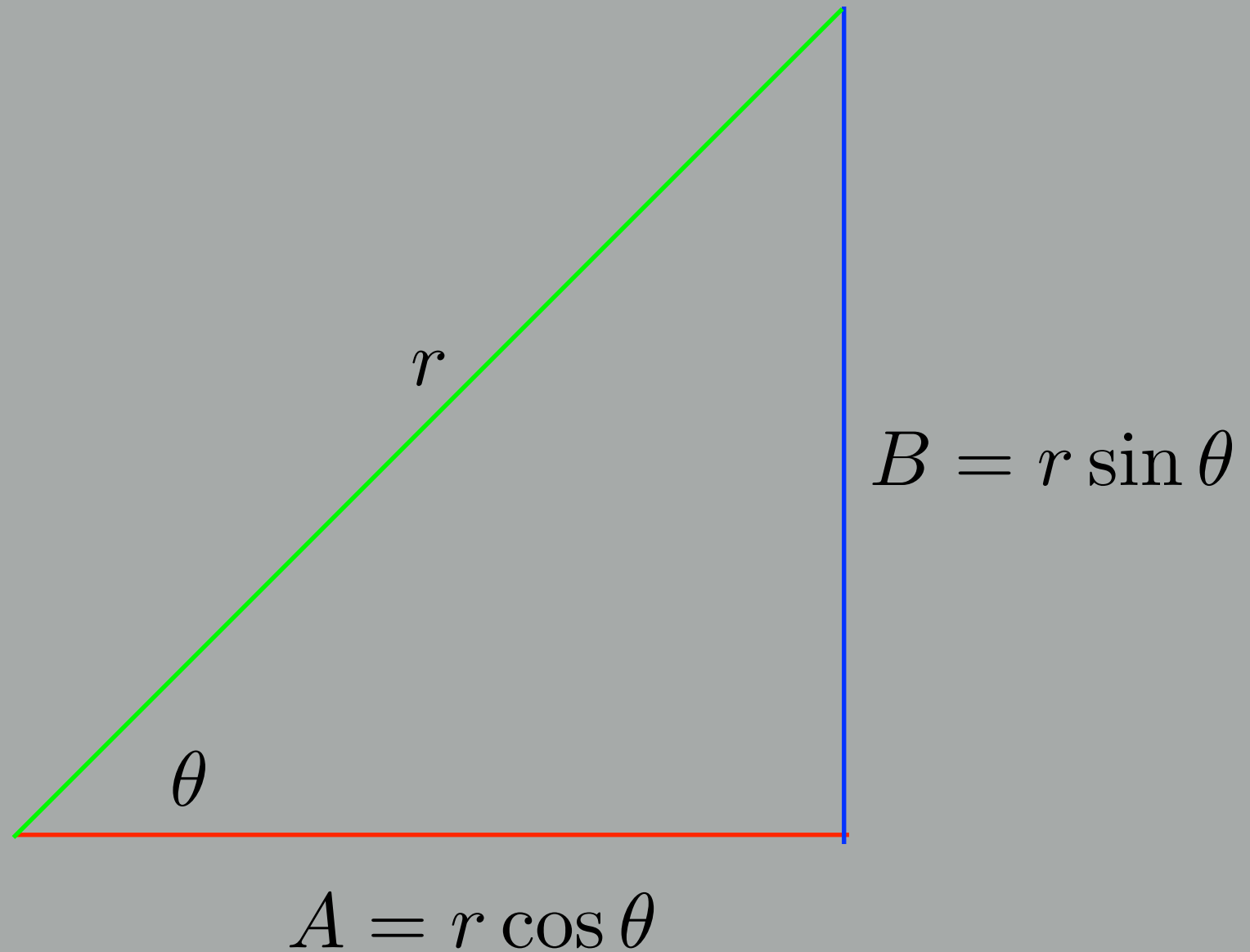
Echo planar

Spiral

Trigonometry

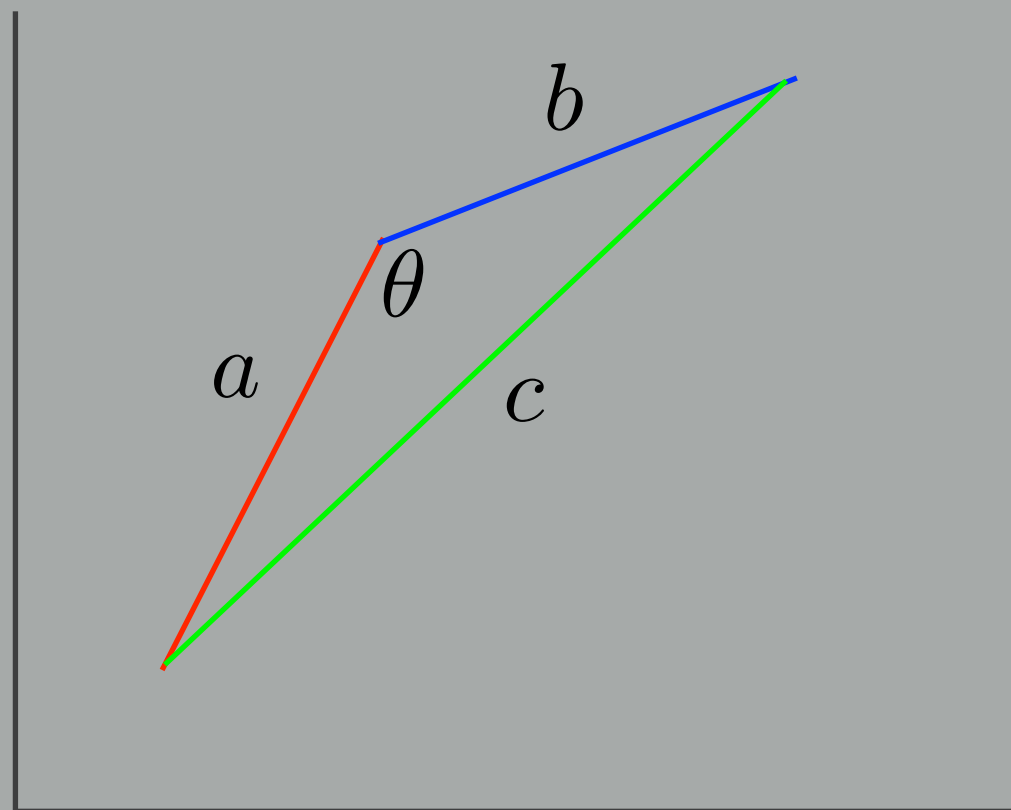
$$\cos \theta = A/r$$

$$\sin \theta = B/r$$



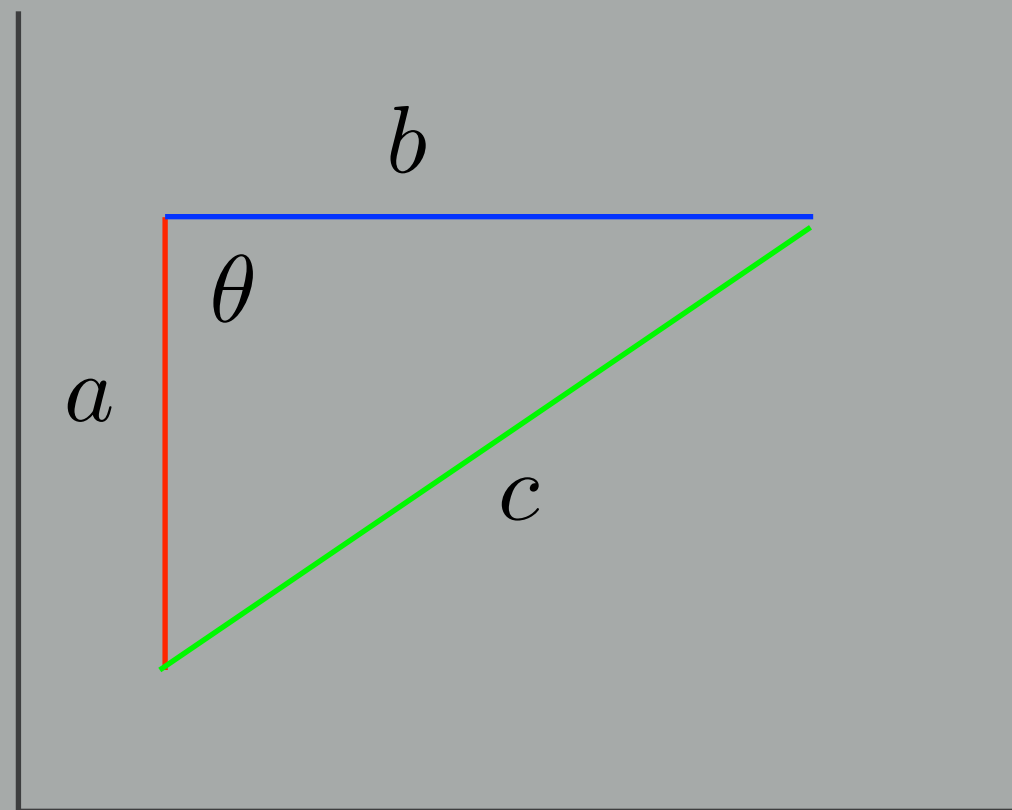
Law of Cosines

$$c^2 = a^2 + b^2 - 2ab \cos \theta$$



Law of Cosines

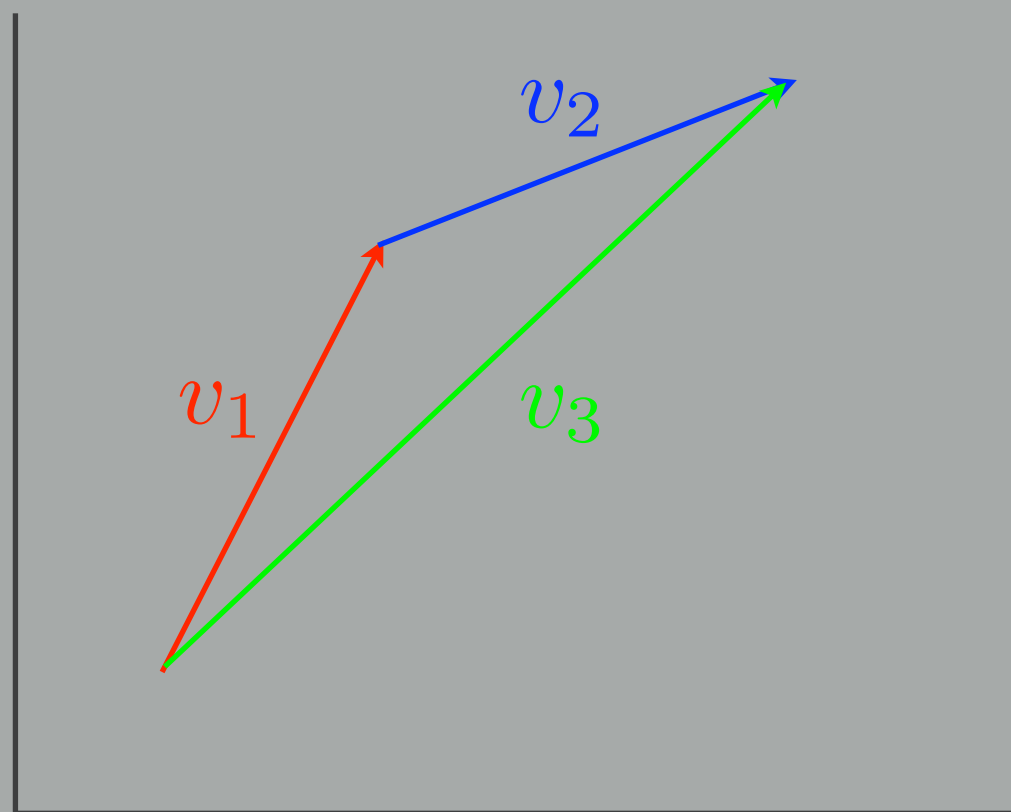
$$c^2 = a^2 + b^2 \quad (\text{ref: Pythagoras})$$



For right triangle: $\cos 90^\circ = 0$

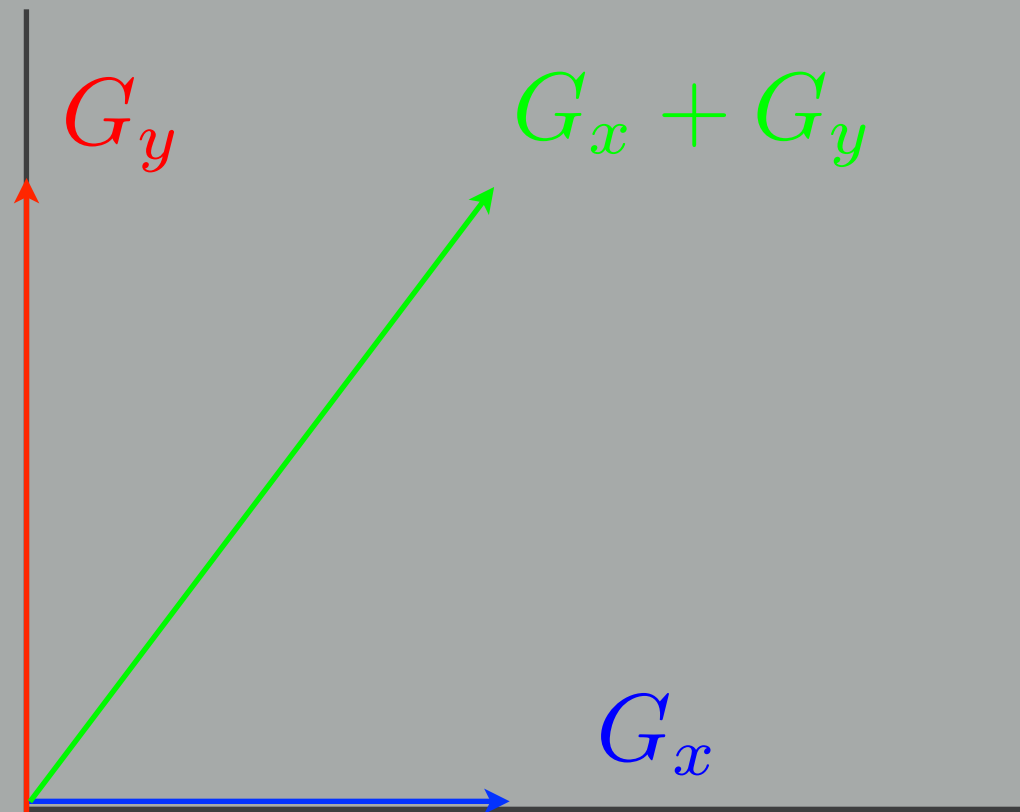
Vector Addition

$$\mathbf{v}_3 = \mathbf{v}_1 + \mathbf{v}_2$$



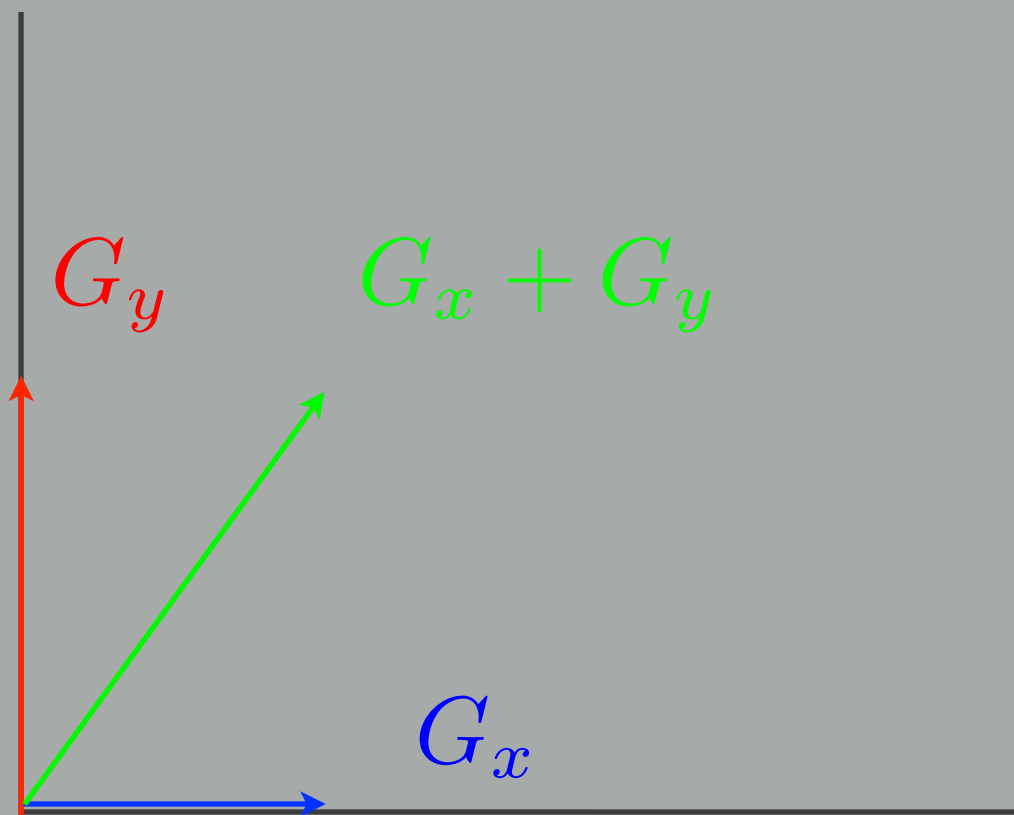
Example

MRI field gradient add like vectors

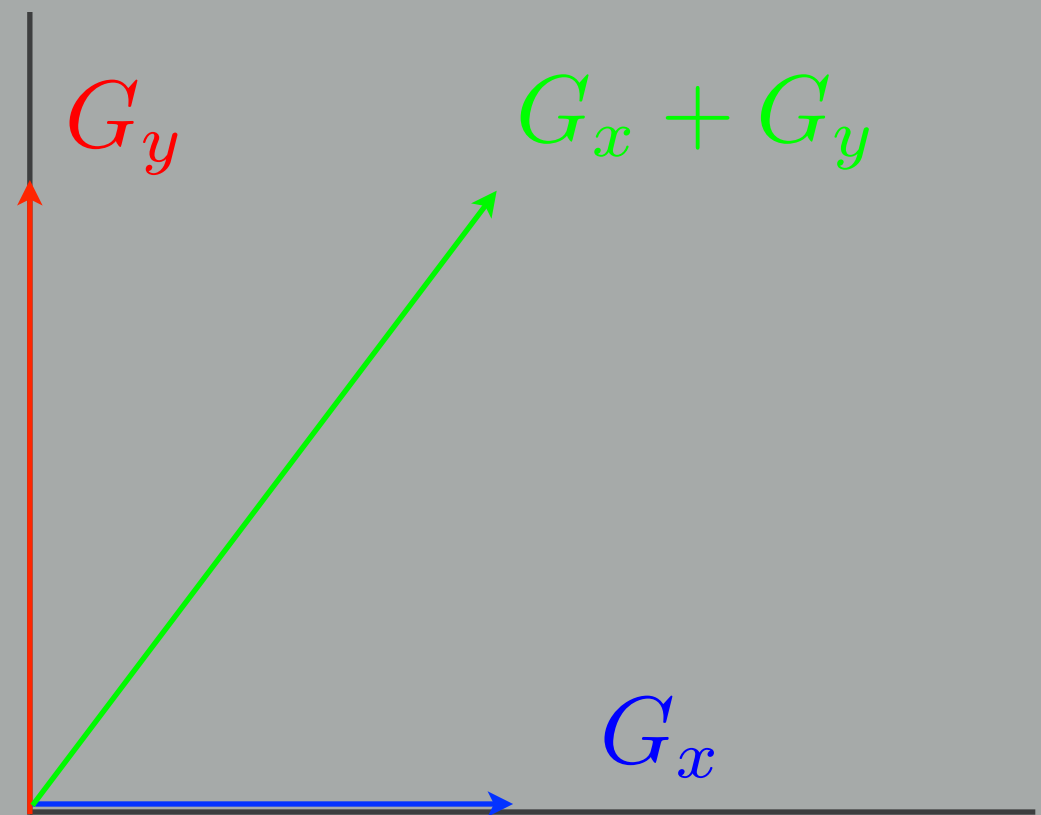


Example

Strength of applied gradients

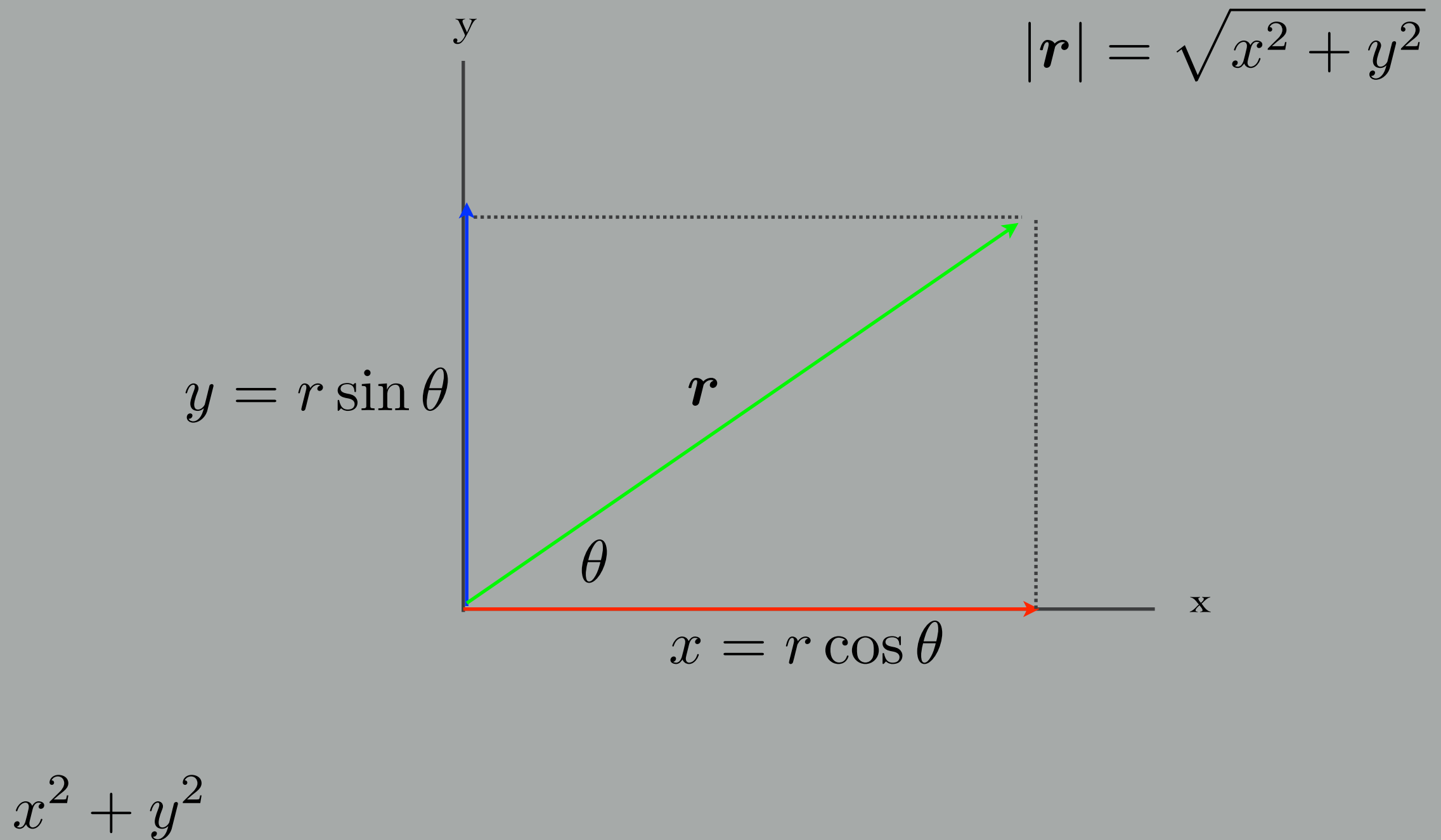


weaker

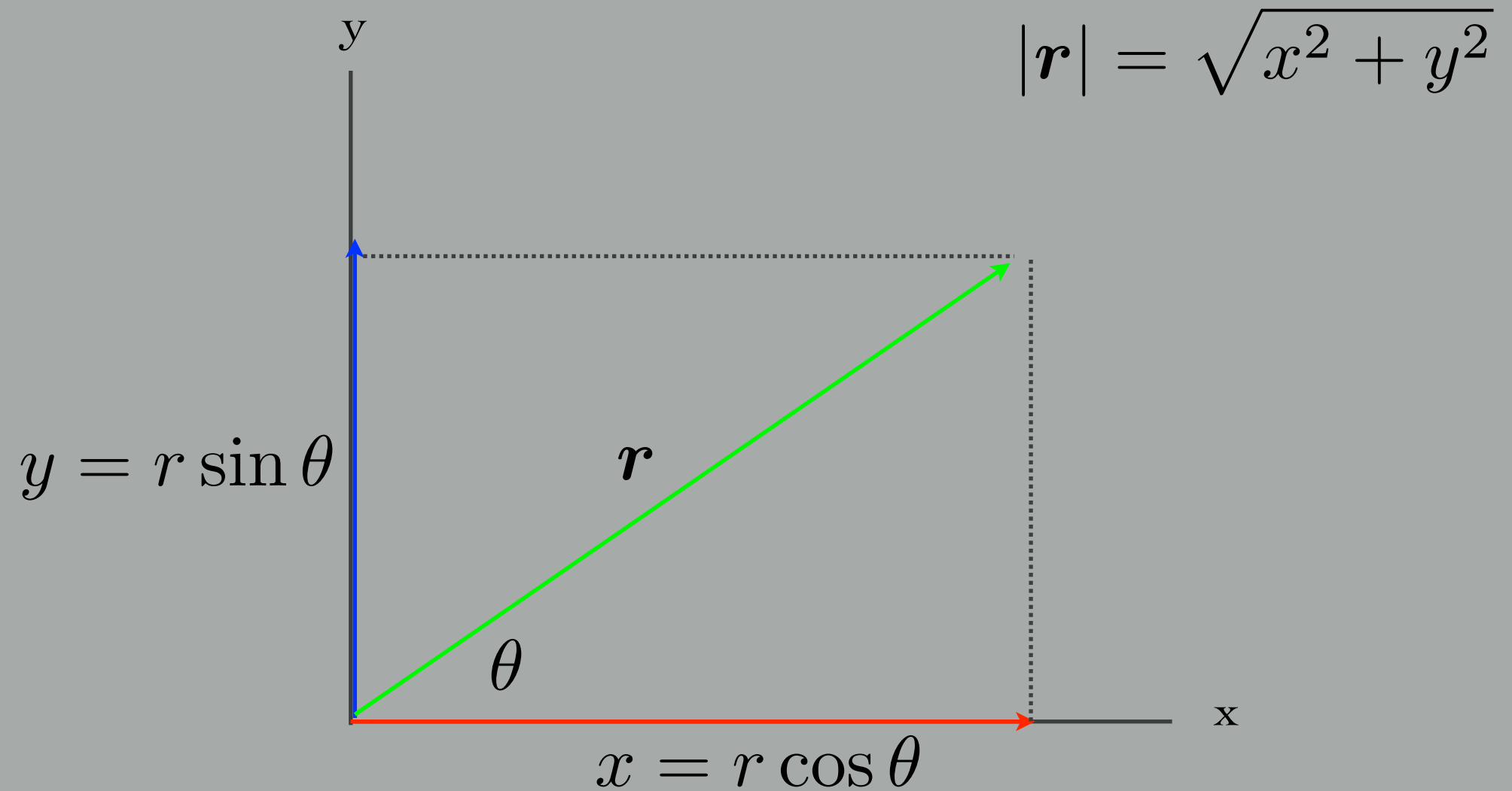


stronger

Vector Length (Orthogonal)

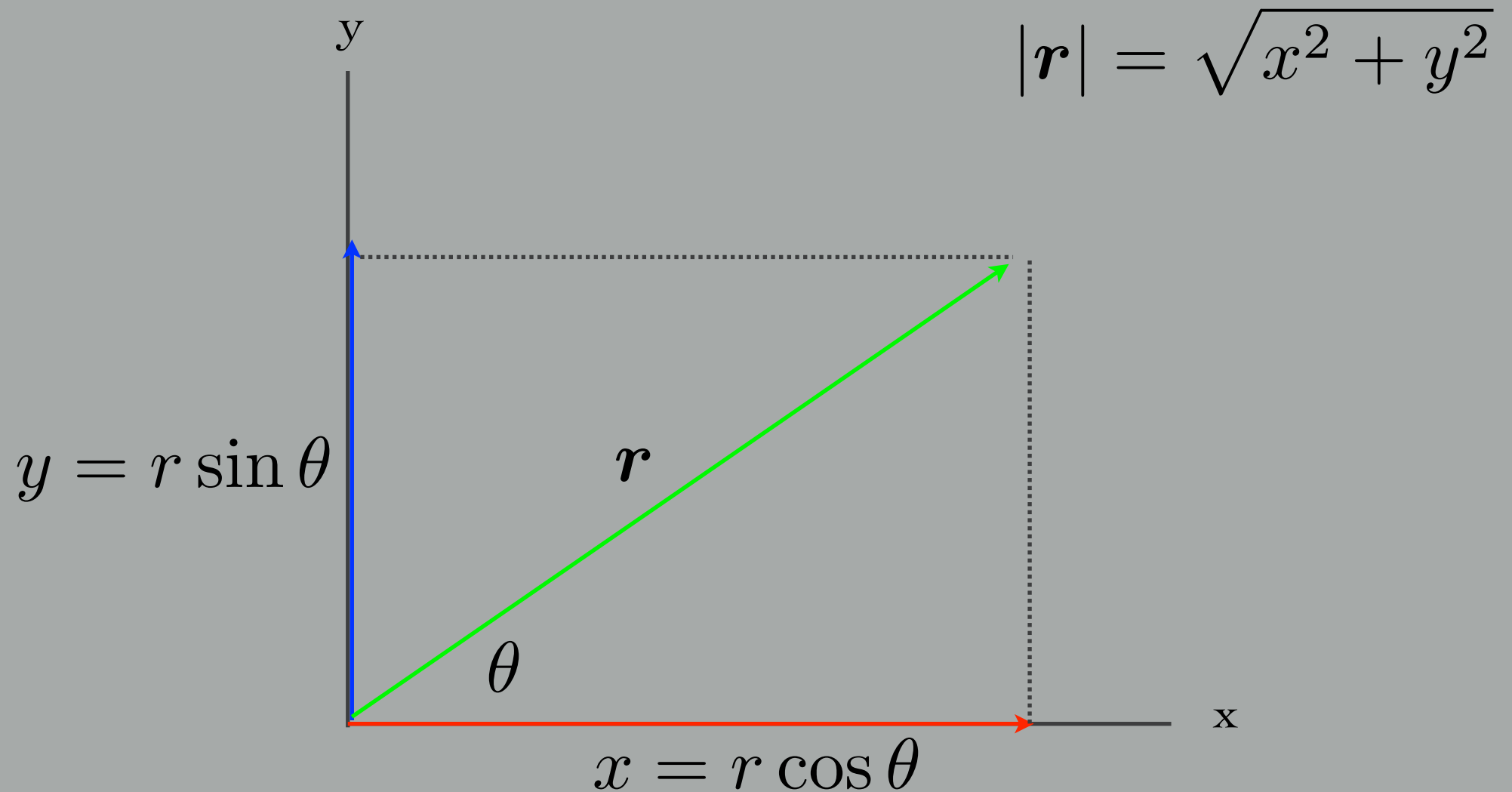


Vector Length (Orthogonal)



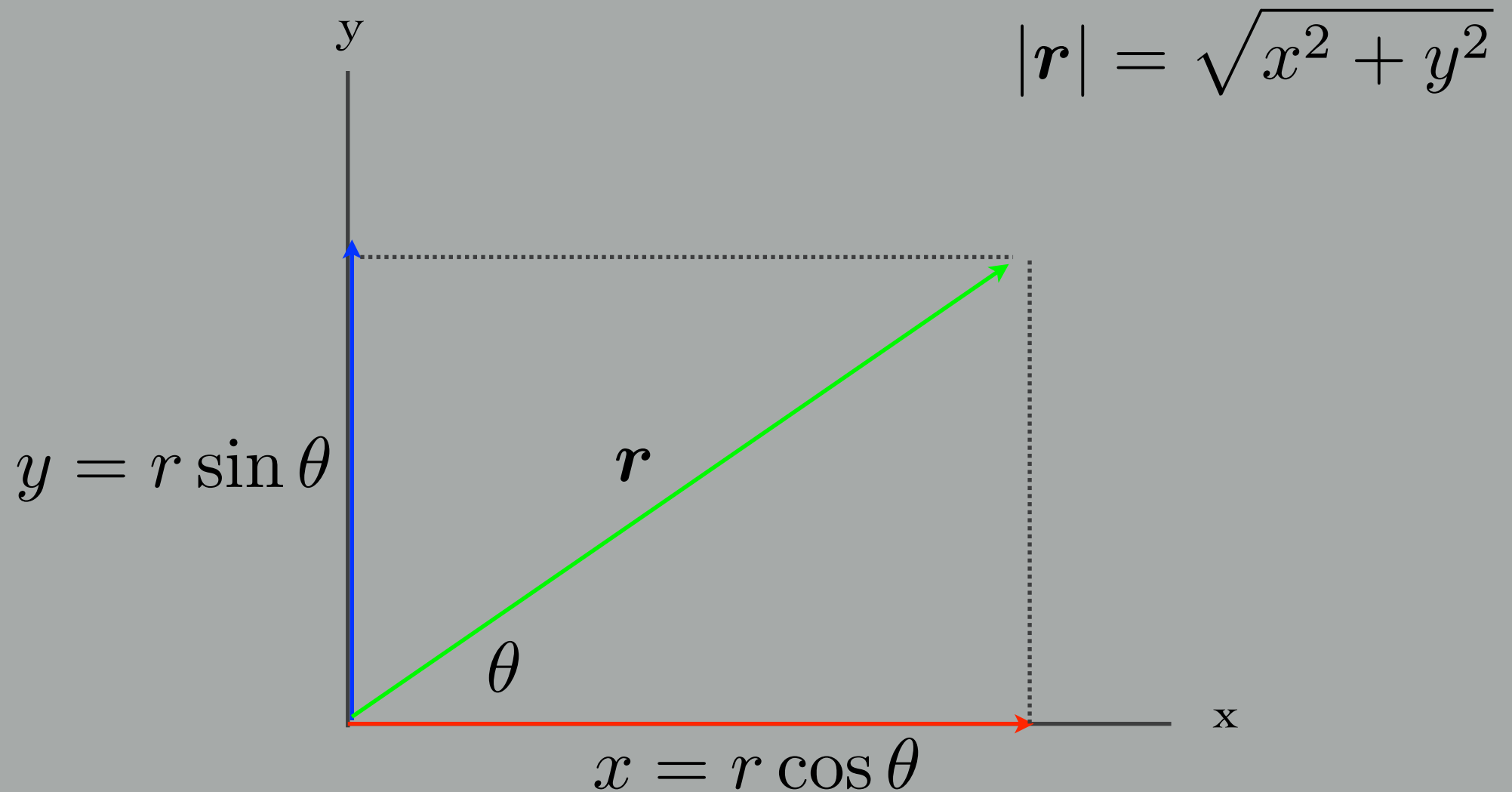
$$x^2 + y^2 = r^2 \cos^2 \theta + r^2 \sin^2 \theta$$

Vector Length (Orthogonal)



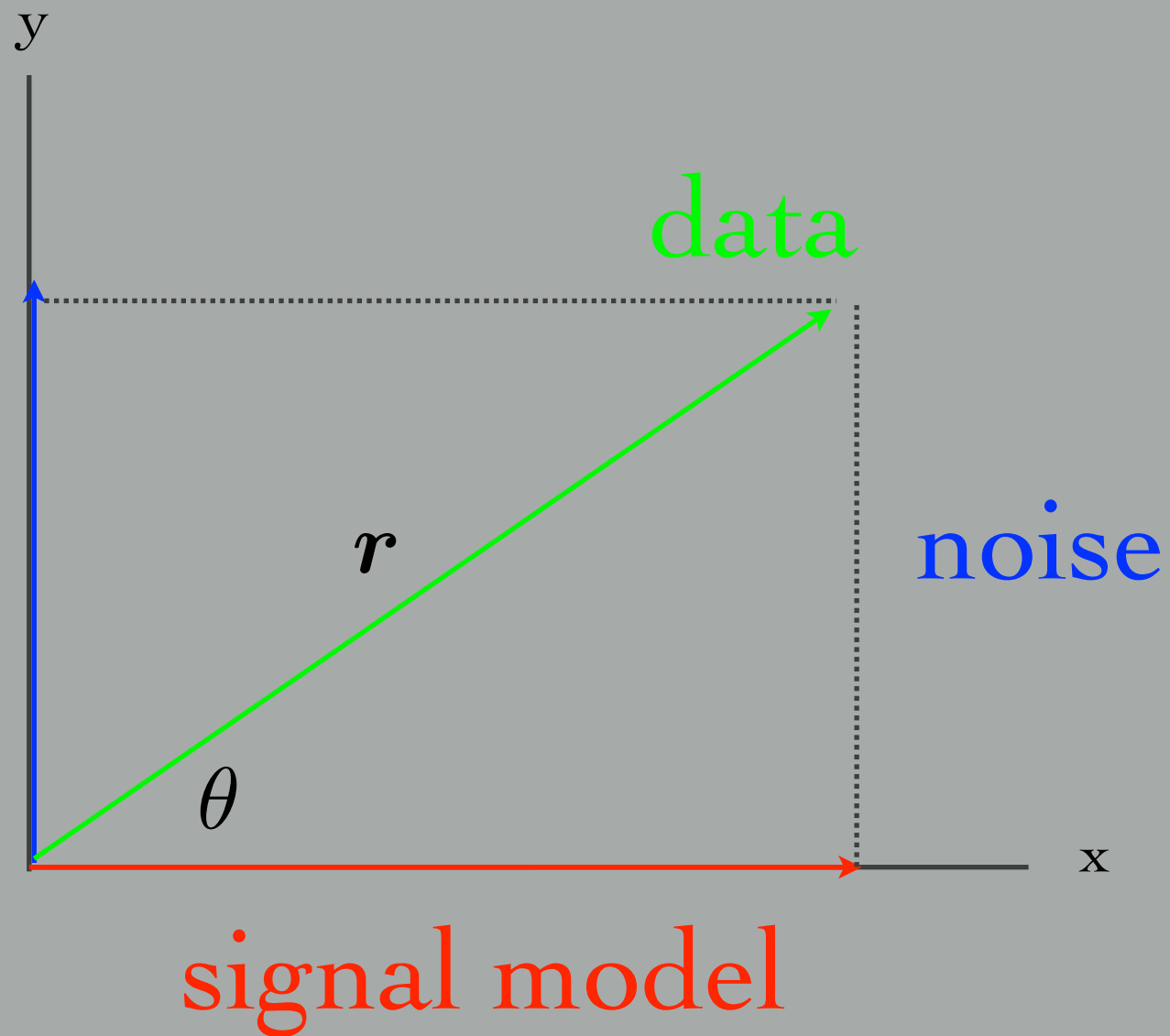
$$x^2 + y^2 = r^2 \cos^2 \theta + r^2 \sin^2 \theta = r^2 \underbrace{(\cos^2 \theta + \sin^2 \theta)}_1$$

Vector Length (Orthogonal)

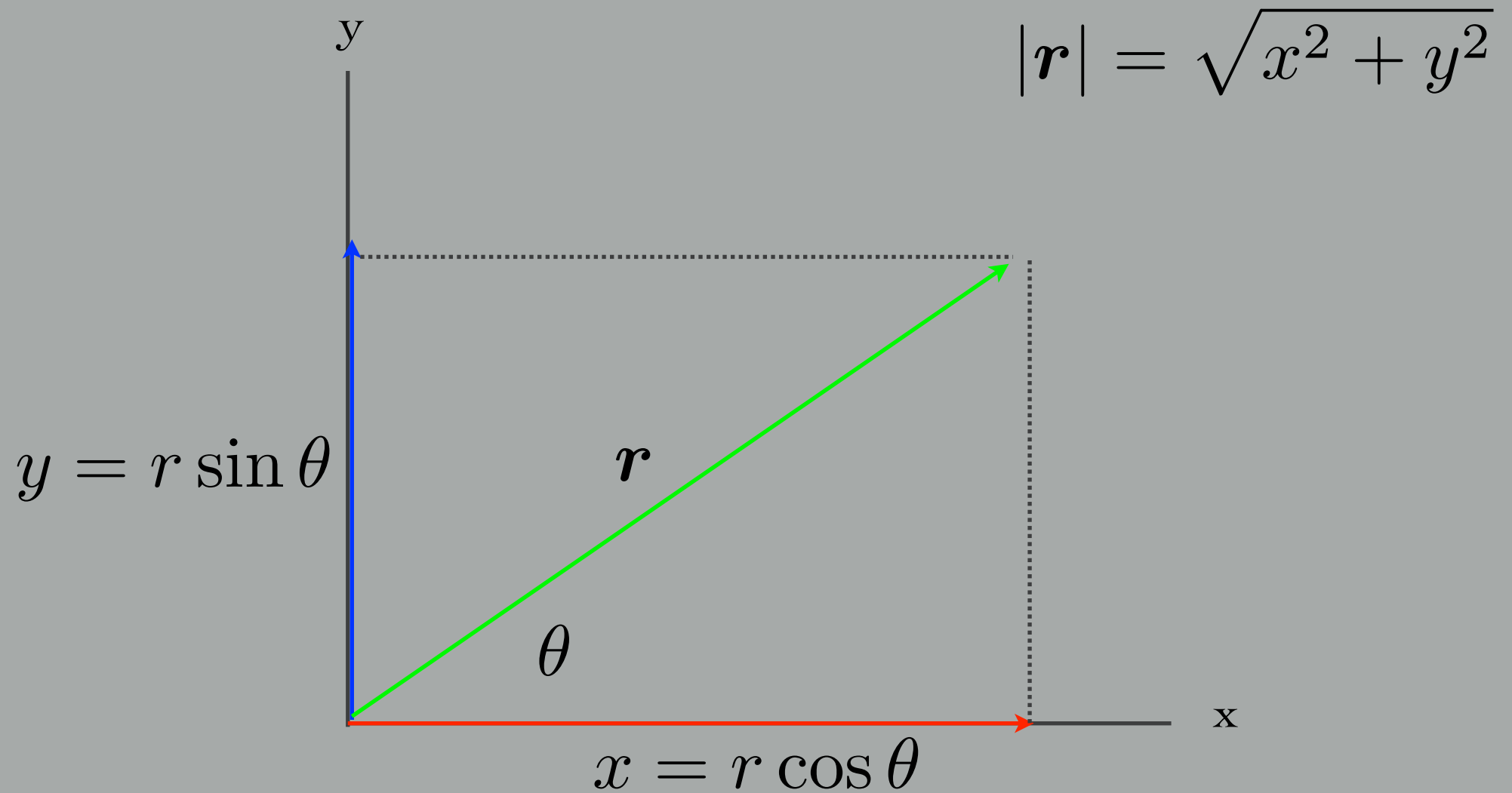


$$x^2 + y^2 = r^2 \cos^2 \theta + r^2 \sin^2 \theta = r^2 \underbrace{(\cos^2 \theta + \sin^2 \theta)}_1 = r^2$$

Example: Least Squares

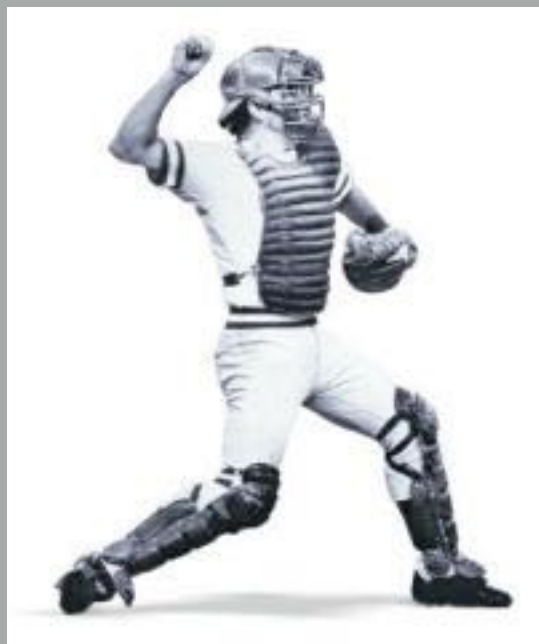
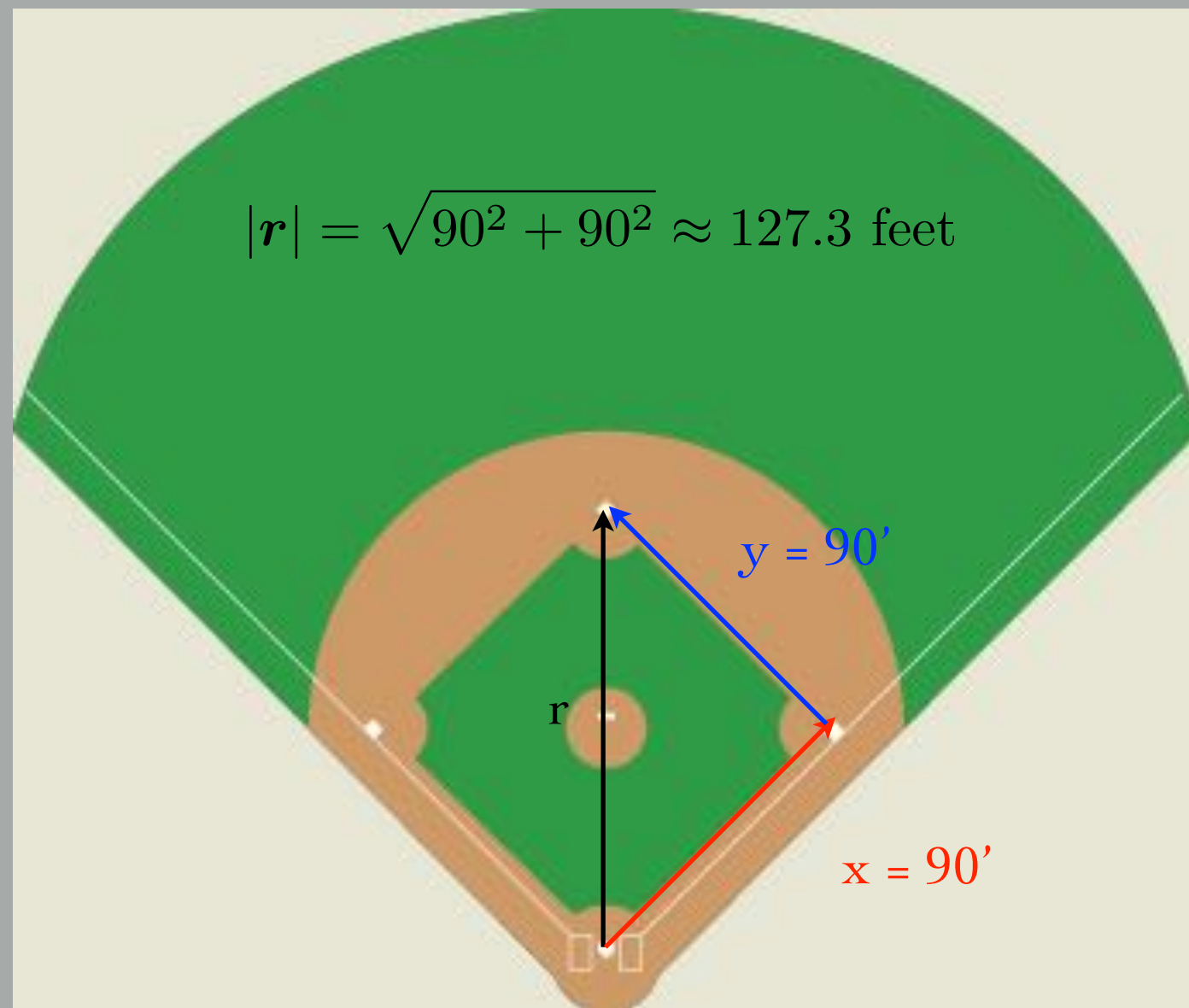


Vector Angle



$$\tan \theta = \frac{y}{x} \quad \longrightarrow \quad \theta = \tan^{-1} \left(\frac{y}{x} \right)$$

Vector Lengths



catcher throws to second

Vector Notation

$$\mathbf{r} = \begin{pmatrix} x \\ y \end{pmatrix} \quad \text{column vector}$$

$$\mathbf{q} = (a \ , \ b) \quad \text{row vector}$$

$$\mathbf{r}^t = (x \ , \ y) \quad \begin{array}{l} \text{vector } \textit{transpose} \\ \text{(exchange rows \& columns)} \end{array}$$

often written in general form with row subscript

$$\mathbf{r} = \begin{pmatrix} r_1 \\ r_2 \end{pmatrix}$$

Vector and Equations

the line

$$y = \alpha + \beta x$$

Vector and Equations

the line

$$y = \alpha + \beta x$$

Write in vector form:

$$\mathbf{a} = \begin{pmatrix} \alpha \\ \beta \end{pmatrix} \quad \mathbf{f} = \begin{pmatrix} 1 \\ x \end{pmatrix}$$

Vector and Equations

the line

$$y = \alpha + \beta x$$

Write in vector form:

$$\mathbf{a} = \begin{pmatrix} \alpha \\ \beta \end{pmatrix} \quad \mathbf{f} = \begin{pmatrix} 1 \\ x \end{pmatrix}$$

“amplitudes”

“model functions”

Vector and Equations

the line

$$y = \alpha + \beta x$$

Write in vector form:

$$\mathbf{a} = \begin{pmatrix} \alpha \\ \beta \end{pmatrix} \quad \mathbf{f} = \begin{pmatrix} 1 \\ x \end{pmatrix}$$

Vector and Equations

the line

$$y = \alpha + \beta x$$

Write in vector form:

$$\mathbf{a} = \begin{pmatrix} \alpha \\ \beta \end{pmatrix} \quad \mathbf{f} = \begin{pmatrix} 1 \\ x \end{pmatrix}$$

Vector Dot Product

$$\mathbf{a} = \begin{pmatrix} \alpha \\ \beta \end{pmatrix} \quad \mathbf{f} = \begin{pmatrix} 1 \\ x \end{pmatrix}$$

Vector multiplication rule:
Row elements times column elements and add

$$y = (\alpha \quad \beta) \begin{pmatrix} 1 \\ x \end{pmatrix}$$

Vector Dot Product

$$\mathbf{a} = \begin{pmatrix} \alpha \\ \beta \end{pmatrix} \quad \mathbf{f} = \begin{pmatrix} 1 \\ x \end{pmatrix}$$

Vector multiplication rule:
Row elements times column elements and add

$$y = (\alpha \quad \beta) \begin{pmatrix} 1 \\ x \end{pmatrix} = \alpha$$

Vector Dot Product

$$\mathbf{a} = \begin{pmatrix} \alpha \\ \beta \end{pmatrix} \quad \mathbf{f} = \begin{pmatrix} 1 \\ x \end{pmatrix}$$

Vector multiplication rule:
Row elements times column elements and add

$$y = (\alpha \quad \beta) \begin{pmatrix} 1 \\ x \end{pmatrix} = \alpha +$$

Vector Dot Product

$$\mathbf{a} = \begin{pmatrix} \alpha \\ \beta \end{pmatrix} \quad \mathbf{f} = \begin{pmatrix} 1 \\ x \end{pmatrix}$$

Vector multiplication rule:
Row elements times column elements and add

$$y = (\alpha \quad \beta) \begin{pmatrix} 1 \\ x \end{pmatrix} = \alpha + \beta x$$

Vector Dot Product

$$\mathbf{a} = \begin{pmatrix} \alpha \\ \beta \end{pmatrix} \quad \mathbf{f} = \begin{pmatrix} 1 \\ x \end{pmatrix}$$

Vector multiplication rule:
Row elements times column elements and add

$$y = \begin{pmatrix} \alpha & \beta \end{pmatrix} \begin{pmatrix} 1 \\ x \end{pmatrix} = \alpha + \beta x$$

columns here must equal *rows* here

Vector Dot Product

$$\mathbf{a} = \begin{pmatrix} \alpha \\ \beta \end{pmatrix} \quad \mathbf{f} = \begin{pmatrix} 1 \\ x \end{pmatrix}$$

Vector multiplication rule:
Row elements times column elements and add

$$y = (\alpha \quad \beta) \begin{pmatrix} 1 \\ x \end{pmatrix} = \alpha + \beta x$$

$$y = \mathbf{a} \cdot \mathbf{f} = \mathbf{a}^t \mathbf{f}$$

Vector and Equations

Vector Dot Product

$$y = a \cdot f = \alpha 1 + \beta x$$

$$y = \sum_{i=1}^2 a_i f_i$$

Vector dot product

linear equation

$$f = ax + by + cz$$

Write in vector form:

$$\mathbf{u} = \begin{pmatrix} a \\ b \\ c \end{pmatrix} \quad \mathbf{v} = \begin{pmatrix} x \\ y \\ z \end{pmatrix}$$

$$f = \underbrace{(x \quad y \quad z)}_{\mathbf{u}^t} \underbrace{\begin{pmatrix} x \\ y \\ z \end{pmatrix}}_{\mathbf{v}} = \mathbf{u}^t \mathbf{v}$$

Vector dot product

linear equation

$$f = ax + by + cz$$

Write in vector form:

$$\mathbf{u} = \begin{pmatrix} a \\ b \\ c \end{pmatrix} \quad \mathbf{v} = \begin{pmatrix} x \\ y \\ z \end{pmatrix}$$

$$f = \underbrace{(x \quad y \quad z)}_{\mathbf{u}^t} \underbrace{\begin{pmatrix} x \\ y \\ z \end{pmatrix}}_{\mathbf{v}} = \mathbf{u}^t \mathbf{v}$$

Vector dot product

linear equation

$$f = ax + by + cz$$

Write in vector form:

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$$f = \underbrace{(x \quad y \quad z)}_{\mathbf{u}^t} \underbrace{\begin{pmatrix} x \\ y \\ z \end{pmatrix}}_{\mathbf{v}} = \mathbf{u}^t \mathbf{v}$$

Vector dot product

linear equation

$$f = ax + by + cz$$

Write in vector form:

$$\mathbf{u} = \begin{pmatrix} a \\ b \\ c \end{pmatrix} \quad \mathbf{v} = \begin{pmatrix} x \\ y \\ z \end{pmatrix}$$

$$f = \underbrace{(x \ y \ z)}_{\mathbf{u}^t} \underbrace{\begin{pmatrix} x \\ y \\ z \end{pmatrix}}_{\mathbf{v}} = \mathbf{u}^t \mathbf{v}$$

Vector Length, revisited

$$\mathbf{r} = \begin{pmatrix} x \\ y \end{pmatrix} \quad \text{column vector}$$

$$\mathbf{r}^t = (x \ , \ y) \quad \text{vector } \textit{transpose}$$

$$\mathbf{r}^t \mathbf{r} = (x \ , \ y) \begin{pmatrix} x \\ y \end{pmatrix} = x^2 + y^2$$

“therefore”


$$\therefore \boxed{\|\mathbf{r}\| = \sqrt{\mathbf{r}^t \mathbf{r}}}$$

Vector in 3D



Vector Length



$$\|\mathbf{r}\| = \sqrt{x^2 + y^2 + z^2} = \sqrt{\mathbf{r}^t \mathbf{r}}$$

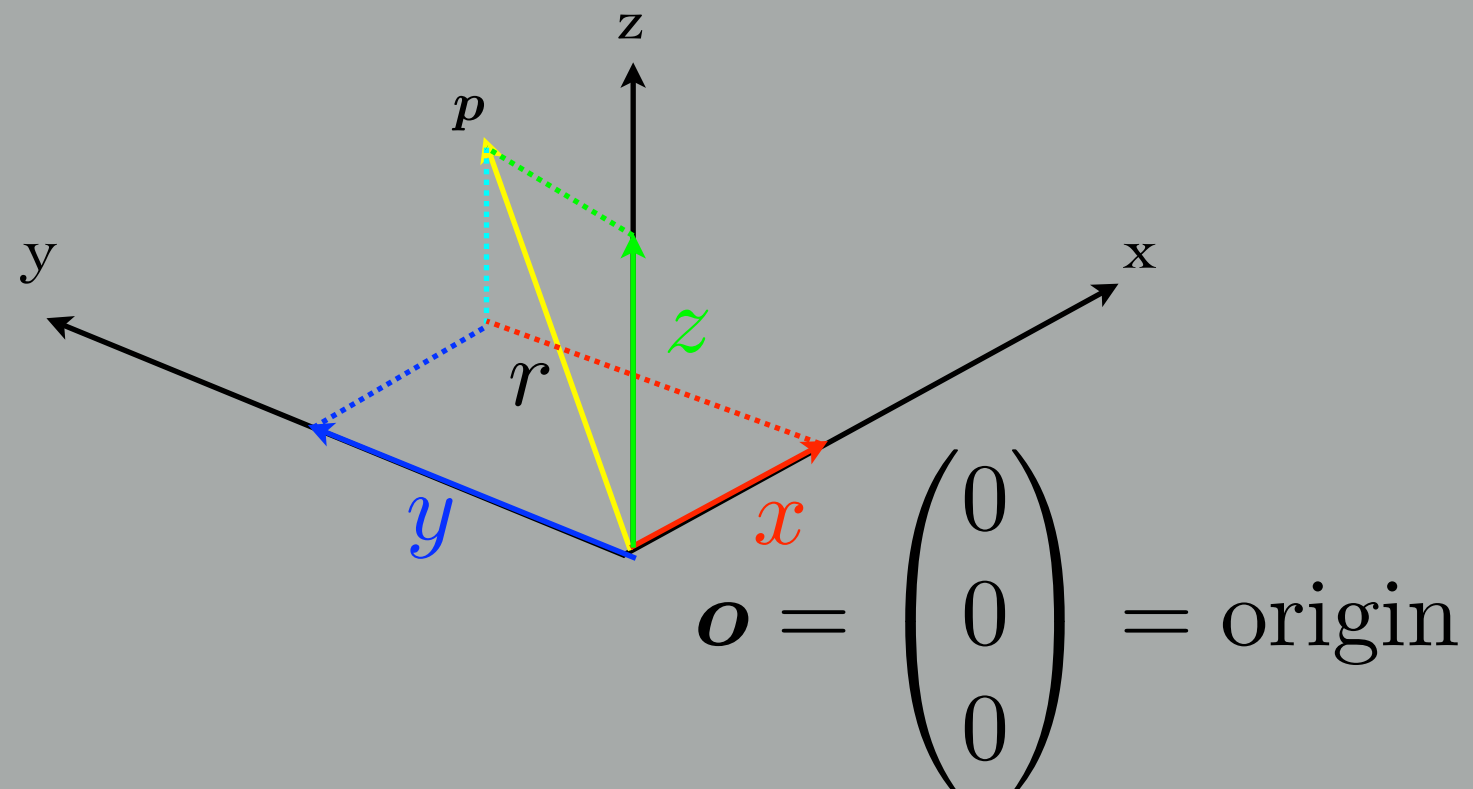
Vectors in 3D

Position vector

$$\mathbf{p} = \begin{pmatrix} x \\ y \\ z \end{pmatrix}$$

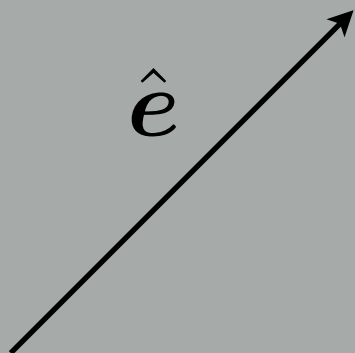
Velocity vector

$$\mathbf{v} = \begin{pmatrix} v_x \\ v_y \\ v_z \end{pmatrix}$$

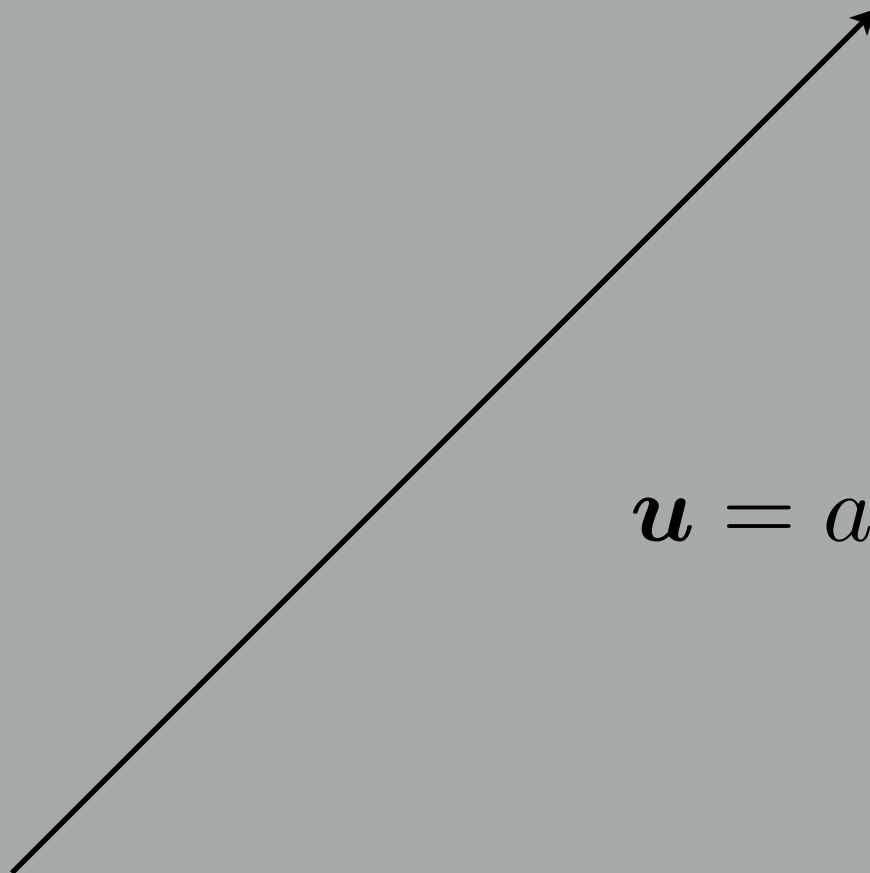


Vector-Scalar Multiplication

the *unit* vector

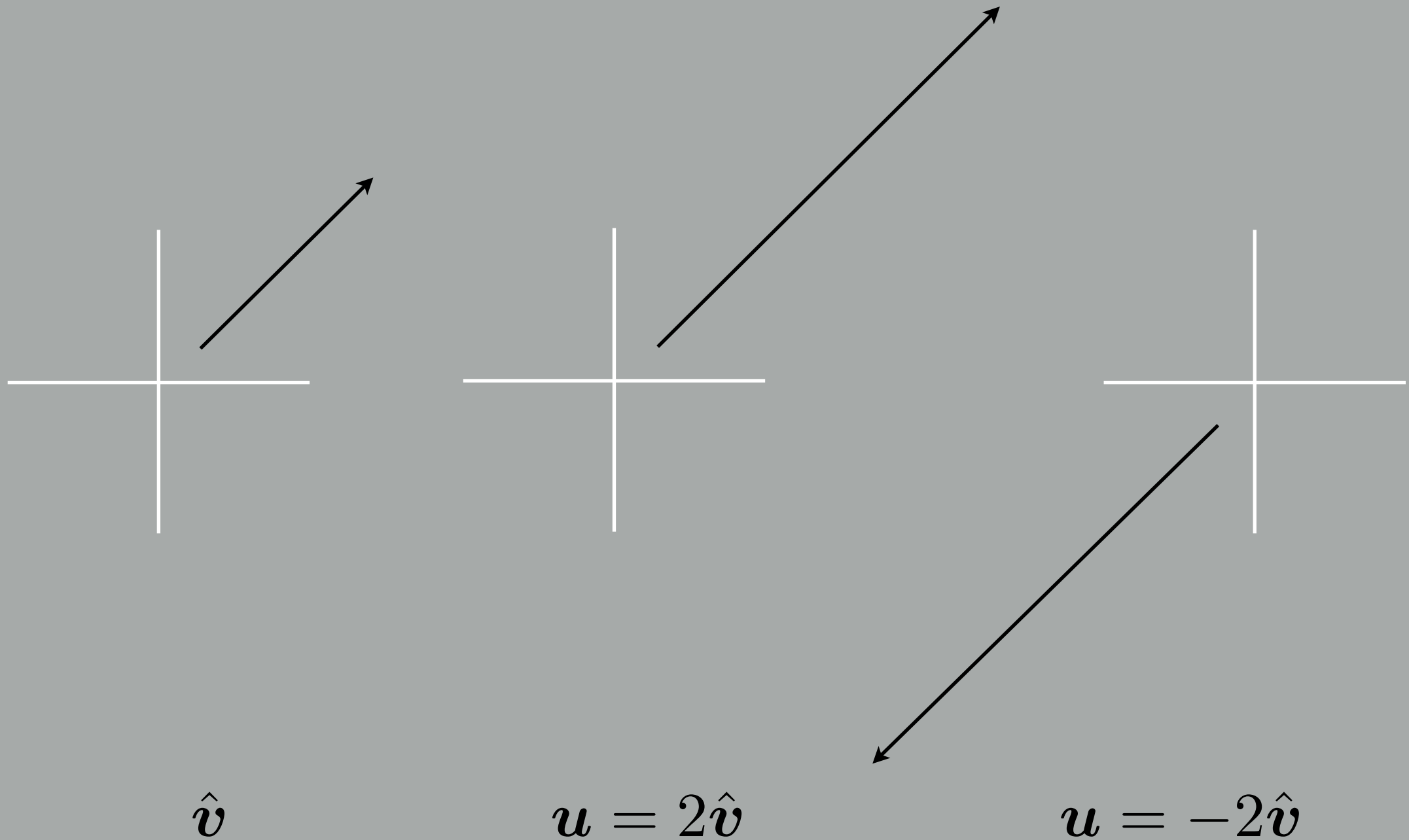


$$\|\hat{e}\| = 1$$

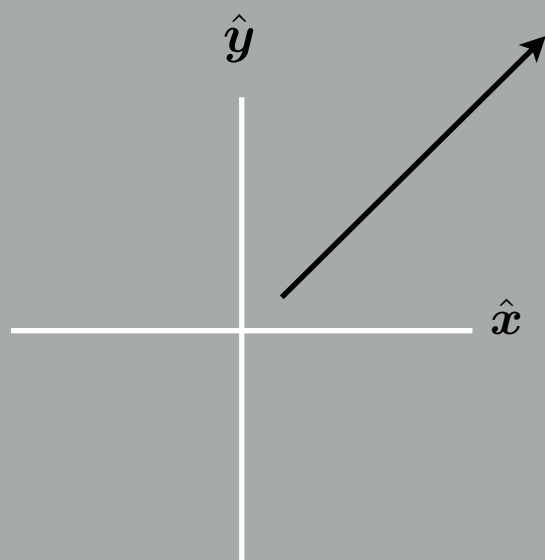


$$\|u\| = a\|\hat{e}\| = a$$

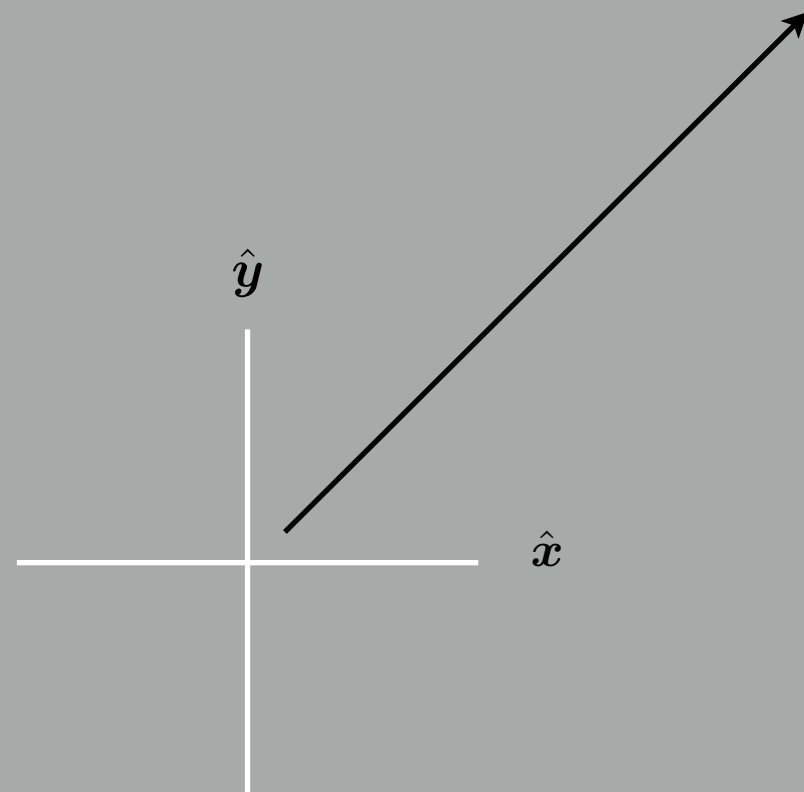
Vector-Scalar Multiplication



Example: Diffusion weighting



$\hat{g}$
↑
(the direction)

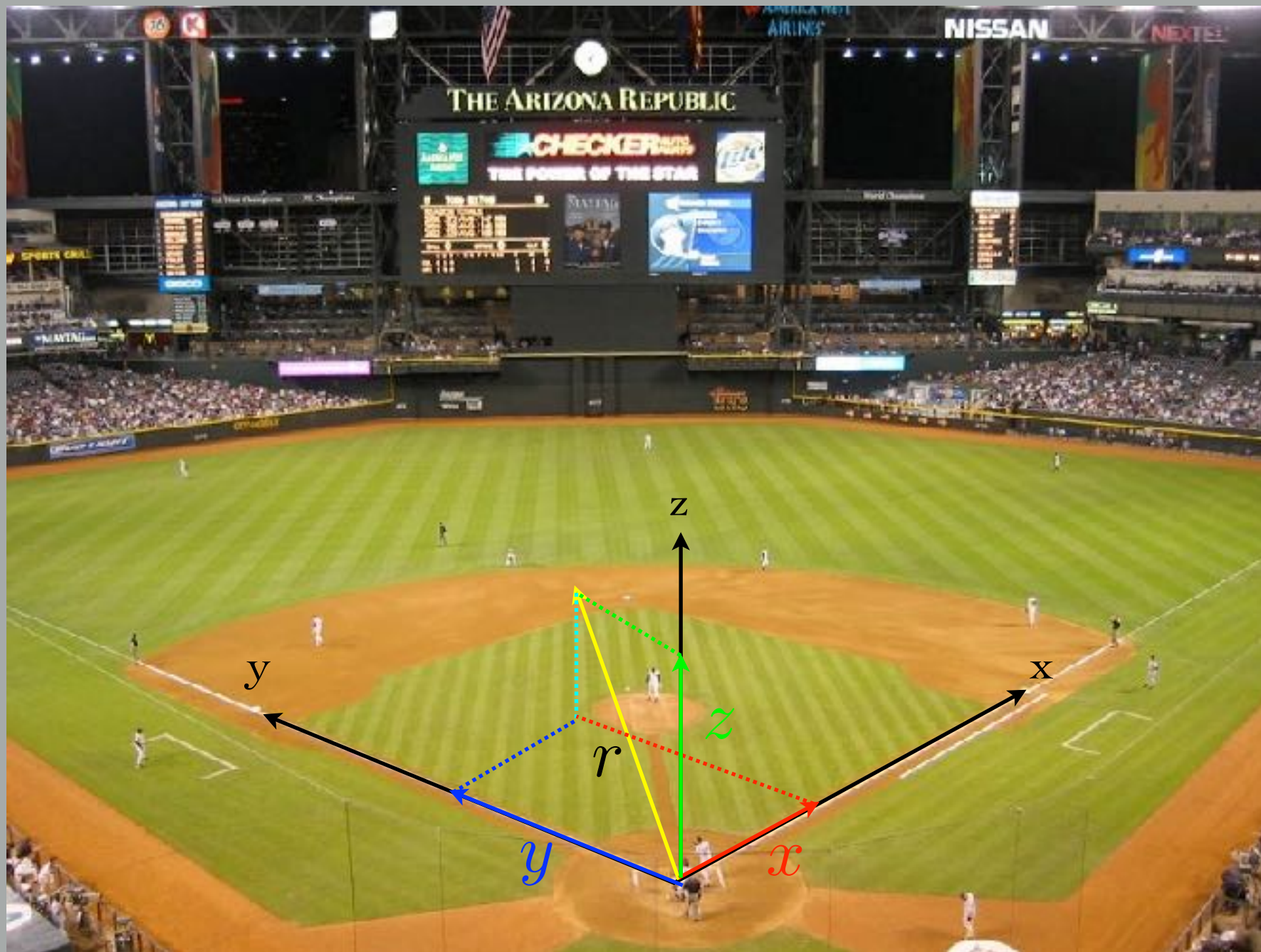


$g = b\hat{g}$
↑
(the weighting)

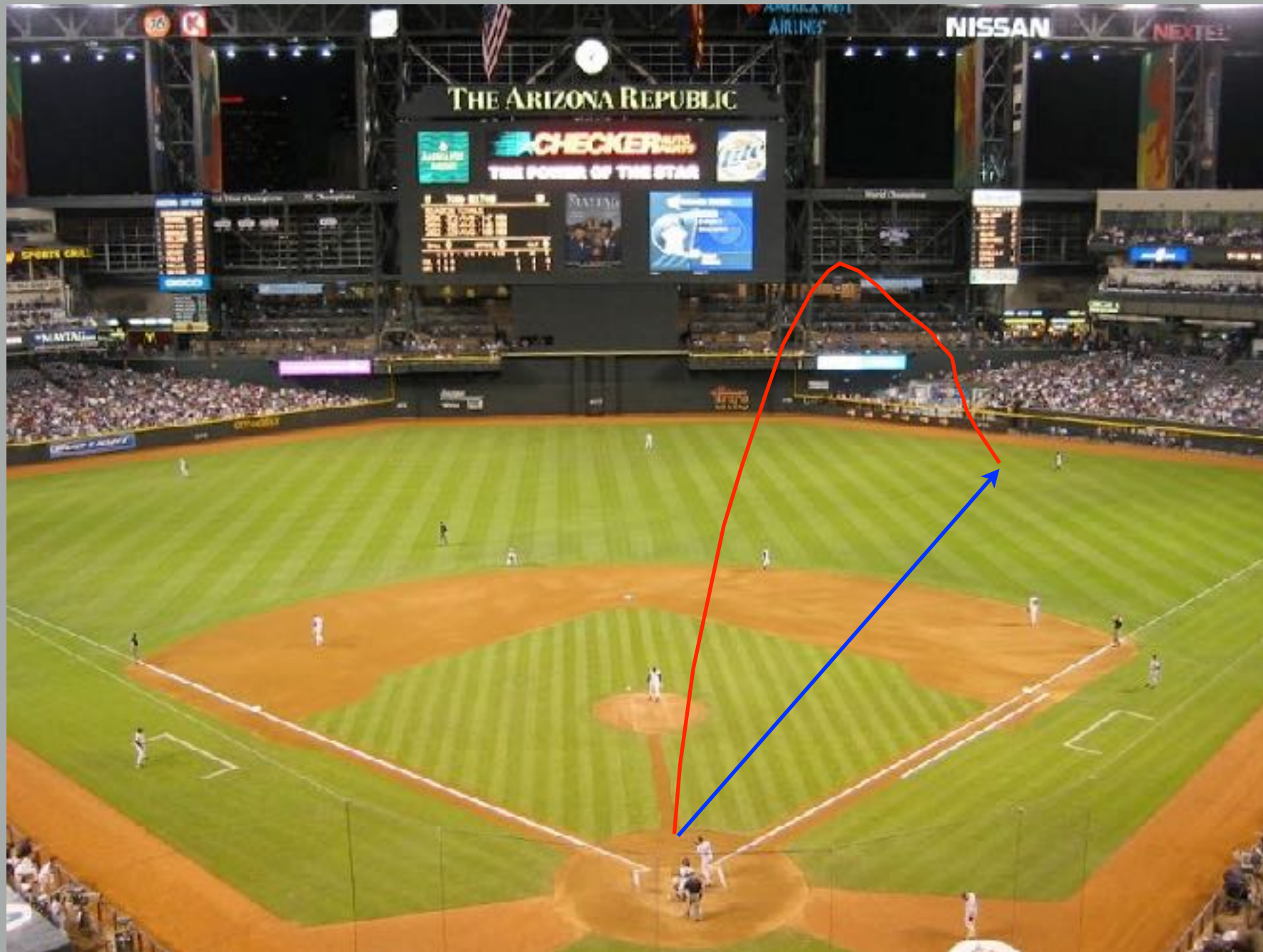
Projections



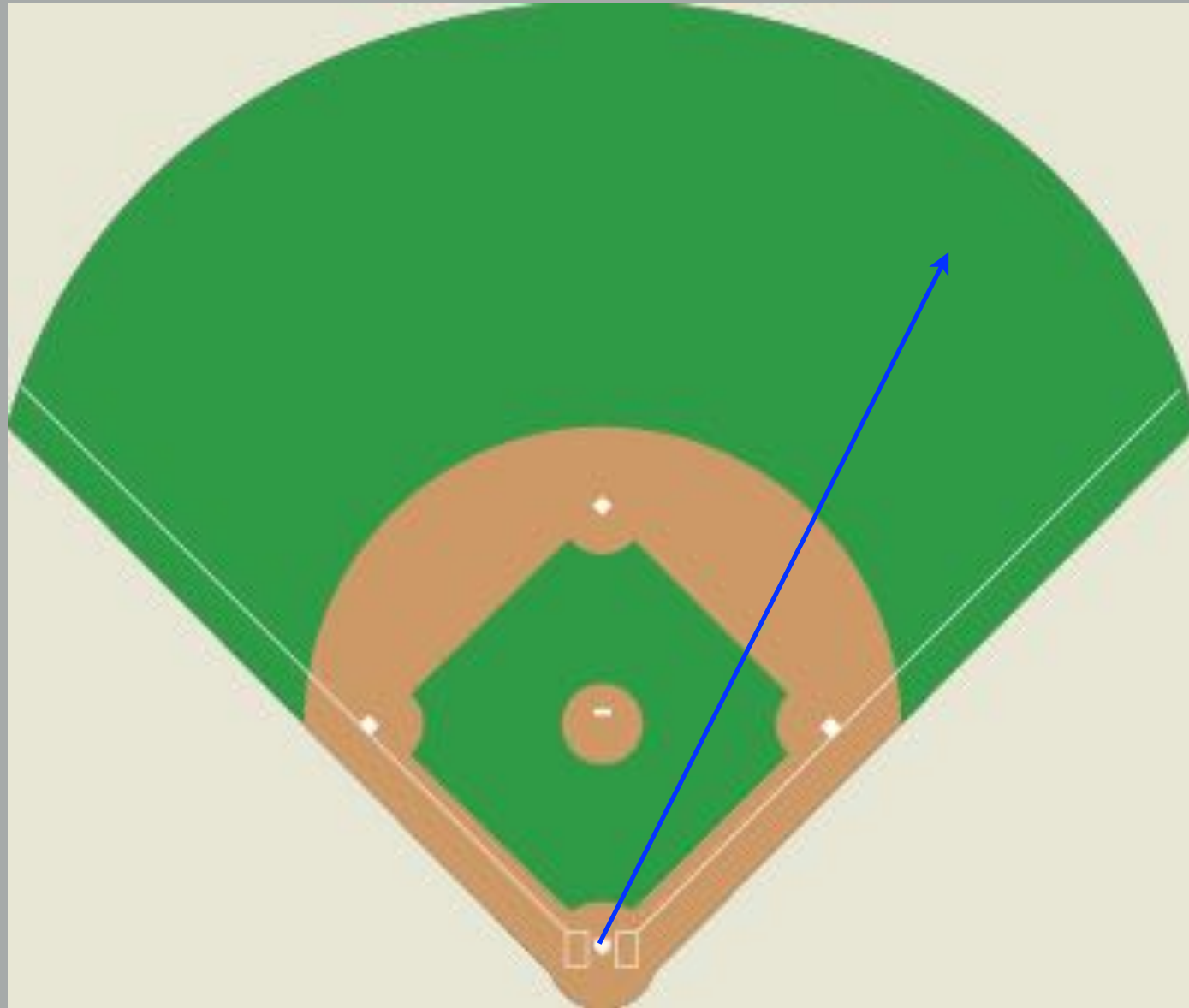
Projections



Projections

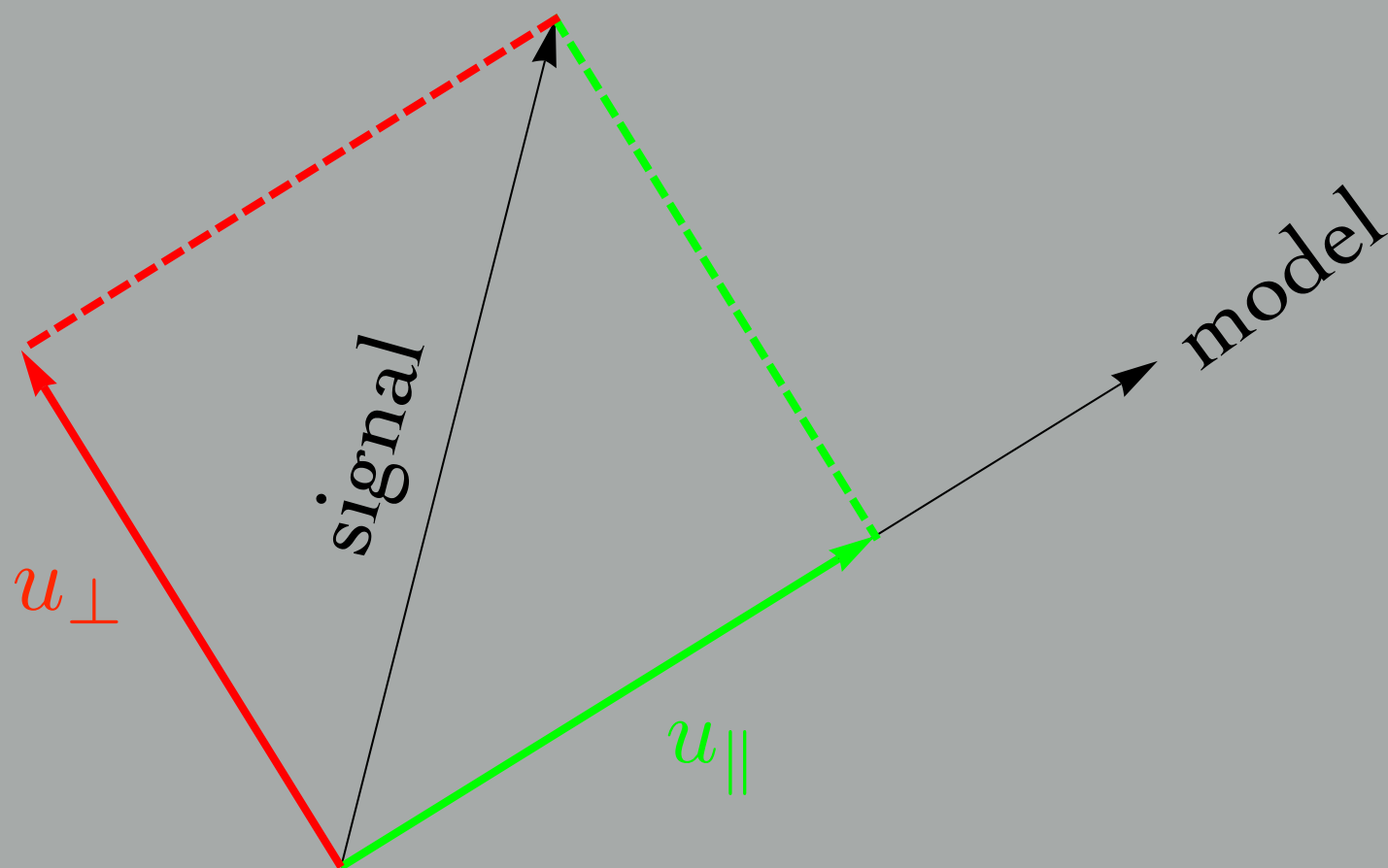


Projections



Projection

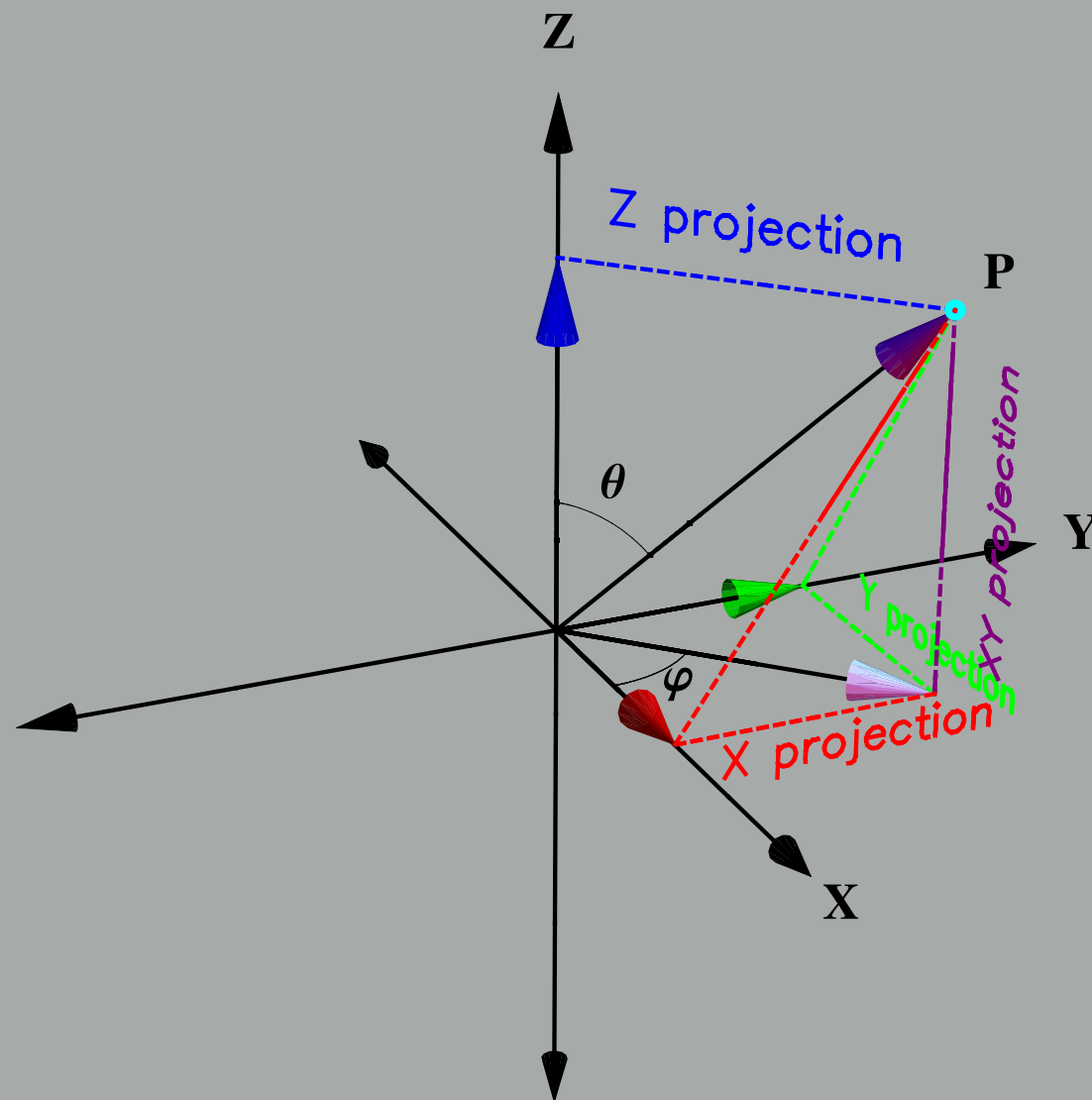
Example: Least squares fitting



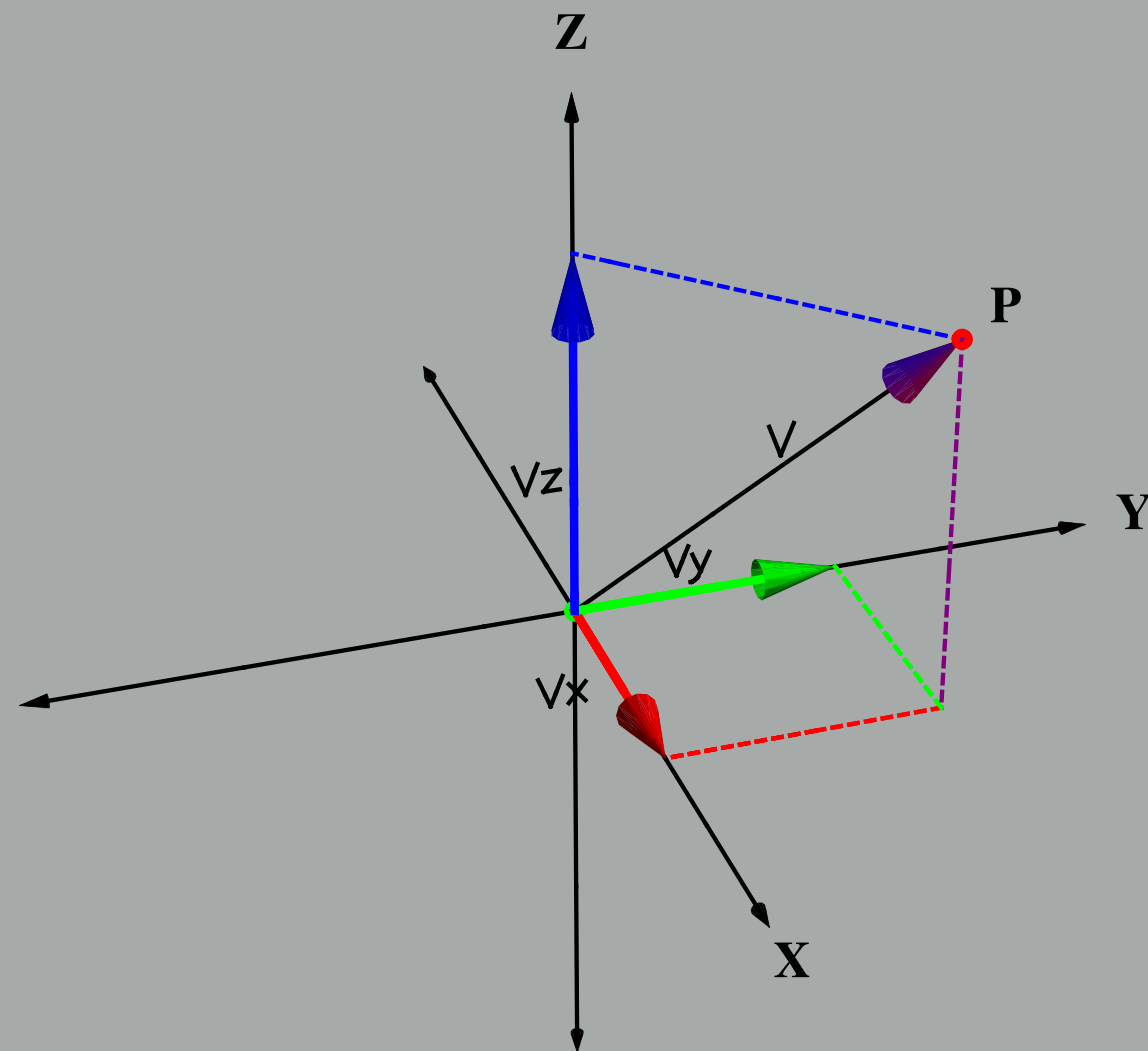
$u_{\parallel}$ = projection of signal onto model

$u_{\perp}$ = noise

Projections



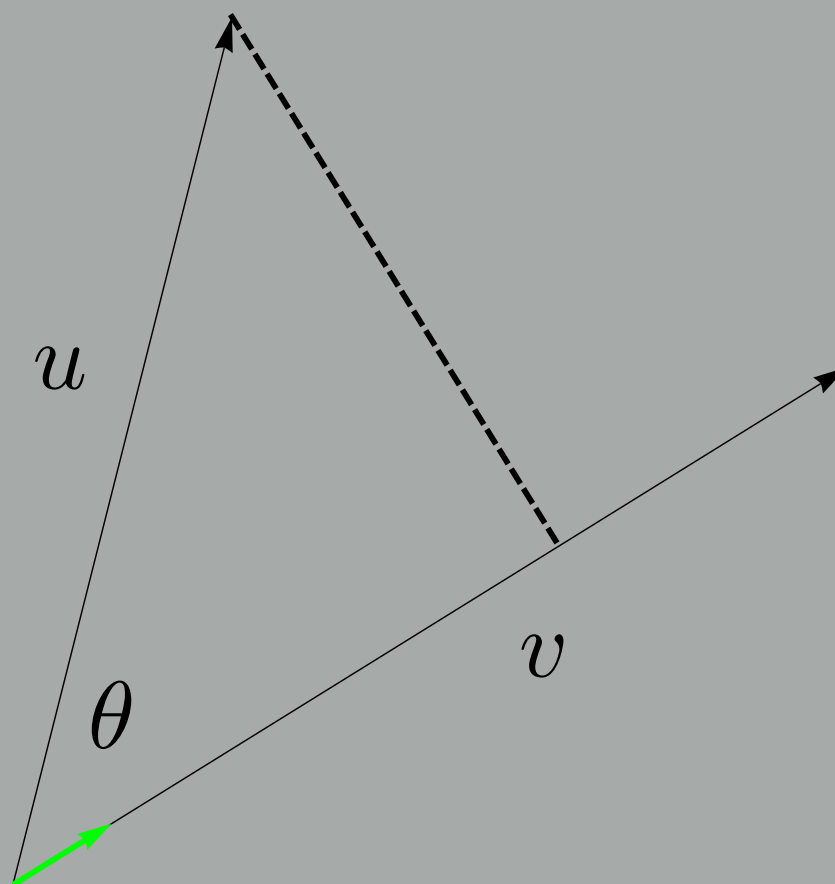
Vector components



In an *orthogonal* coordinate system, the vector length is the square root of the sum of the squares of the projections along the different axes

$$\|\mathbf{v}\| = \sqrt{v_x^2 + v_y^2 + v_z^2} = \sqrt{\mathbf{v}^t \mathbf{v}}$$

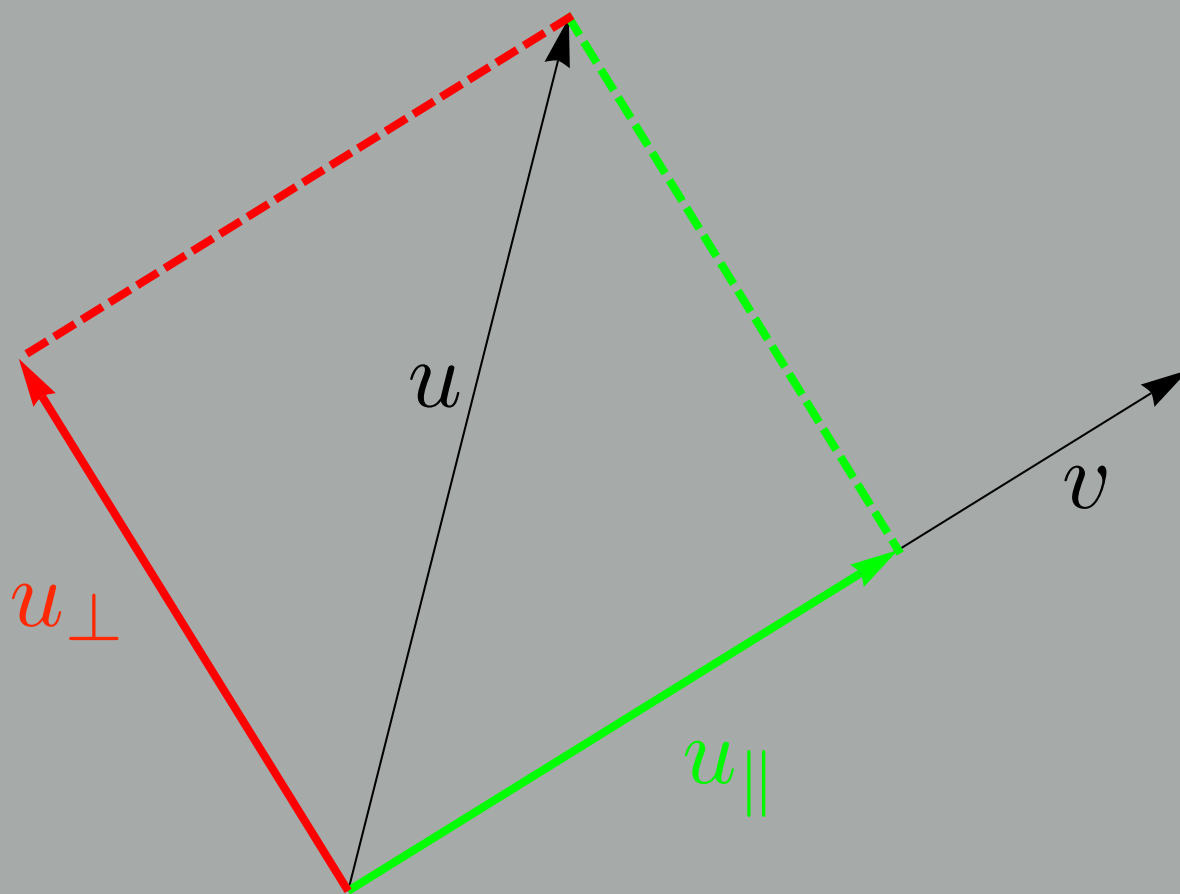
The Dot Product



$$\mathbf{u} \cdot \mathbf{v} = \|\mathbf{u}\| \|\mathbf{v}\| \cos \theta$$

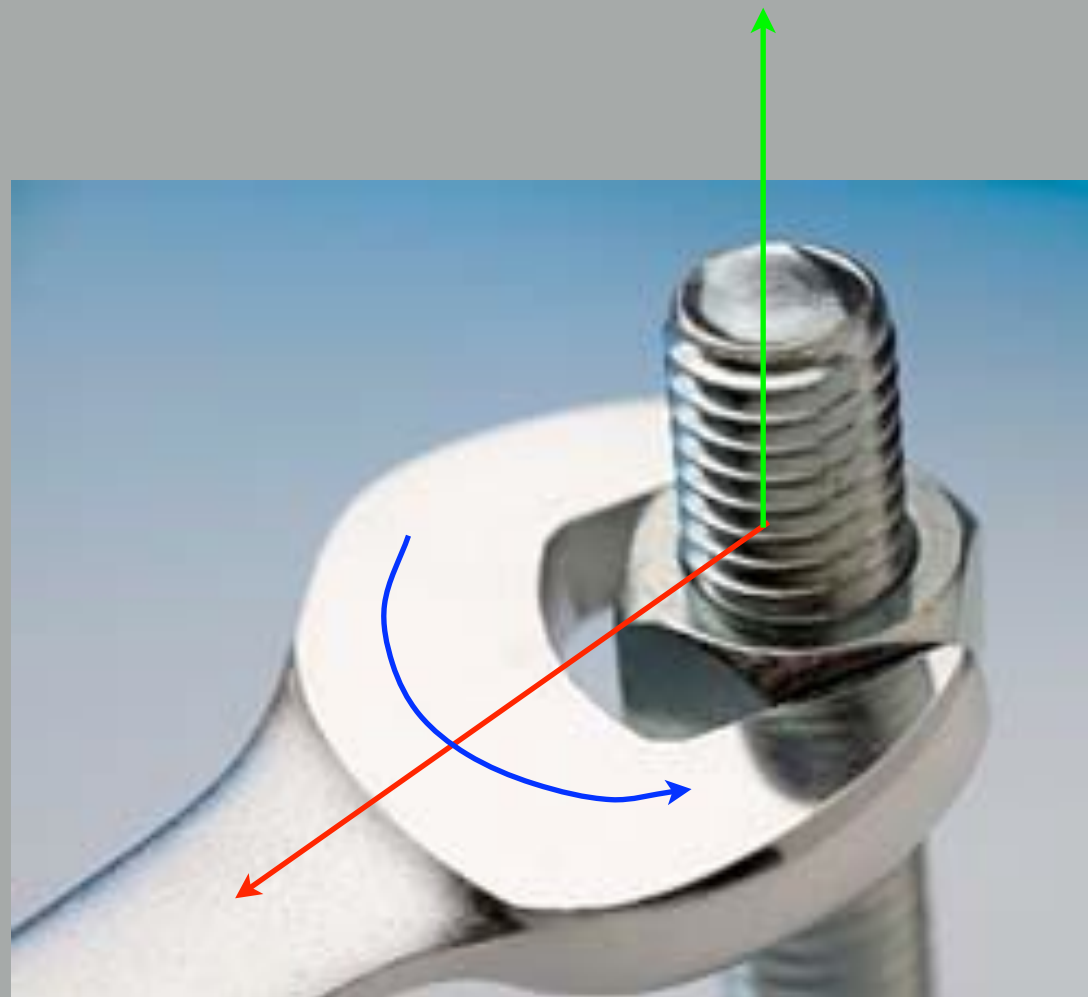
Projection

orthogonal complement: $u_{\perp} = u - u_{\parallel}$



projection of u onto v : $u_{\parallel} = \left(\frac{u \cdot v}{v \cdot v} \right) v$

The Cross Product



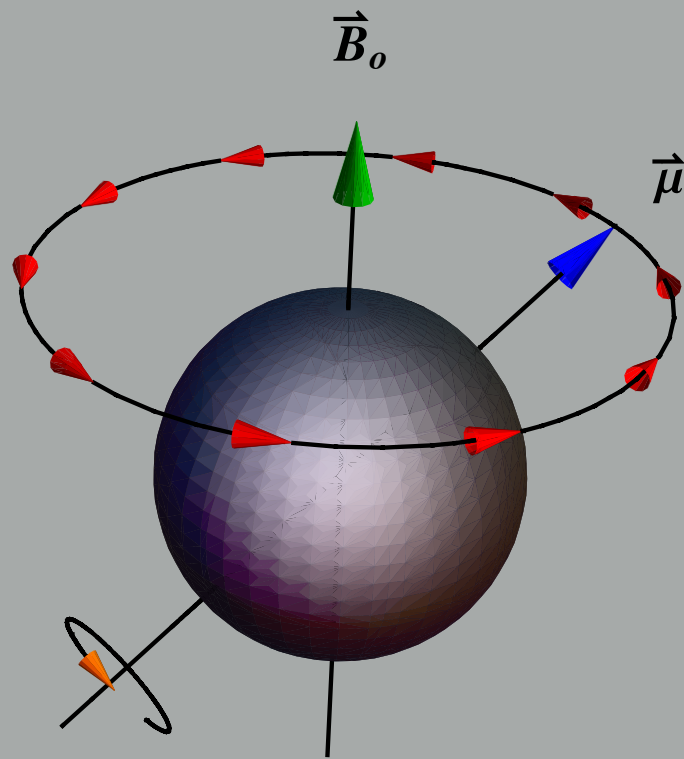
torque: $\tau = r \times F$

The Cross Product



angular momentum: $L = r \times p$

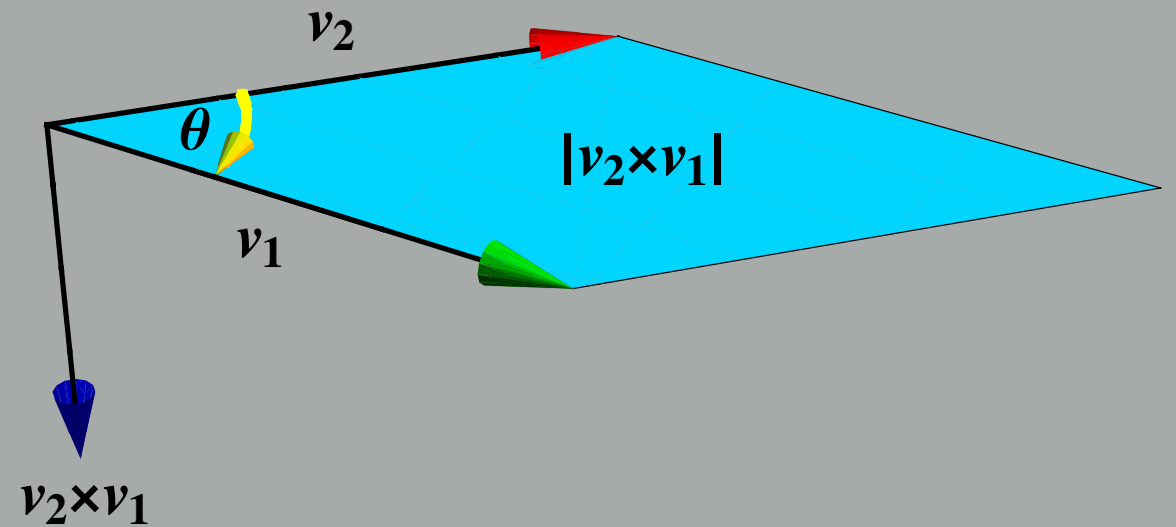
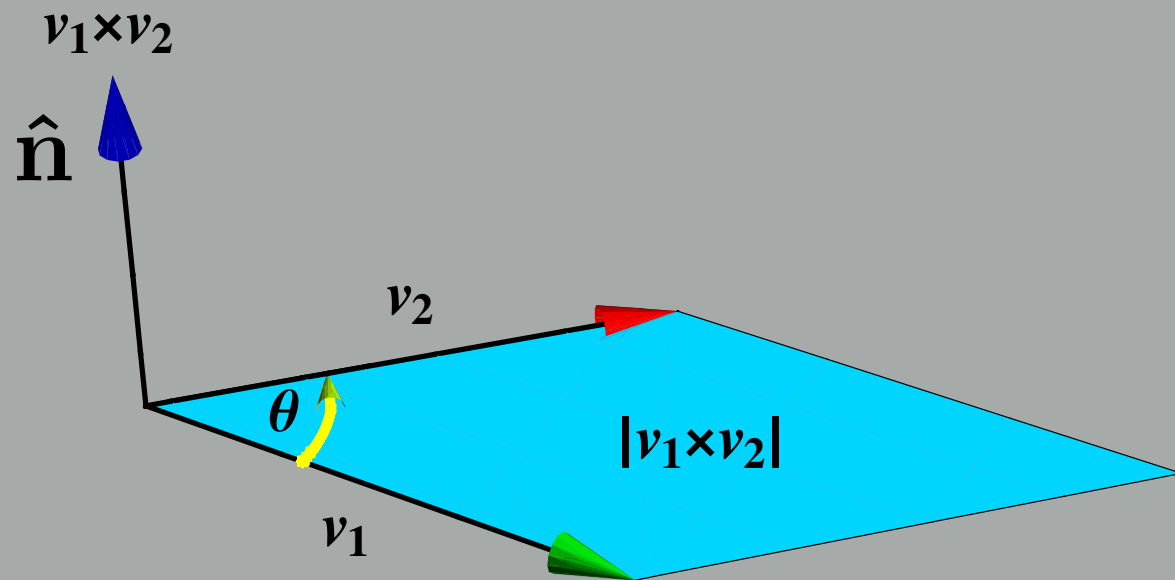
The Cross Product



nuclear precession

torque: $\tau = \mu \times B_0$

The Cross Product



$$\mathbf{v}_1 \times \mathbf{v}_2 = |\mathbf{v}_1| |\mathbf{v}_2| \sin \theta \mathbf{\hat{n}}$$

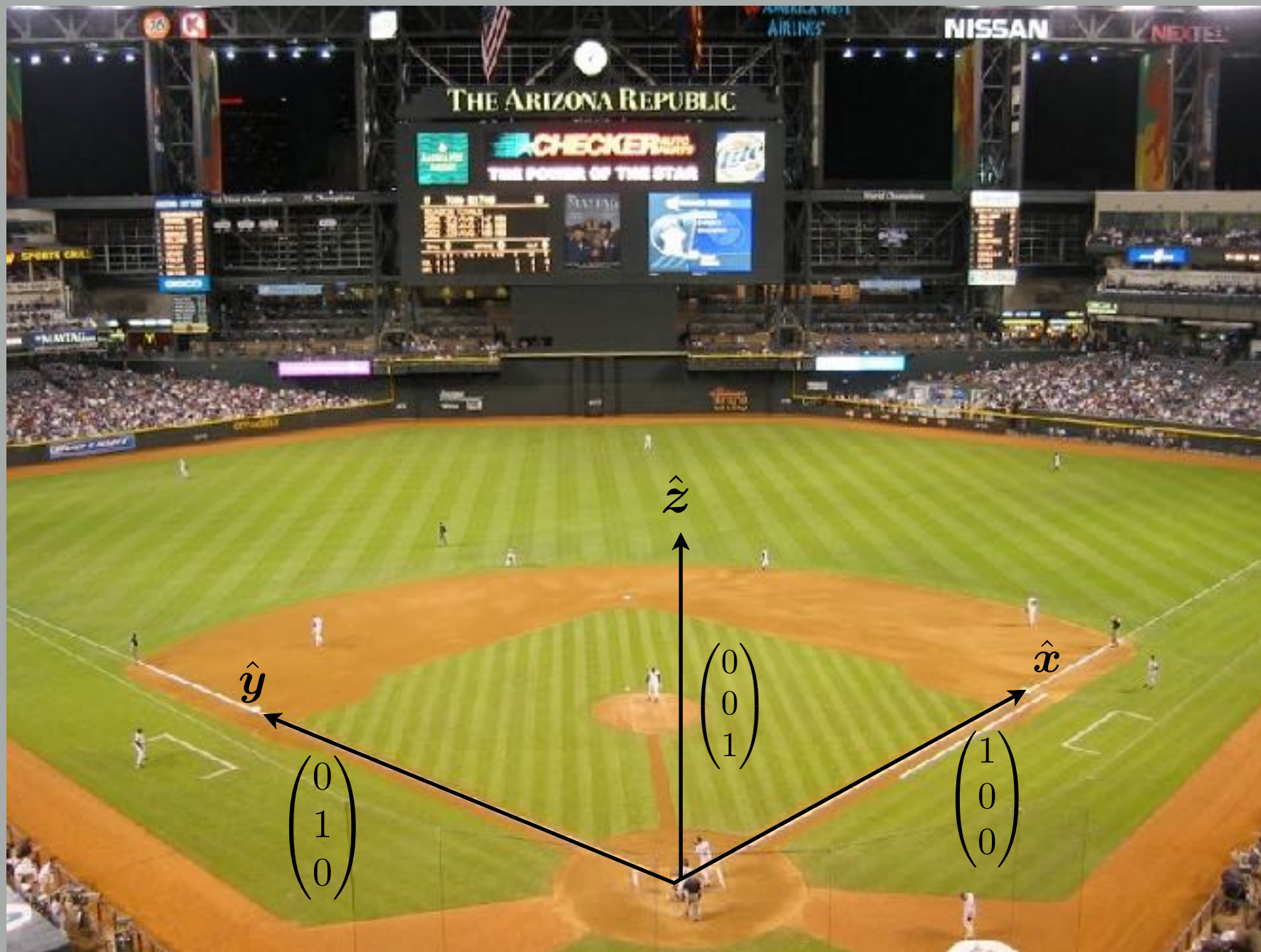
Matrices

Matrices as collections of vectors



Coordinate, in units of the distance to bases (90')

Coordinate system

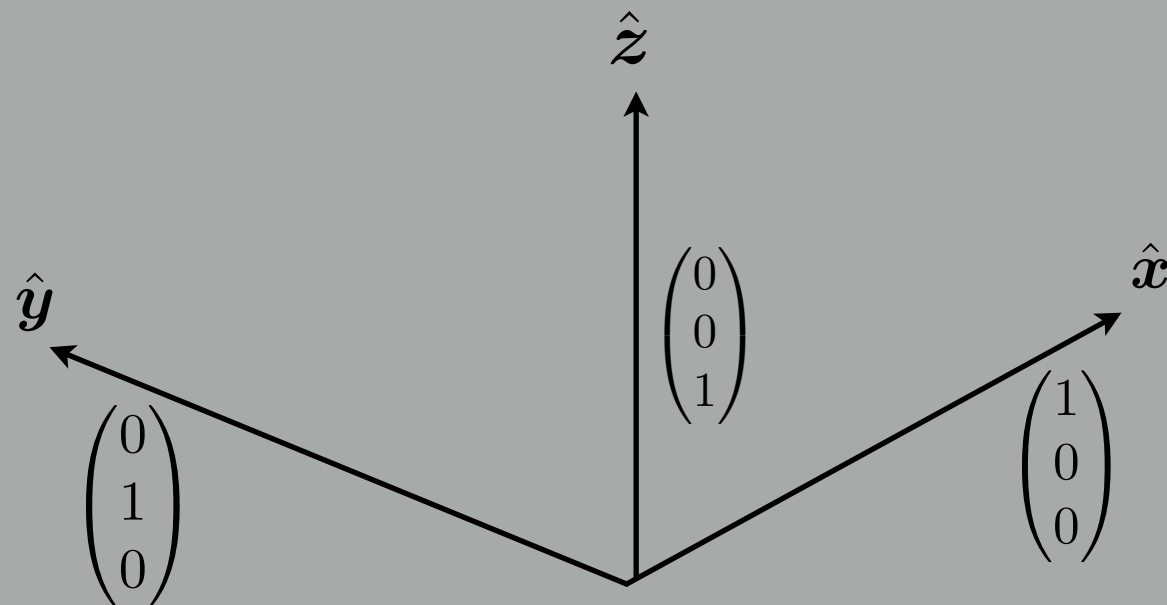


$$\begin{pmatrix} x \\ y \\ z \end{pmatrix}$$

Coordinate, in units of the distance to bases (90')

Coordinate system

$$(\hat{x}, \hat{y}, \hat{z}) = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \leftarrow \text{“diagonal matrix”}$$



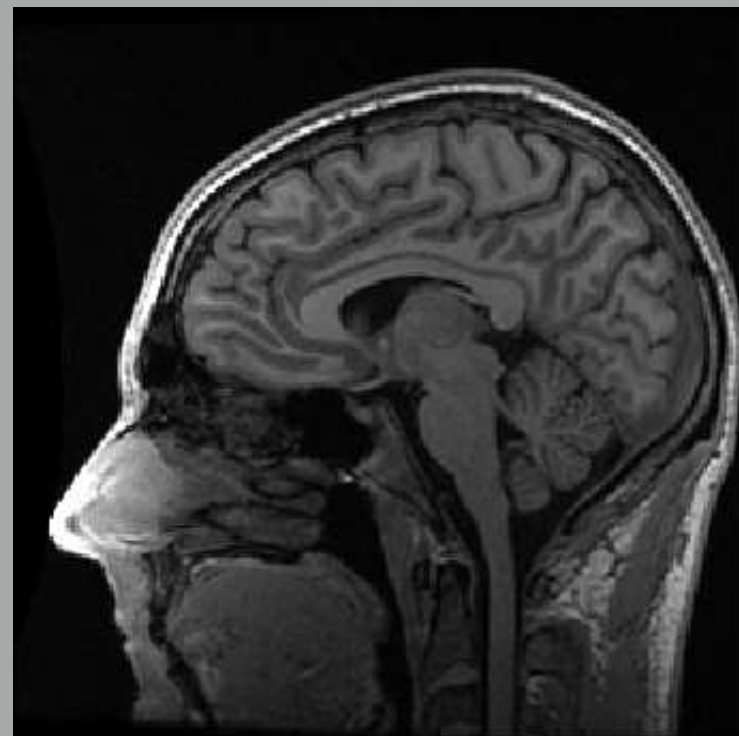
Two-dimensional matrix



intensities as a function of x and y

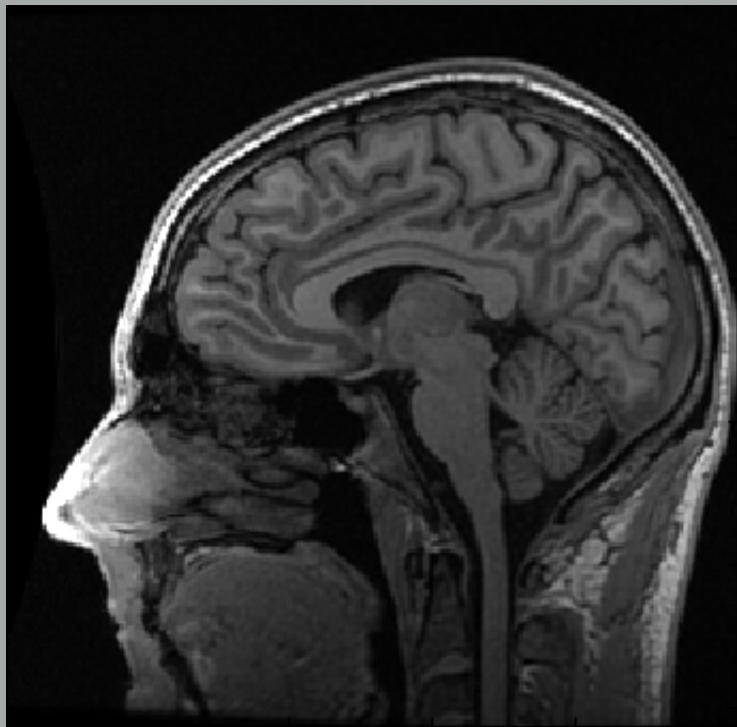
Matrix Notation

$$\mathbf{A} = \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & \ddots & & \\ \vdots & & & \\ a_{n1} & a_{n2} & \dots & a_{nn} \end{pmatrix} =$$



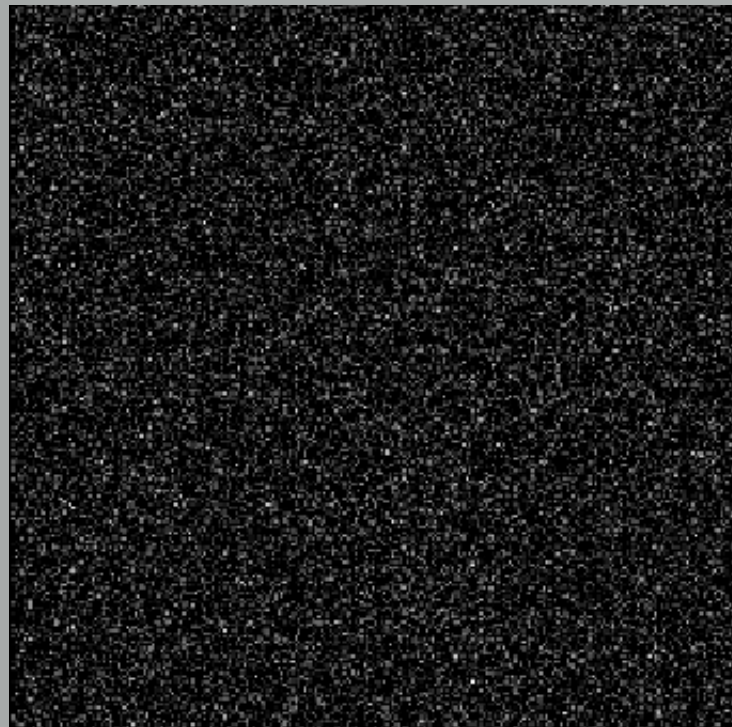
a_{ij} = intensity of voxel in row i and column j

Matrix addition



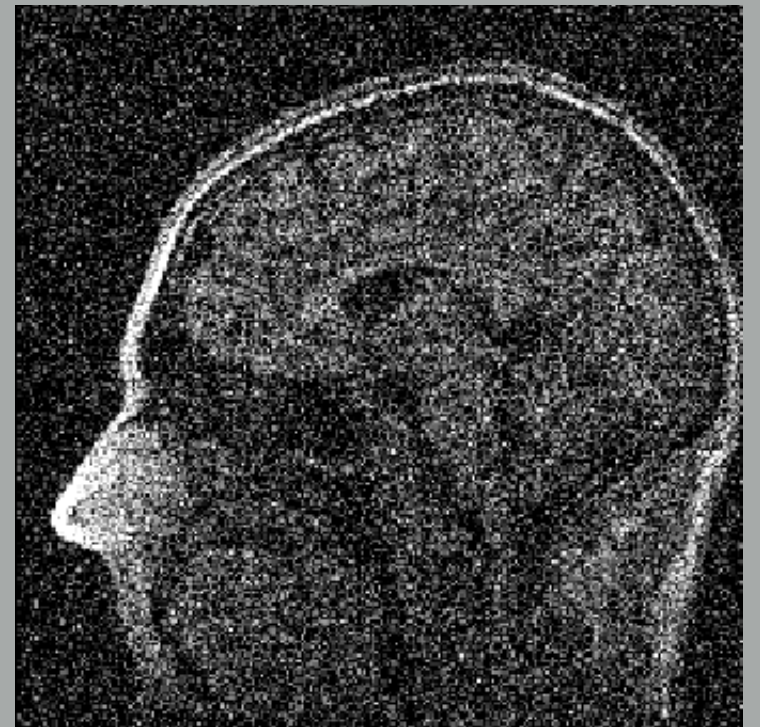
A

$$\begin{pmatrix} a & b \\ c & d \end{pmatrix}$$



B

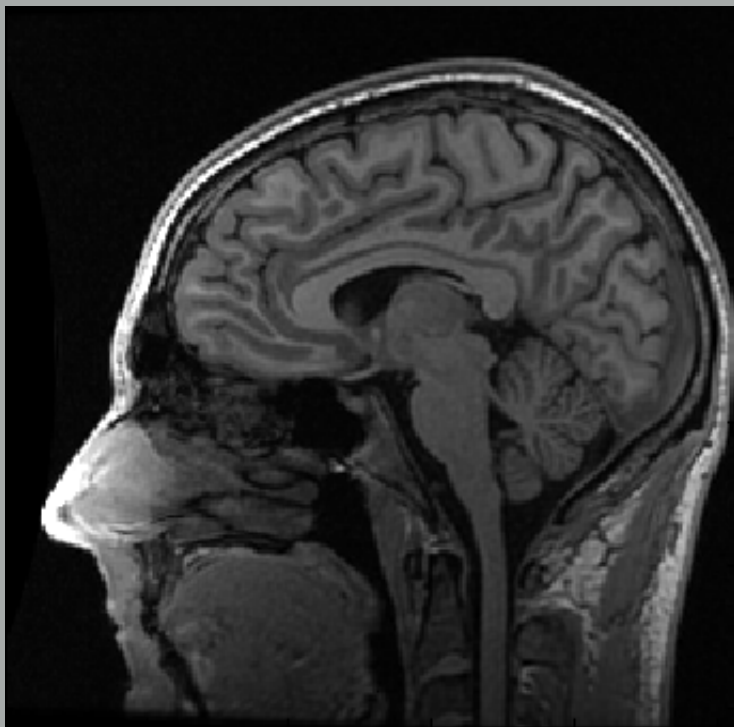
$$\begin{pmatrix} \alpha & \beta \\ \gamma & \delta \end{pmatrix}$$



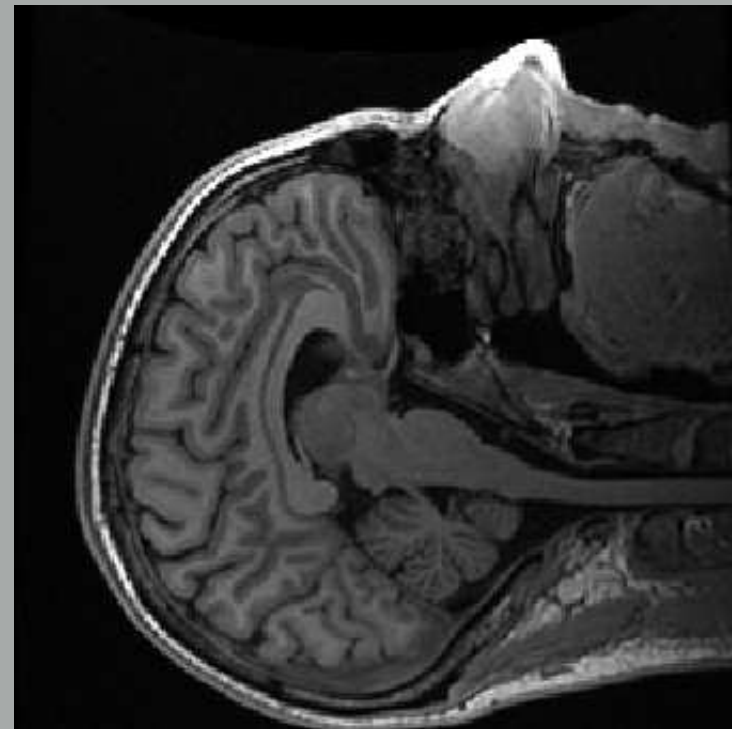
$A + B$

$$\begin{pmatrix} a + \alpha & b + \beta \\ c + \gamma & d + \delta \end{pmatrix}$$

Matrix transpose



A

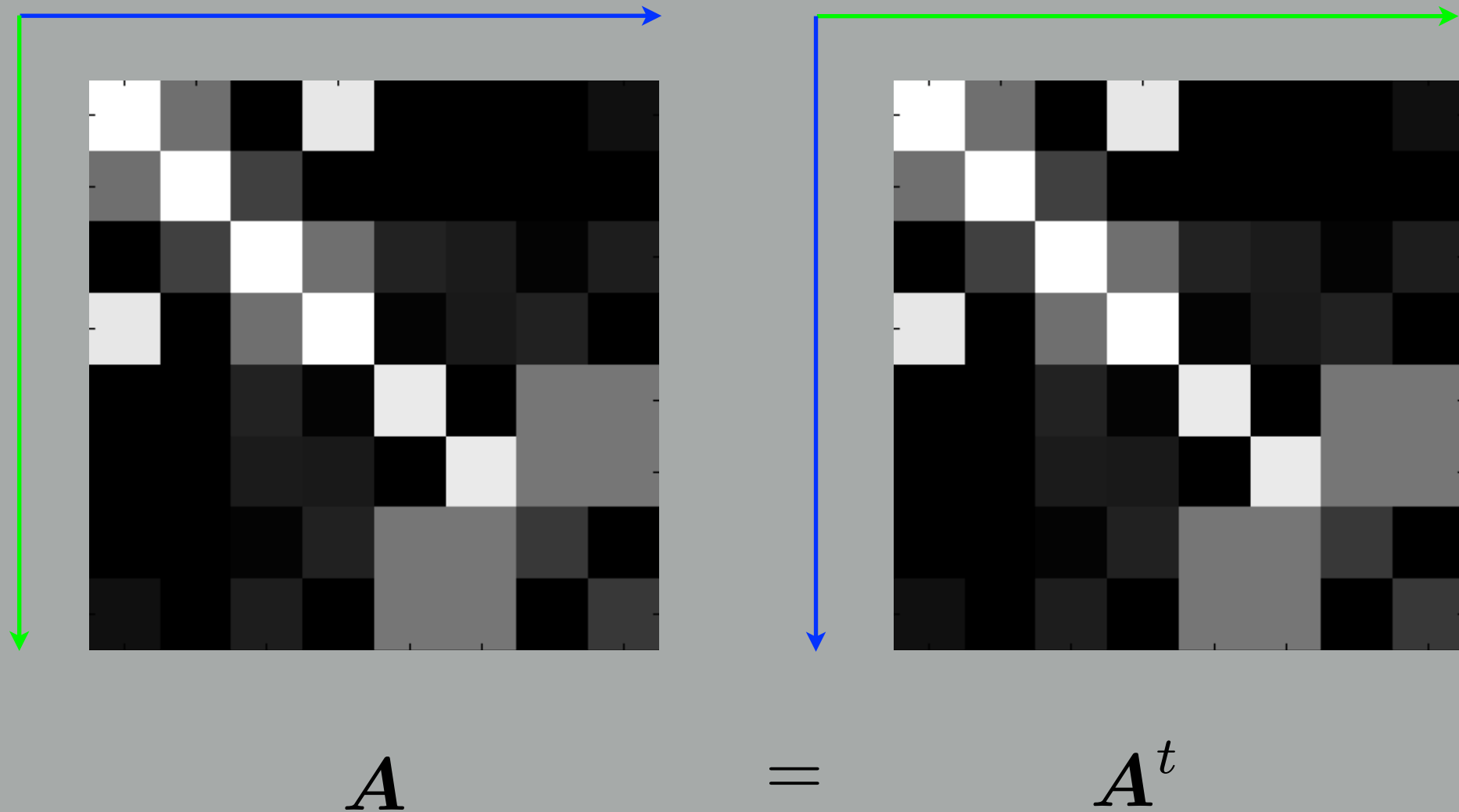


A^t

$$A = \begin{pmatrix} a & b & c \\ d & e & f \end{pmatrix}$$

$$A^t = \begin{pmatrix} a & d \\ b & e \\ c & f \end{pmatrix}$$

Symmetric Matrix



Example: Diffusion Tensor

$$\begin{pmatrix} D_{xx} & \boxed{D_{xy}} & \boxed{D_{xz}} \\ \boxed{D_{yx}} & D_{yy} & \boxed{D_{yz}} \\ \boxed{D_{zx}} & \boxed{D_{zy}} & D_{zz} \end{pmatrix} = \begin{pmatrix} D_{xx} & \boxed{D_{xy}} & \boxed{D_{xz}} \\ \boxed{D_{xy}} & D_{yy} & \boxed{D_{yz}} \\ \boxed{D_{xz}} & \boxed{D_{yz}} & D_{zz} \end{pmatrix}$$

$$D = D^t$$

Matrix Multiplication

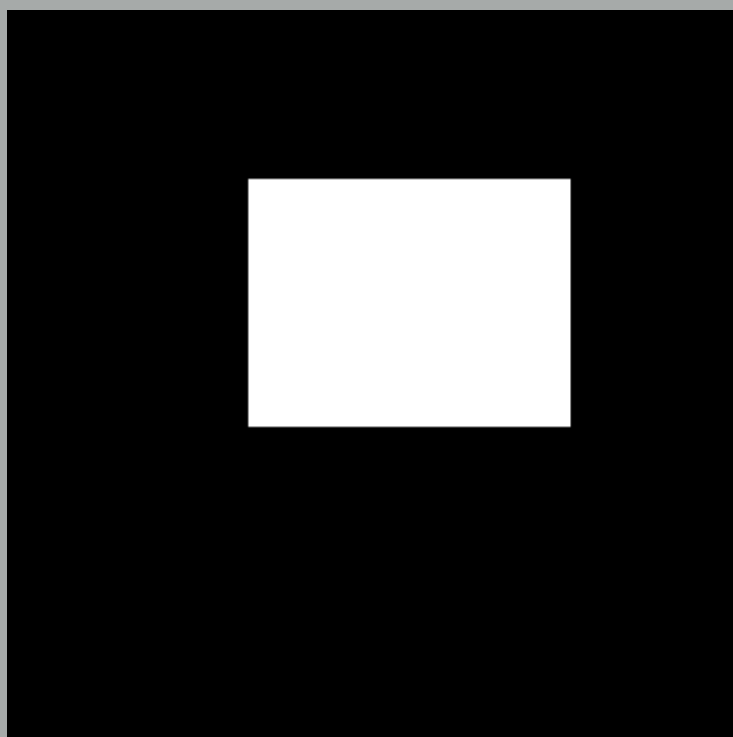
Hadamard product

Kronecker product

Dot product

Matrix Multiplication

Hadamard product



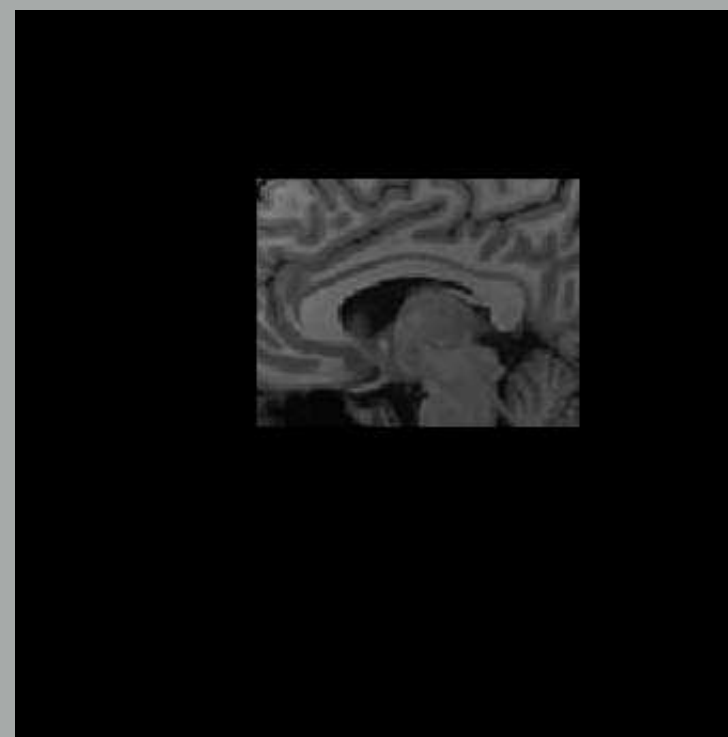
A

$$\begin{pmatrix} a & b \\ c & d \end{pmatrix}$$



B

$$\begin{pmatrix} \alpha & \beta \\ \gamma & \delta \end{pmatrix}$$

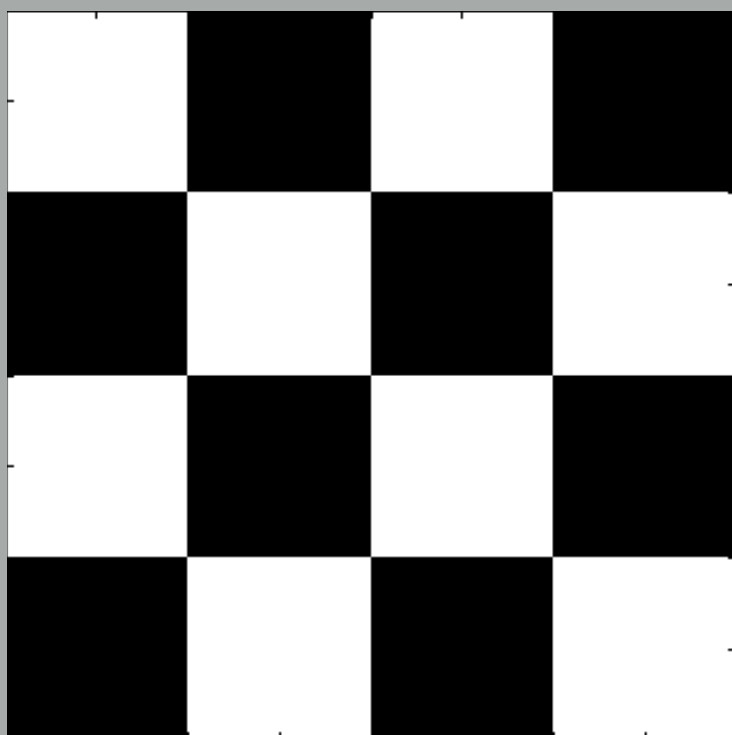


AB

$$\begin{pmatrix} \alpha a & \beta b \\ \gamma c & \delta d \end{pmatrix}$$

Matrix Multiplication

Kroneker product



A

$$\begin{pmatrix} a & b \\ c & d \end{pmatrix}$$



B

$$\begin{pmatrix} \alpha & \beta \\ \gamma & \delta \end{pmatrix}$$



$A \otimes B$

$$\begin{pmatrix} aB & bB \\ cB & dB \end{pmatrix}$$

Vector dot product

linear equation

$$f = ax + by + cz$$

Write in vector form:

$$\mathbf{u} = \begin{pmatrix} a \\ b \\ c \end{pmatrix} \quad \mathbf{v} = \begin{pmatrix} x \\ y \\ z \end{pmatrix}$$

$$f = \underbrace{(x \quad y \quad z)}_{\mathbf{u}^t} \underbrace{\begin{pmatrix} x \\ y \\ z \end{pmatrix}}_{\mathbf{v}} = \mathbf{u}^t \mathbf{v}$$

Vector dot product

linear equation

$$f = ax + by + cz$$

Write in vector form:

$$\mathbf{u} = \begin{pmatrix} a \\ b \\ c \end{pmatrix} \quad \mathbf{v} = \begin{pmatrix} x \\ y \\ z \end{pmatrix}$$

$$f = \underbrace{(x \quad y \quad z)}_{\mathbf{u}^t} \underbrace{\begin{pmatrix} x \\ y \\ z \end{pmatrix}}_{\mathbf{v}} = \mathbf{u}^t \mathbf{v}$$

Vector dot product

linear equation

$$f = ax + by + cz$$

Write in vector form:

$$\mathbf{u} = \begin{pmatrix} a \\ b \\ c \end{pmatrix} \quad \mathbf{v} = \begin{pmatrix} x \\ y \\ z \end{pmatrix}$$

$$f = \underbrace{(x \quad y \quad z)}_{\mathbf{u}^t} \underbrace{\begin{pmatrix} x \\ y \\ z \end{pmatrix}}_{\mathbf{v}} = \mathbf{u}^t \mathbf{v}$$

Vector dot product

linear equation

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$$f = \underbrace{(x \quad y \quad z)}_{\mathbf{u}^t} \underbrace{\begin{pmatrix} x \\ y \\ z \end{pmatrix}}_{\mathbf{v}} = \mathbf{u}^t \mathbf{v}$$

Matrices dot product

a system of linear equations

$$f_1 = a_1x + b_1y + c_1z$$

$$f_2 = a_2x + b_2y + c_2z$$

$$f_3 = a_3x + b_3y + c_3z$$

$$\begin{pmatrix} f_1 \\ f_2 \\ f_3 \end{pmatrix} = \begin{pmatrix} a_1 & b_1 & c_1 \\ a_2 & b_2 & c_2 \\ a_3 & b_3 & c_3 \end{pmatrix} \begin{pmatrix} x \\ y \\ z \end{pmatrix}$$

Matrices dot product

a system of linear equations

$$f_1 = a_1x + b_1y + c_1z$$

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Matrices dot product

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$$\begin{pmatrix} f_1 \\ f_2 \\ f_3 \end{pmatrix} = \begin{pmatrix} a_1 & b_1 & c_1 \\ a_2 & b_2 & c_2 \\ a_3 & b_3 & c_3 \end{pmatrix} \begin{pmatrix} x \\ y \\ z \end{pmatrix}$$

Identity Matrix

$$\mathbf{I} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

$$\mathbf{I}\mathbf{x} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \mathbf{x}$$

Matrix dot product

What about

$$\mathbf{AB} = \underbrace{\begin{pmatrix} a_1 & b_1 & c_1 \\ a_2 & b_2 & c_2 \\ a_3 & b_3 & c_3 \end{pmatrix}}_A \underbrace{\begin{pmatrix} x_1 & x_2 \\ y_1 & y_2 \\ z_1 & z_3 \end{pmatrix}}_B$$

Matrices dot product

$$\begin{pmatrix} ? \end{pmatrix} = \begin{pmatrix} a_1 & b_1 & c_1 \\ a_2 & b_2 & c_2 \\ a_3 & b_3 & c_3 \end{pmatrix} \begin{pmatrix} x_1 & x_2 \\ y_1 & y_2 \\ z_1 & z_3 \end{pmatrix}$$

Matrices dot product

$$\begin{pmatrix} f_{11} \\ \phantom{f_{11}} \\ \phantom{f_{11}} \end{pmatrix} = \begin{pmatrix} a_1 & b_1 & c_1 \\ a_2 & b_2 & c_2 \\ a_3 & b_3 & c_3 \end{pmatrix} \begin{pmatrix} x_1 & x_2 \\ y_1 & y_2 \\ z_1 & z_3 \end{pmatrix}$$

Matrices dot product

$$\begin{pmatrix} f_{11} & f_{12} \end{pmatrix} = \begin{pmatrix} a_1 & b_1 & c_1 \\ a_2 & b_2 & c_2 \\ a_3 & b_3 & c_3 \end{pmatrix} \begin{pmatrix} x_1 & x_2 \\ y_1 & y_2 \\ z_1 & z_3 \end{pmatrix}$$

Matrices dot product

$$\begin{pmatrix} f_{11} & f_{12} \\ f_{21} & \end{pmatrix} = \begin{pmatrix} a_1 & b_1 & c_1 \\ a_2 & b_2 & c_2 \\ a_3 & b_3 & c_3 \end{pmatrix} \begin{pmatrix} x_1 & x_2 \\ y_1 & y_2 \\ z_1 & z_3 \end{pmatrix}$$

Matrices dot product

$$\begin{pmatrix} f_{11} & f_{12} \\ f_{21} & f_{22} \end{pmatrix} = \begin{pmatrix} a_1 & b_1 & c_1 \\ a_2 & b_2 & c_2 \\ a_3 & b_3 & c_3 \end{pmatrix} \begin{pmatrix} x_1 & x_2 \\ y_1 & y_2 \\ z_1 & z_3 \end{pmatrix}$$

Matrices dot product

$$\begin{pmatrix} f_{11} & f_{12} \\ f_{21} & f_{22} \\ f_{31} & \end{pmatrix} = \begin{pmatrix} a_1 & b_1 & c_1 \\ a_2 & b_2 & c_2 \\ a_3 & b_3 & c_3 \end{pmatrix} \begin{pmatrix} x_1 & x_2 \\ y_1 & y_2 \\ z_1 & z_3 \end{pmatrix}$$

Matrices dot product

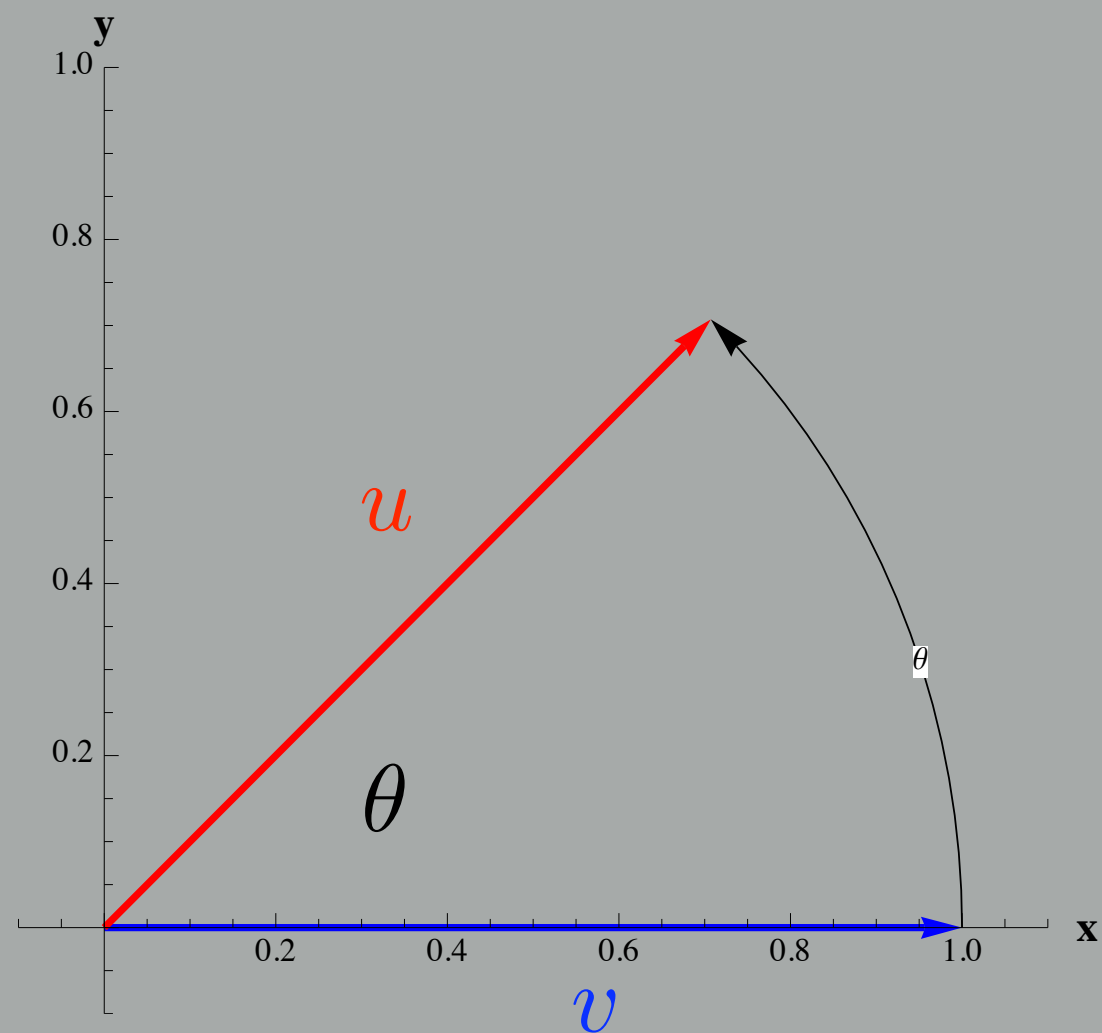
$$\begin{pmatrix} f_{11} & f_{12} \\ f_{21} & f_{22} \\ f_{31} & f_{32} \end{pmatrix} = \begin{pmatrix} a_1 & b_1 & c_1 \\ a_2 & b_2 & c_2 \\ a_3 & b_3 & c_3 \end{pmatrix} \begin{pmatrix} x_1 & x_2 \\ y_1 & y_2 \\ z_1 & z_3 \end{pmatrix}$$

Matrices dot product

dimensions = [rows $\times$ cols]

$$\begin{array}{ccc} \begin{pmatrix} f_{11} & f_{12} \\ f_{21} & f_{22} \\ f_{31} & f_{32} \end{pmatrix} & = & \begin{pmatrix} a_1 & b_1 & c_1 \\ a_2 & b_2 & c_2 \\ a_3 & b_3 & c_3 \end{pmatrix} \begin{pmatrix} x_1 & x_2 \\ y_1 & y_2 \\ z_1 & z_3 \end{pmatrix} \\ [n \times p] & & [n \times m] \qquad [m \times p] \end{array}$$

Example: Rotations

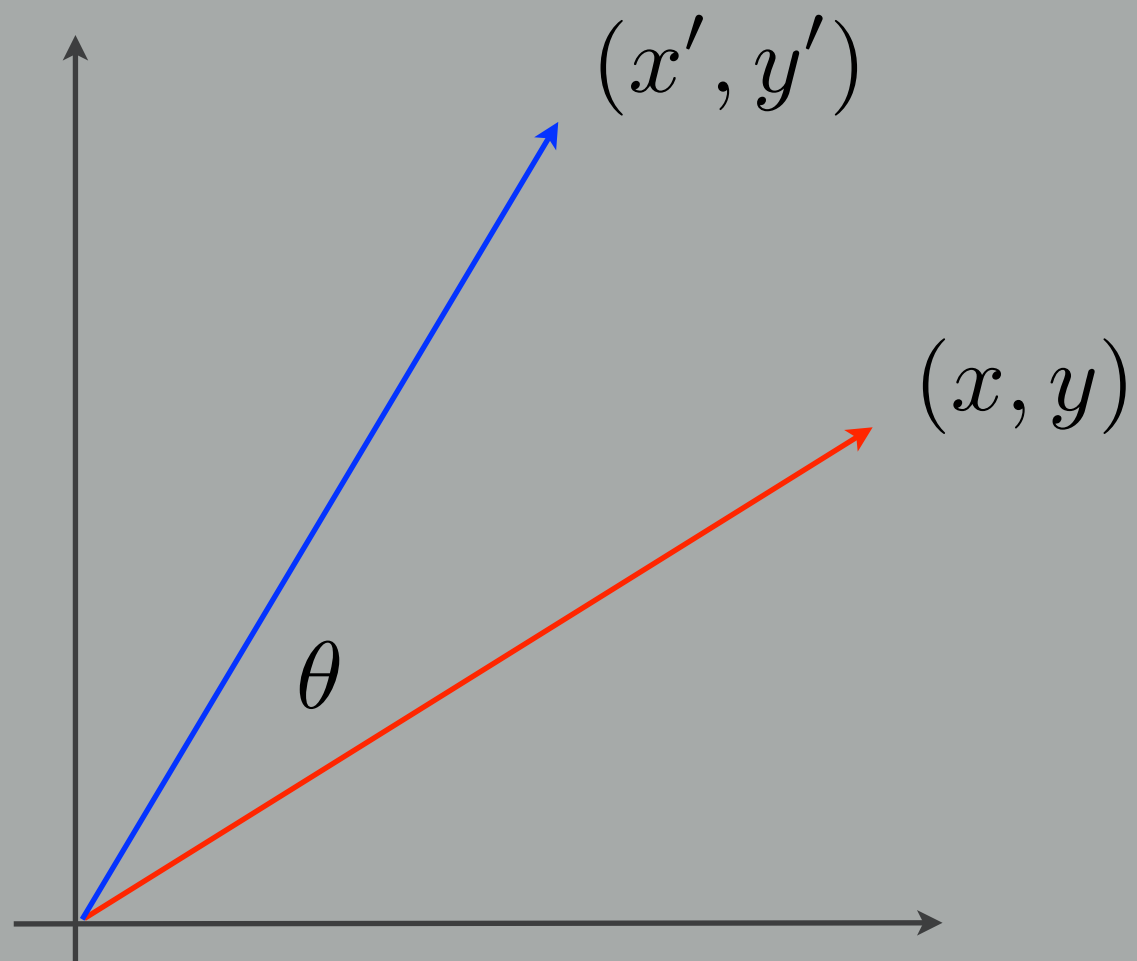


$$R(\uparrow) = \uparrow \quad u = Rv$$

Rotations

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

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

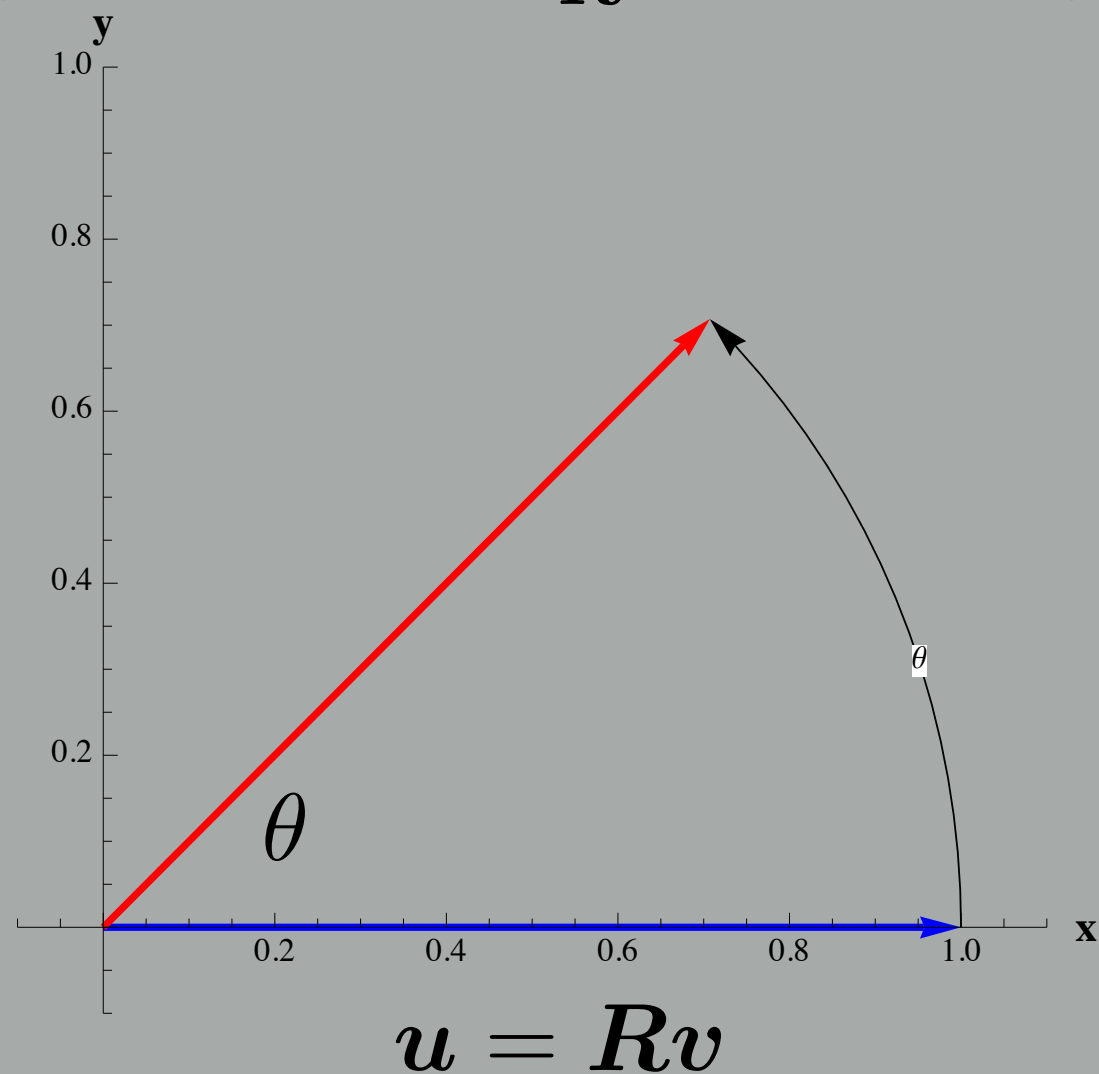


Rotations

$$\begin{pmatrix} x' \\ y' \end{pmatrix} = \begin{pmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix}$$

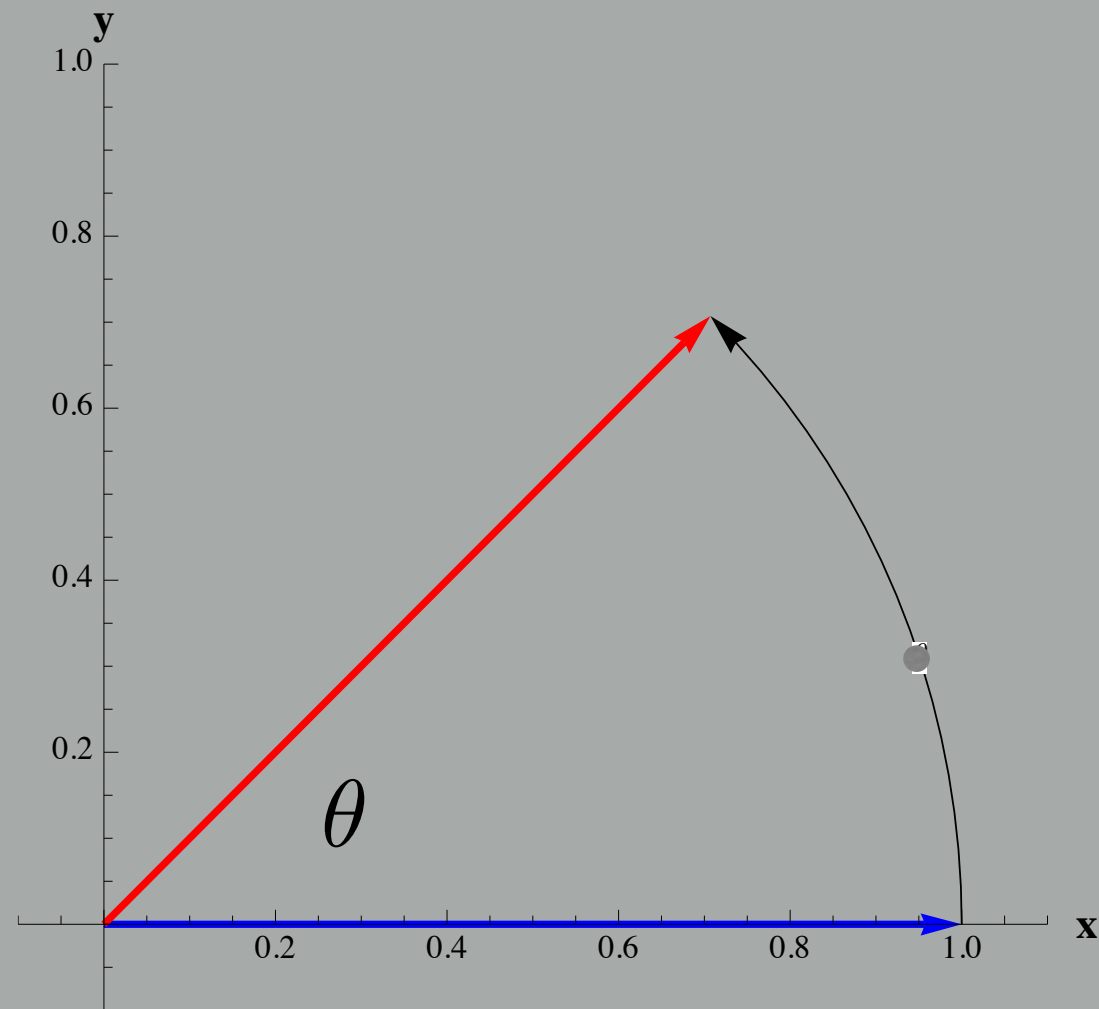
Rotations

$$\underbrace{\begin{pmatrix} u_x \\ u_y \end{pmatrix}}_{\mathbf{u}} = \underbrace{\begin{pmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{pmatrix}}_{\mathbf{R}} \underbrace{\begin{pmatrix} v_x \\ v_y \end{pmatrix}}_{\mathbf{v}}$$



2D Rotation Matrix

$$R = \begin{pmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{pmatrix}$$



$$R(\uparrow) = \uparrow$$

$$u = Rv$$

What about rotating an ellipse?

Equation of ellipse

$$Ax^2 + 2Bxy + Cy^2$$

In matrix form

$$\vec{x}^t \vec{Q} \vec{x}$$

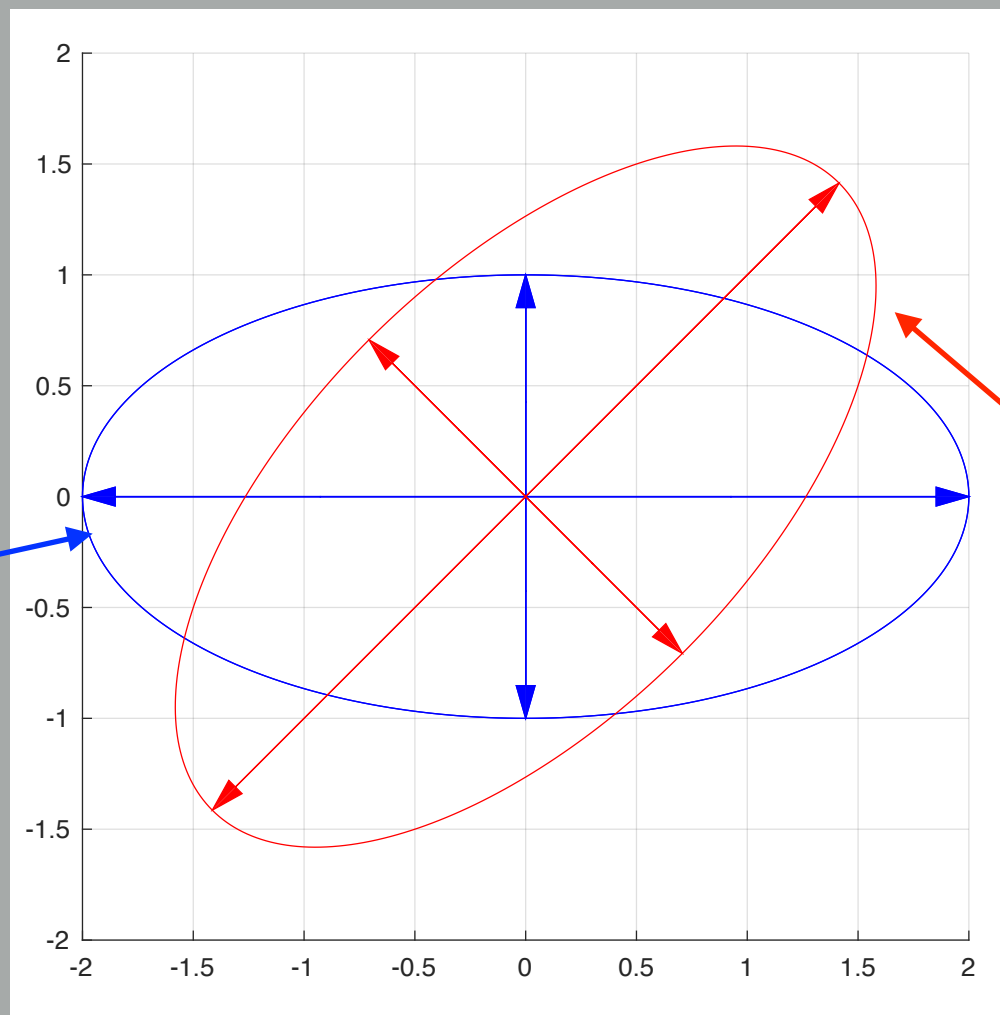
where

$$\vec{x} = \begin{pmatrix} x \\ y \end{pmatrix} \qquad \vec{Q} = \begin{pmatrix} A & B \\ B & C \end{pmatrix}$$

What about rotating an ellipse?

$$\vec{x} = \begin{pmatrix} x \\ y \end{pmatrix}$$

$$\vec{x}^t \vec{Q} \vec{x}$$



$$\vec{x}'^t = \vec{R} \vec{x}$$

$$\vec{x}'^t \vec{Q} \vec{x}'$$

What about rotating an ellipse?

rotated axes: $x' = Rx$ ellipse along rotated axes: $x'^t Q x'$

$$x'^t Q x' = (Rx)^t Q (Rx) = x^t R^t Q R x = x^t Q' x$$

where $Q' \equiv R^t Q R$

Similarity transform

rotation of ellipse is different than rotation of vector!

Complex number motivation

MR image



$$m_{\perp}(\boldsymbol{x}) = \int s(\boldsymbol{k}) e^{i\boldsymbol{k} \cdot \boldsymbol{x}} d\boldsymbol{k}$$

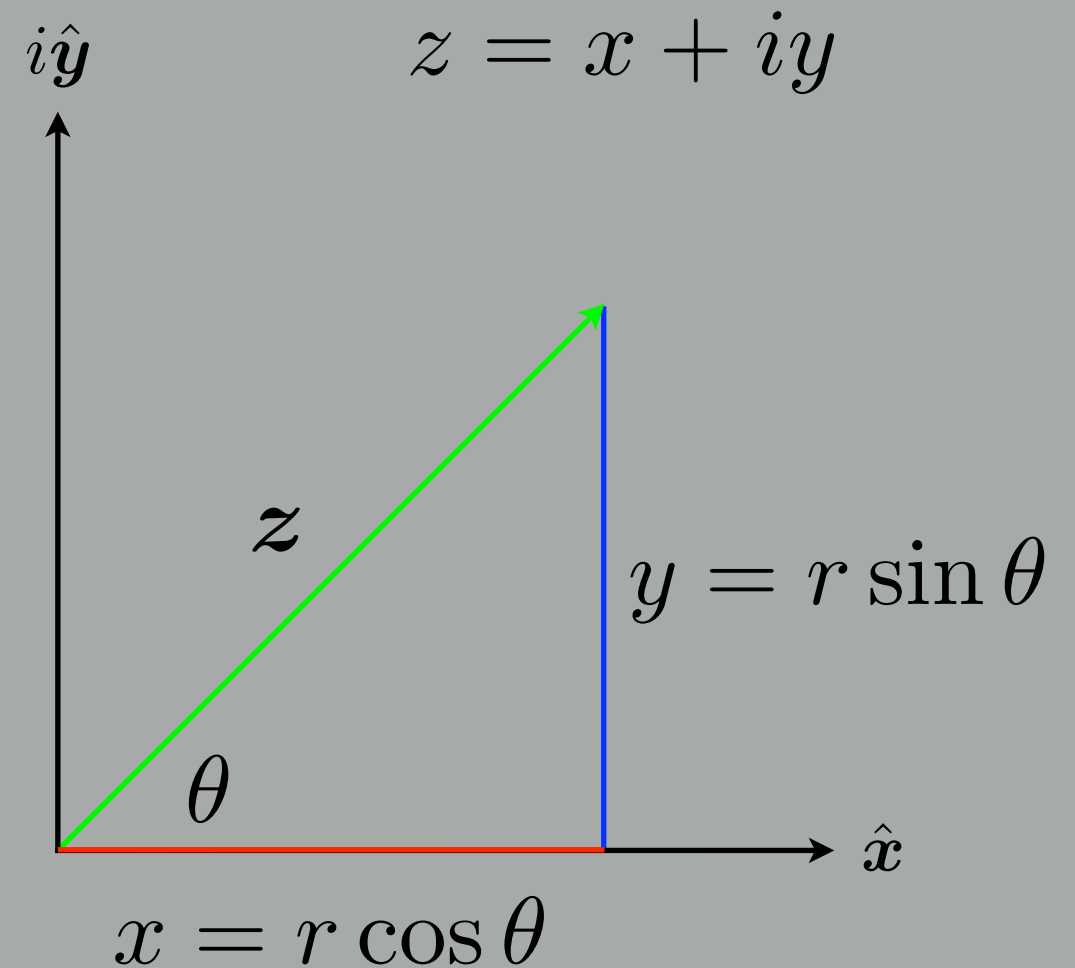
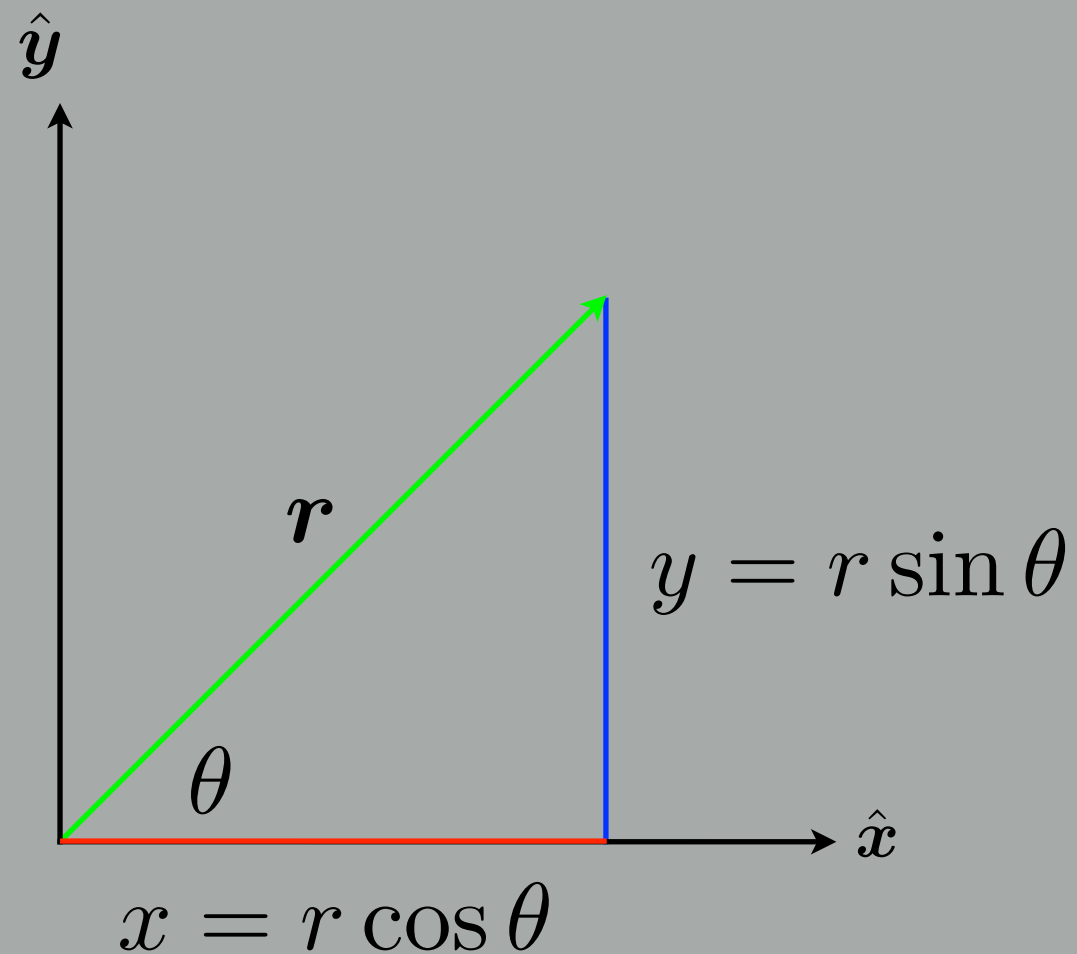
Euler's Relation

$$e^{i\theta} = \cos \theta + i \sin \theta$$

This is the key to the relationship between
complex numbers and *rotations*

$$i = \sqrt{-1}$$

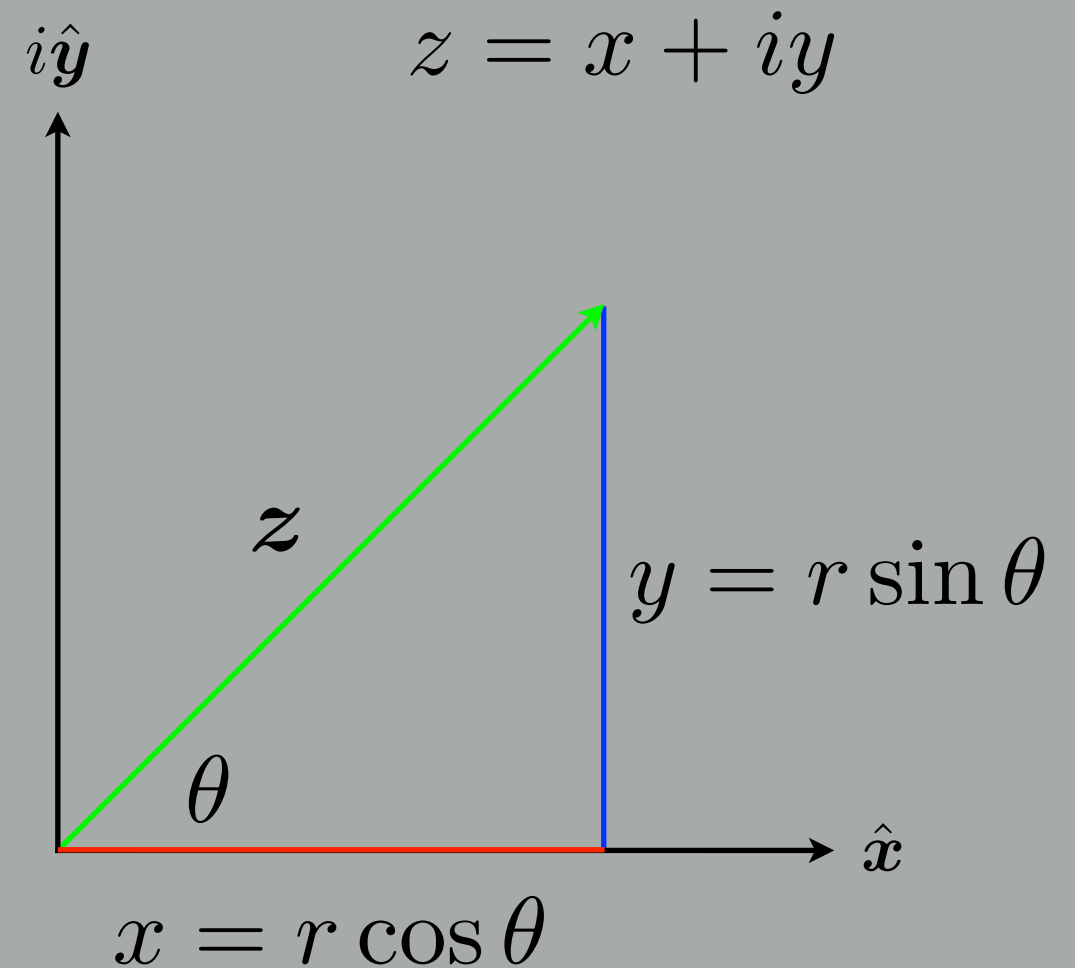
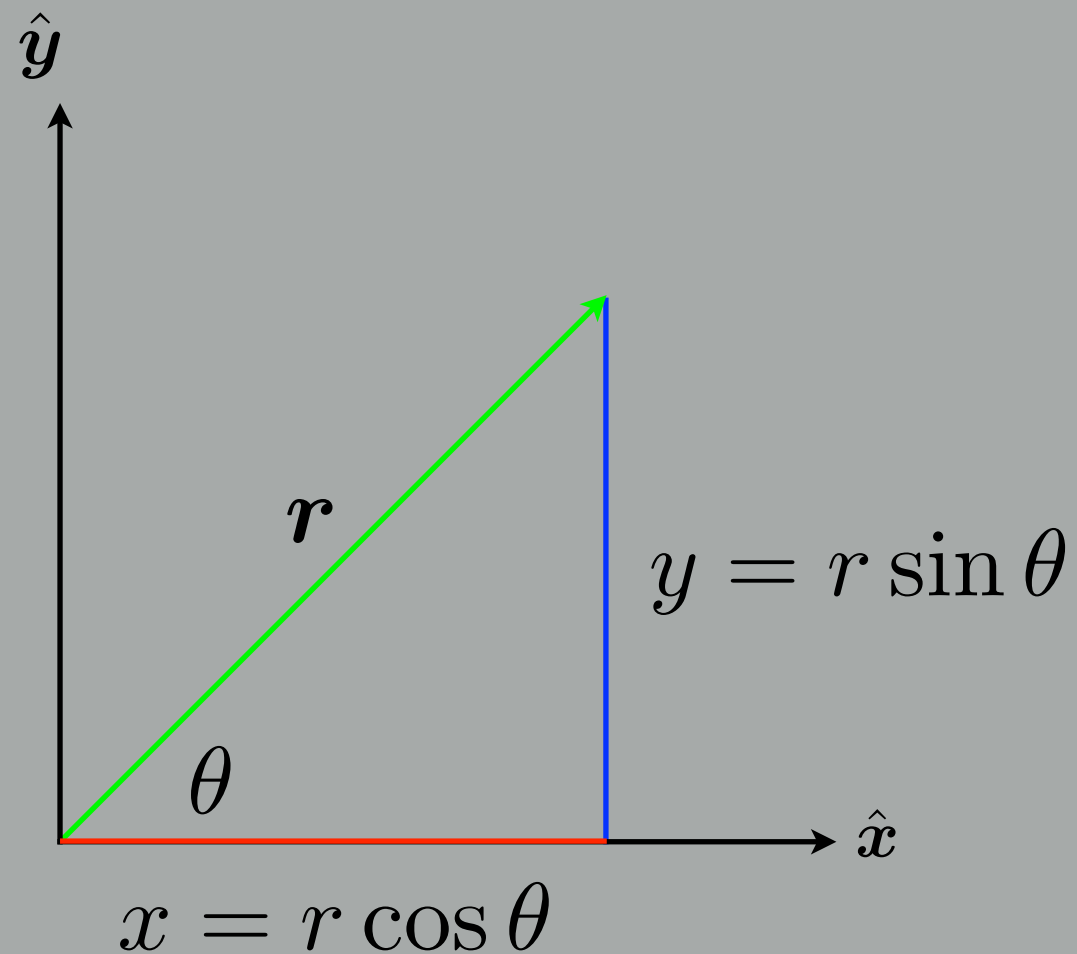
The Complex Plane



$$\|\mathbf{r}\| = \sqrt{x^2 + y^2} = \sqrt{r^2 \cos^2 \theta + r^2 \sin^2 \theta} = \sqrt{r^2 (\cos^2 \theta + \sin^2 \theta)} = \sqrt{r^2}$$

What is $\|\mathbf{z}\|$?

The Complex Plane



$$\|\mathbf{r}\| = \sqrt{x^2 + y^2} = \sqrt{r^2 \cos^2 \theta + r^2 \sin^2 \theta} = \sqrt{r^2 (\cos^2 \theta + \sin^2 \theta)} = \sqrt{r^2}$$

it has to be: $\|z\| = r$

Imaginary Numbers

$$i = \sqrt{-1}$$

$$w = ia, \text{ } a \text{ is real } (a \in \mathfrak{R})$$

What is magnitude of w ?

$$w^2 = i^2 a^2 = -a^2$$

Oops! Can't have negative magnitude

Complex Conjugate

$$w = ia$$

a is real, w is complex

Define: $w^* = -ia$

complex conjugate of w

Replace i with $-i$

Notice what happens when we multiply
a complex number by its complex conjugate

$$ww^* = (ia)(-ia) = -i^2a^2 = a^2$$

we get a real number

Complex Number

$$z = x + iy, \quad x \text{ and } y \text{ real}$$

$$z^* = x - iy \quad \text{Complex conjugate: Replace } i \text{ with } -i$$

Magnitude of z

$$zz^* = (x + iy)(x - iy) = x^2 + y^2$$

$$\|z\| = \sqrt{z^* z}$$

Complex Conjugation

Now let's create a complex number using Euler's relation multiplied by a real number a

$$w = ae^{i\theta} = a(\cos \theta + i \sin \theta)$$

and create its complex conjugate:

$$w^* = ae^{-i\theta} = a(\cos \theta - i \sin \theta)$$

and confirm the result from the previous page:

$$ww^* = (ae^{i\theta})(ae^{-i\theta}) = a^2 e^{i\theta - i\theta} = a^2 e^0 = a^2$$

Complex Number

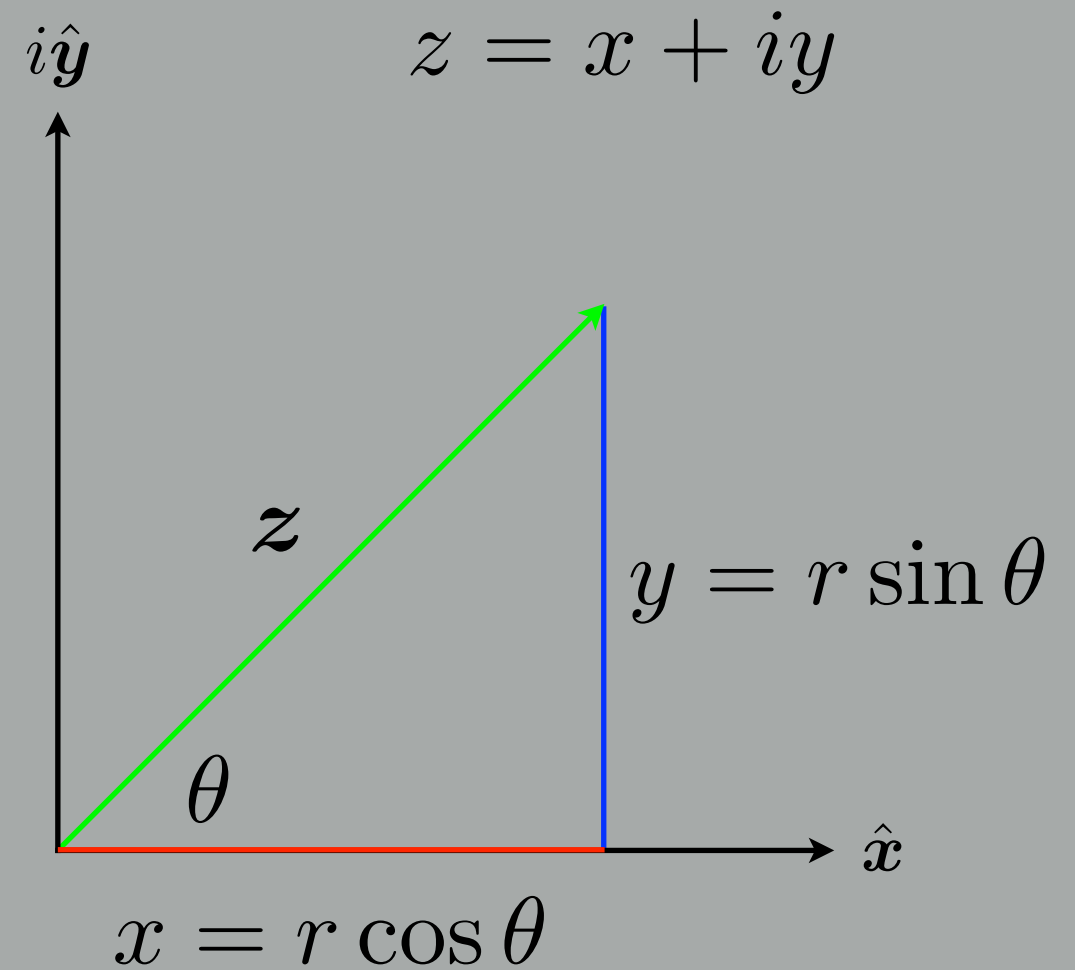
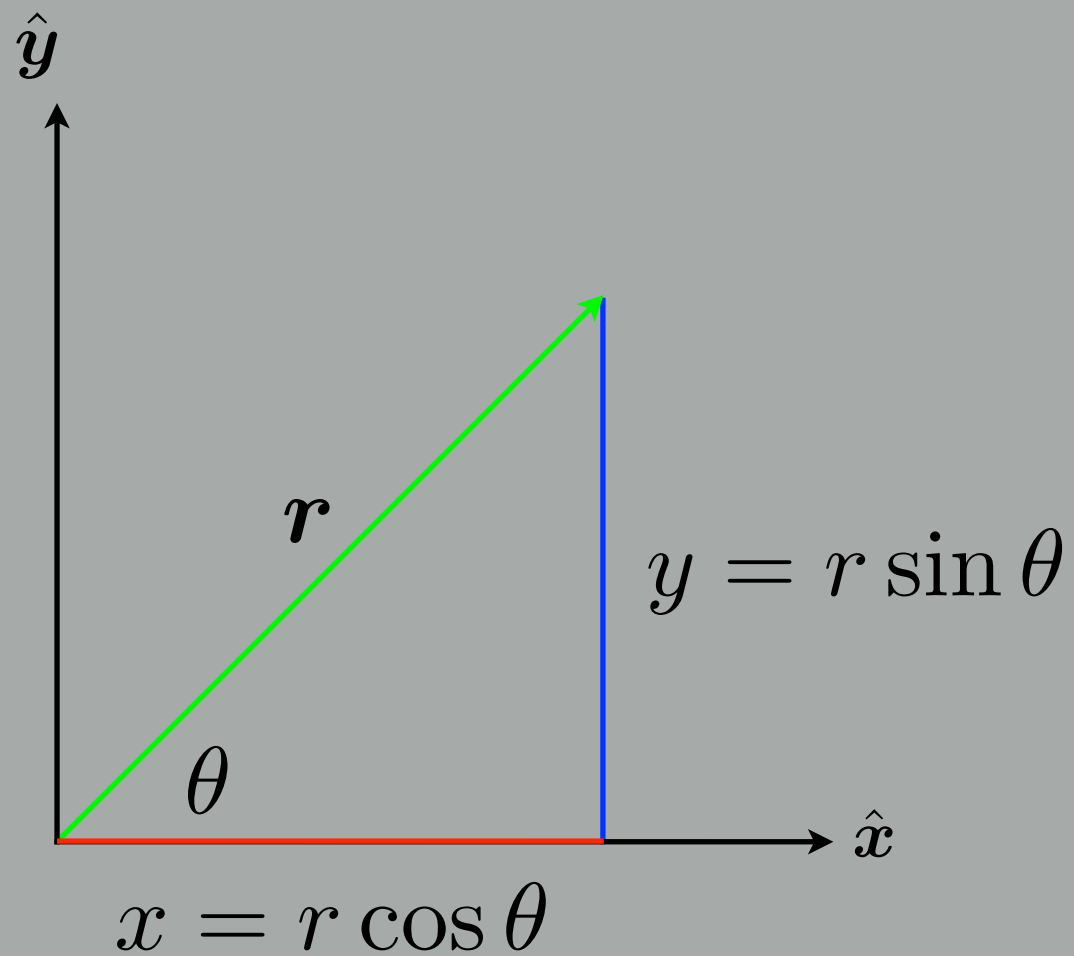
Magnitude of z

$$zz^* = (x + iy)(x - iy) = x^2 + y^2$$

$$\|z\| = \sqrt{z^*z}$$

$$\|z\| = \sqrt{(re^{i\theta})(re^{-i\theta})} = r$$

The Complex Plane



$$|\mathbf{r}| = \sqrt{x^2 + y^2} = \sqrt{r^2 \cos^2 \theta + r^2 \sin^2 \theta} = \sqrt{r^2 (\cos^2 \theta + \sin^2 \theta)} = \sqrt{r^2}$$

$$|z| = \sqrt{(x + iy)(x - iy)} = \sqrt{x^2 + y^2}$$

Magnitude and phase

Writing complex numbers $z = x + iy$ by using Euler's relation

$$z = re^{i\theta}$$

is very convenient since the magnitude is

$$|z| = \sqrt{zz^*} = \begin{cases} \sqrt{(z + iy)(z - iy)} = \sqrt{x^2 + y^2} & \text{complex Cartesian} \\ \sqrt{re^{i\theta}re^{-i\theta}} = \sqrt{r^2} = r & \text{using Euler's relation} \end{cases}$$

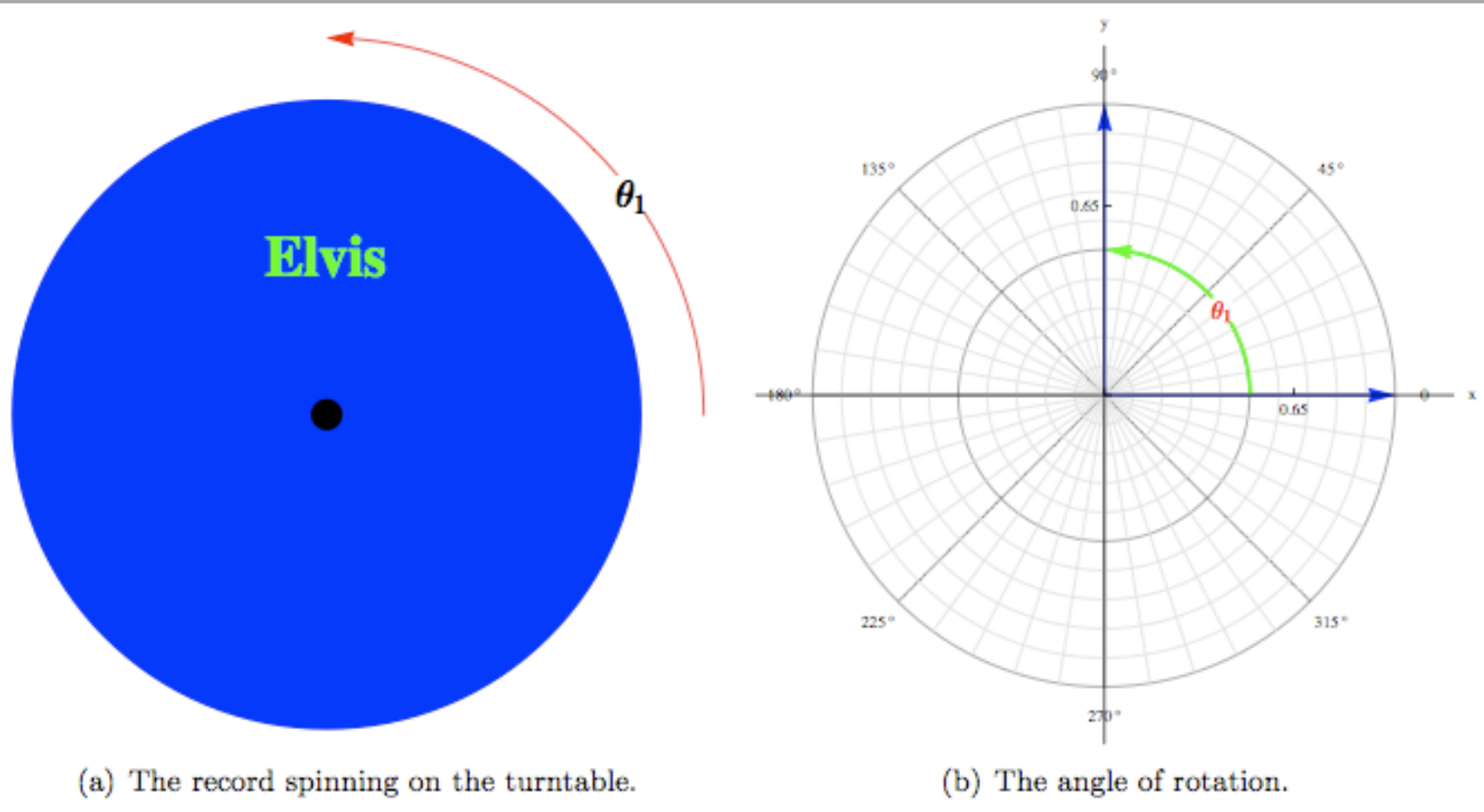
and the phase is

$$\angle |z| = \begin{cases} \tan^{-1} \left(\frac{y}{x} \right) & \text{complex Cartesian} \\ \theta = \arg z & \text{using Euler's relation} \end{cases}$$

symbol for angle

stands for "argument" and mean whatever multiplies i in exponent, which *is* the phase

Phase angle

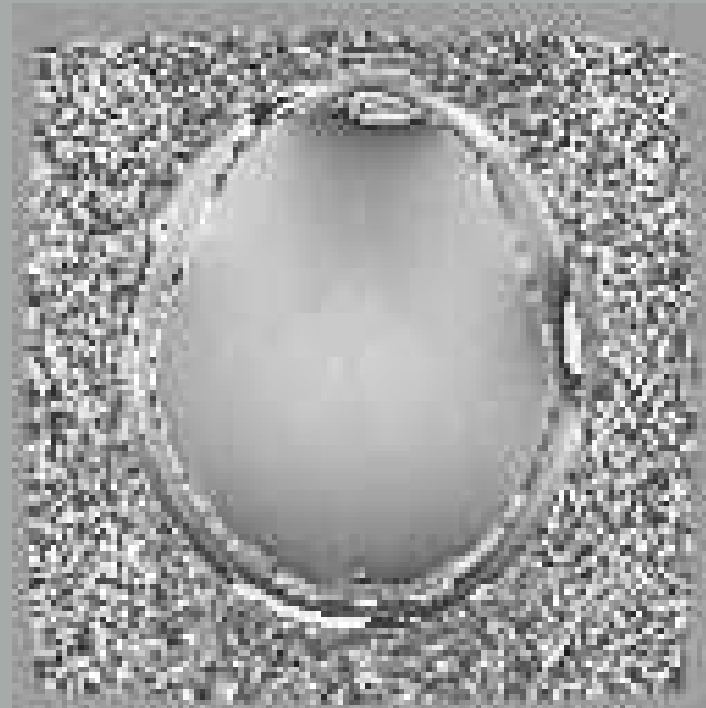


Example: Field Maps

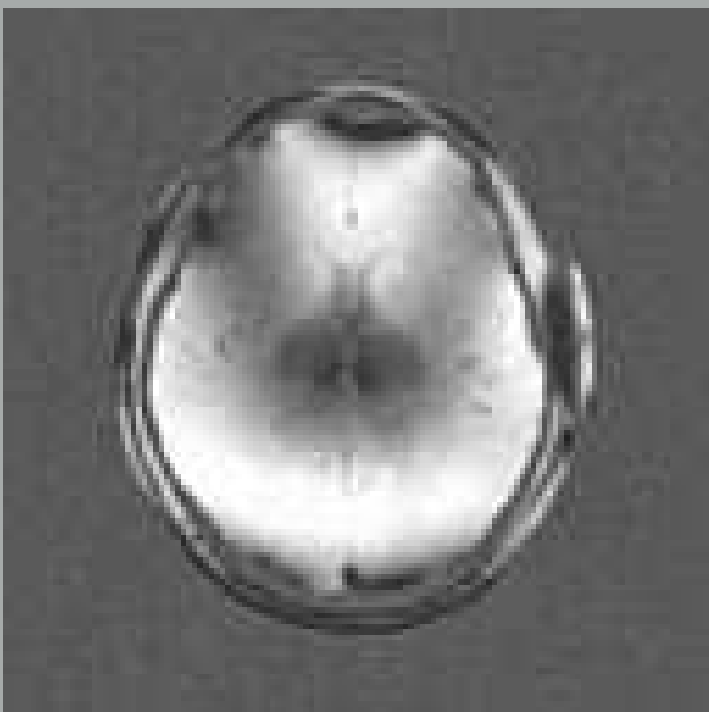
magnitude



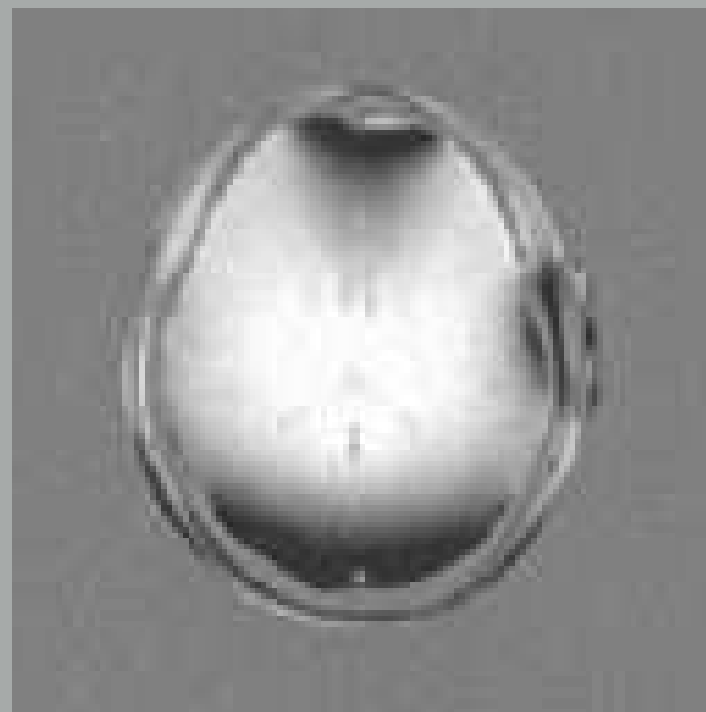
phase



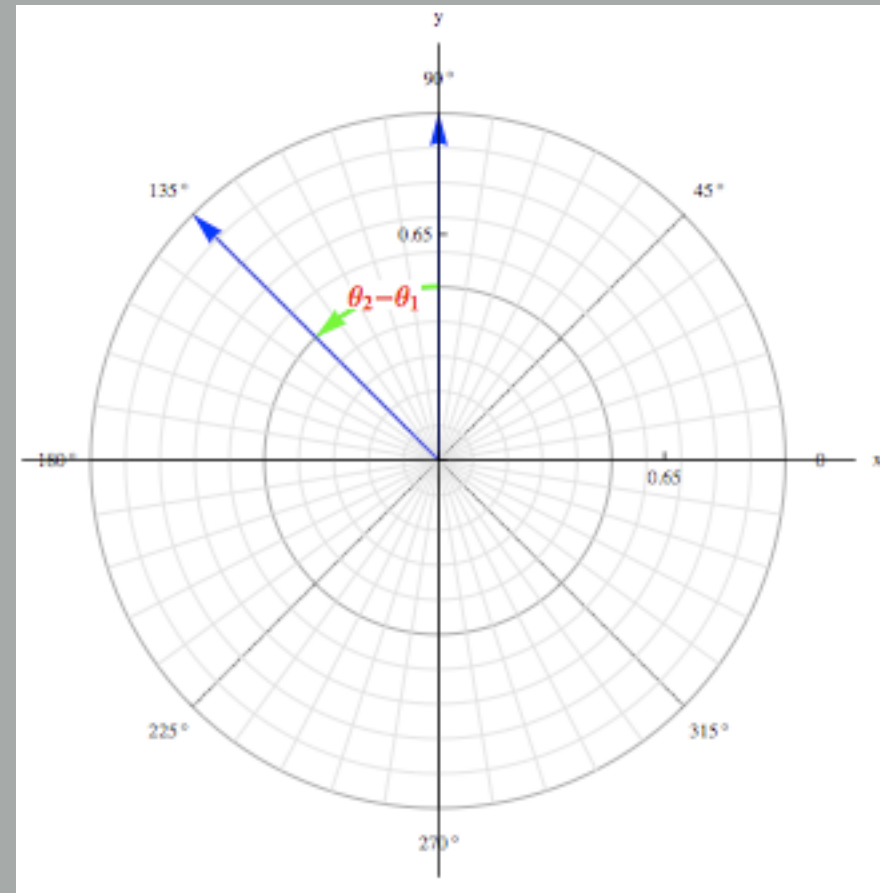
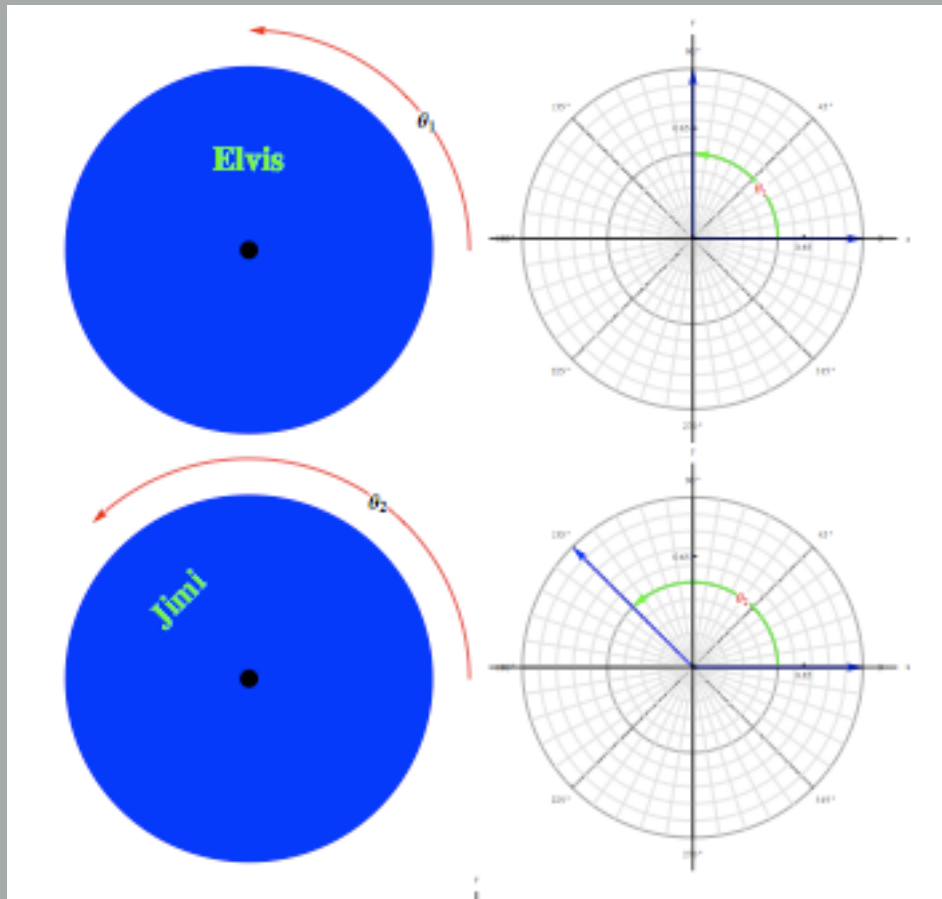
real



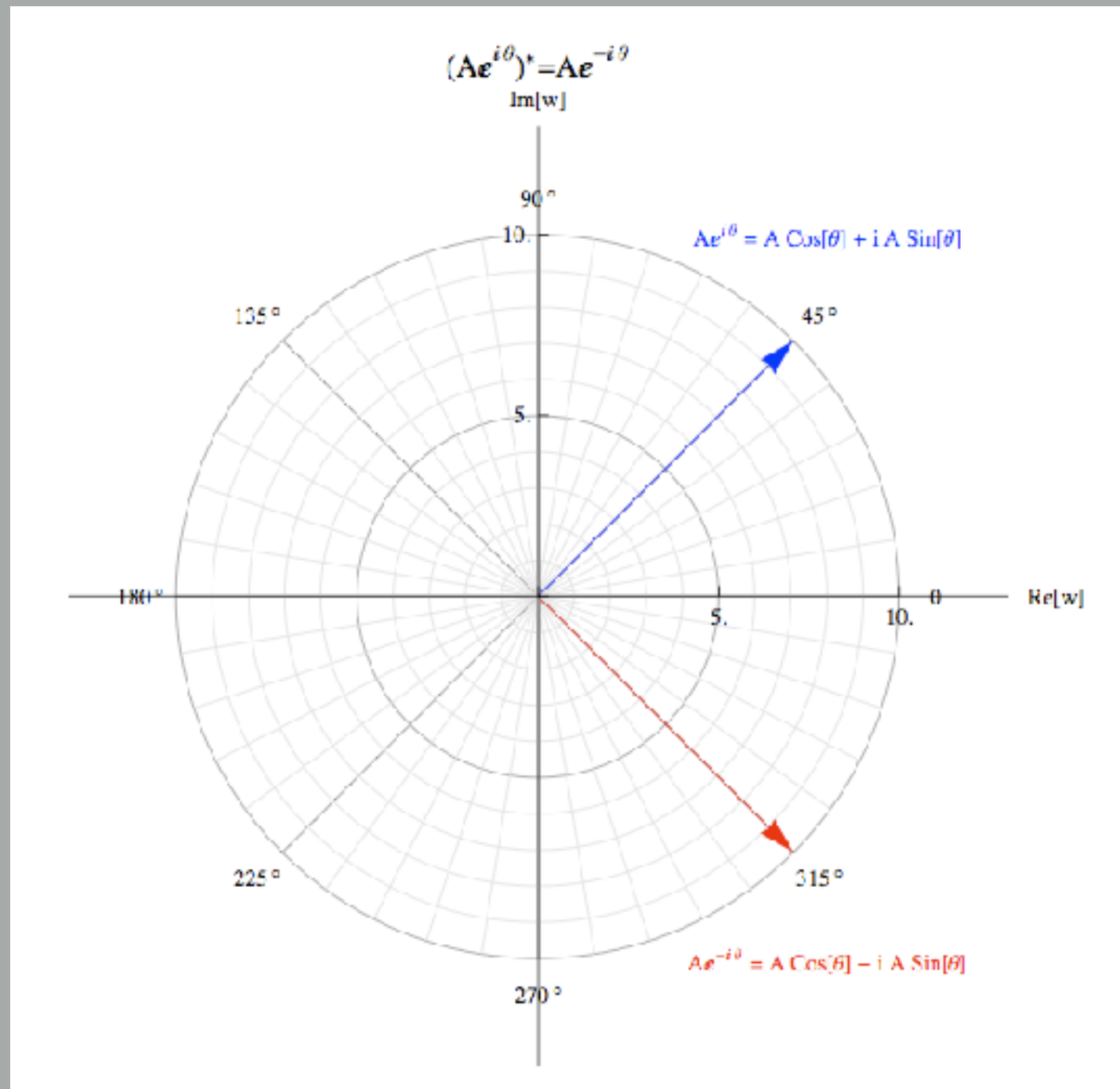
imaginary



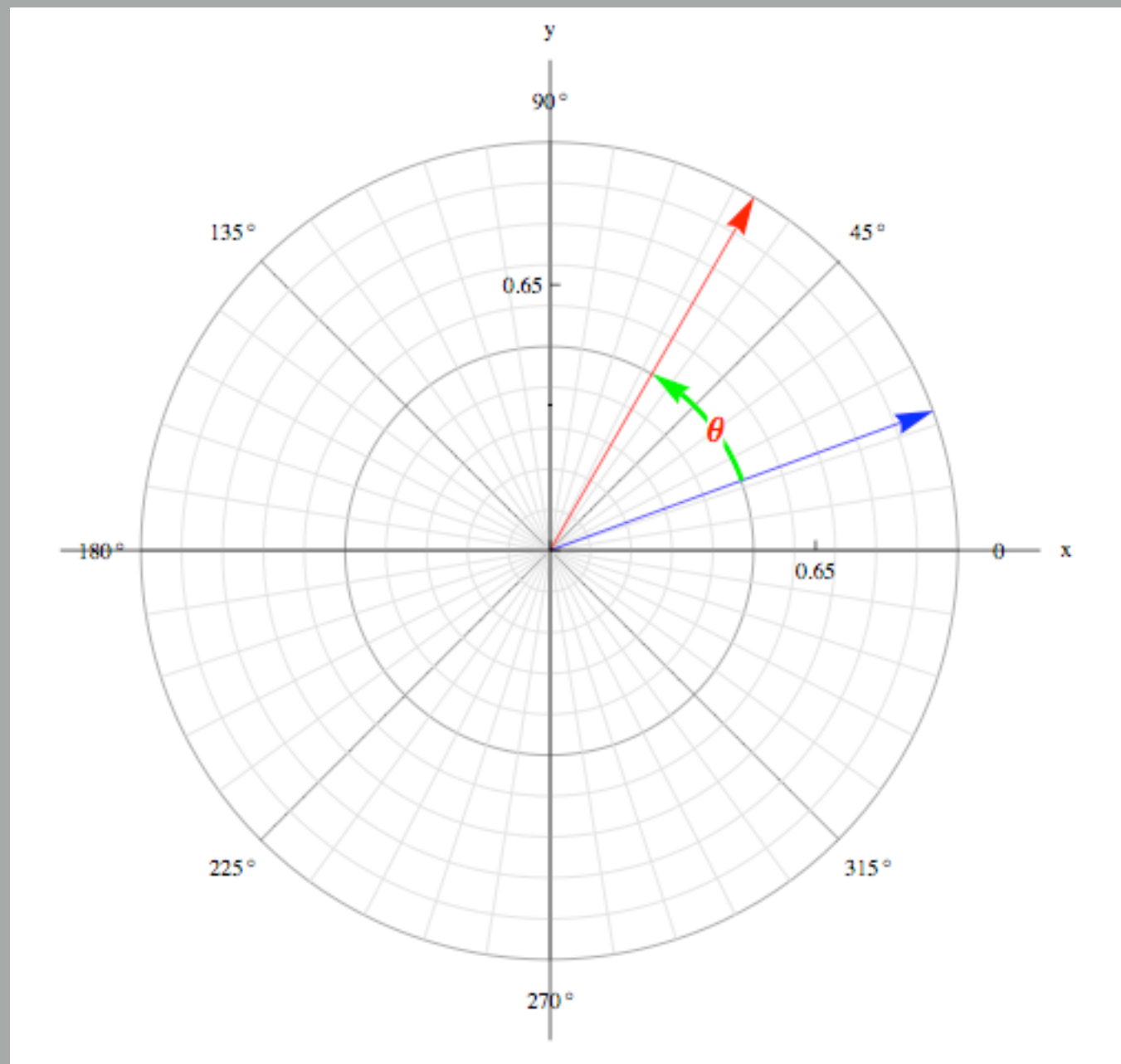
Phase difference



Complex Conjugation

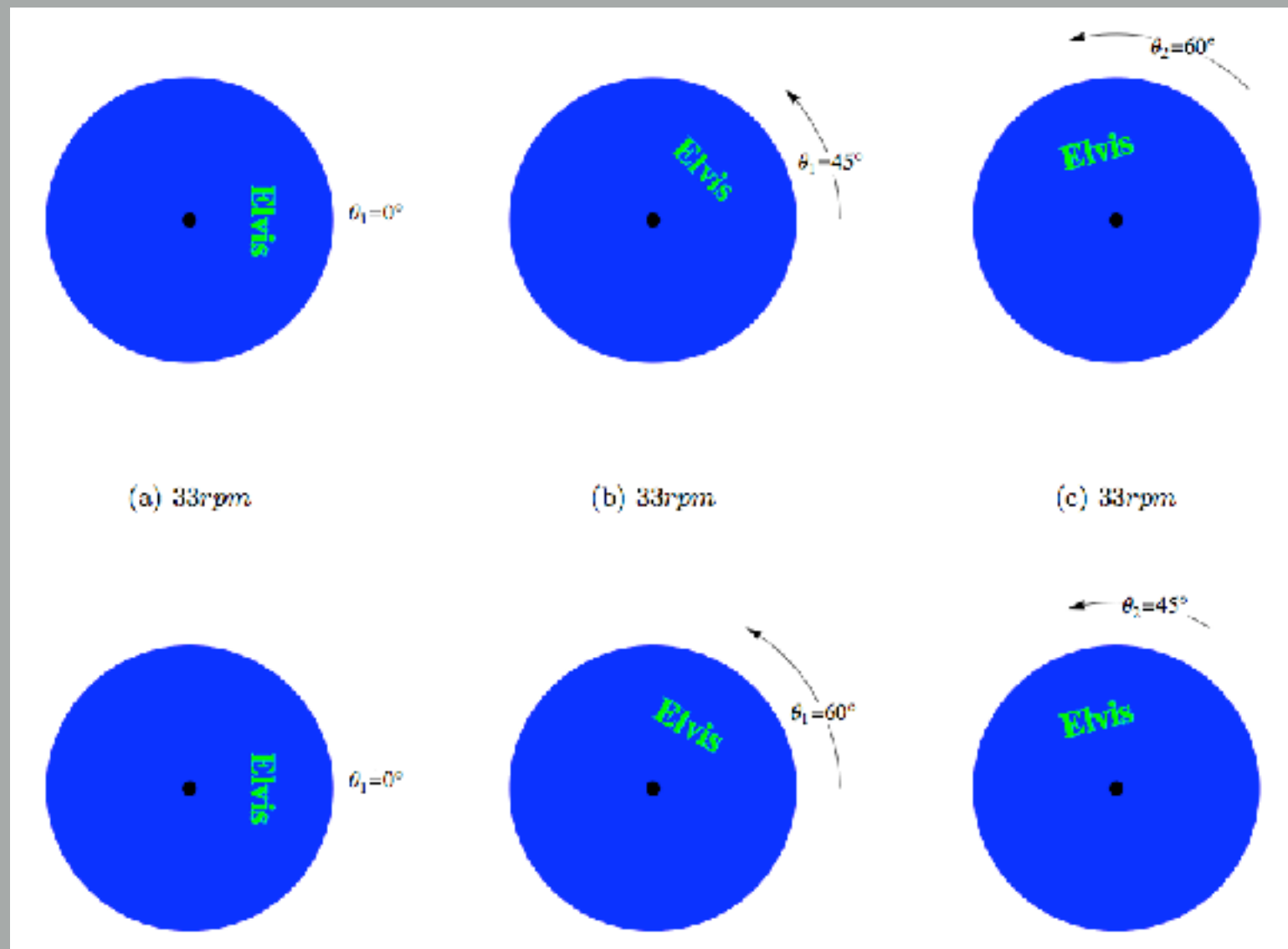


Rotations in 2D



$$u = R(\theta)v$$

Order of Rotations in 2D



$$R(\theta_1)R(\theta_2) = R(\theta_2)R(\theta_1)$$

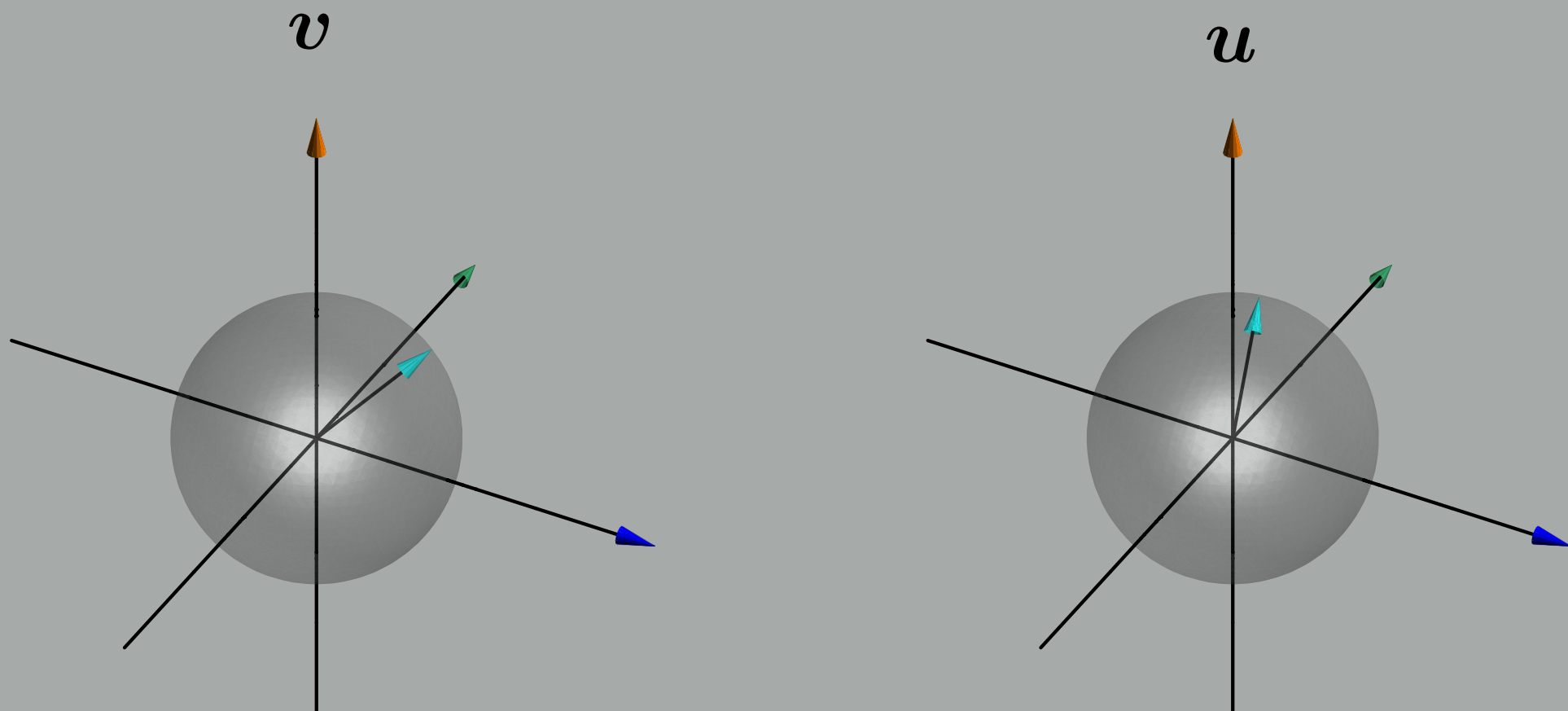
Order of Rotations in 2D

Order of rotations in 2D does not matter
i.e.,

rotations in 2D *commute*

$$\mathbf{R}(\theta_1)\mathbf{R}(\theta_2) = \mathbf{R}(\theta_2)\mathbf{R}(\theta_1)$$

Vector rotations in 3D



Just as in 2D,

$$u = Rv$$

Where R is now a 3×3 rotation matrix

Rotation Matrices in 3D

$$\mathbf{R}_x(\alpha) = \begin{pmatrix} 1 & 0 & 0 \\ 0 & \cos \alpha & \sin \alpha \\ 0 & -\sin \alpha & \cos \alpha \end{pmatrix}$$

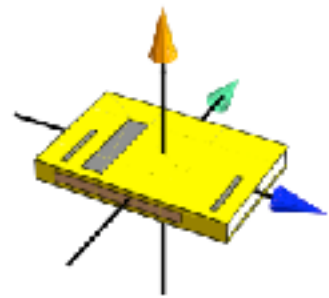
$$\mathbf{R}_y(\beta) = \begin{pmatrix} \cos \beta & 0 & \sin \beta \\ 0 & 1 & 0 \\ -\sin \beta & 0 & \cos \beta \end{pmatrix}$$

$$\mathbf{R}_z(\gamma) = \begin{pmatrix} \cos \gamma & \sin \gamma & 0 \\ -\sin \gamma & \cos \gamma & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

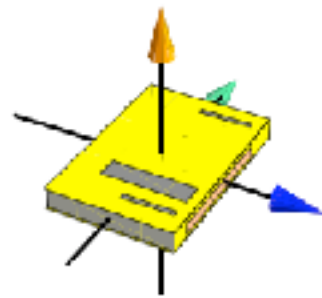
Object Rotations in 3D

But *object* rotations are another story...

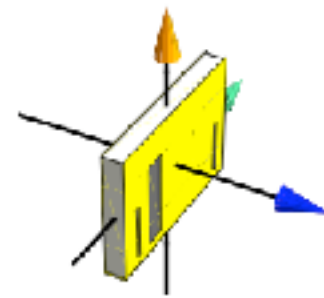
Order of Rotations in 3D



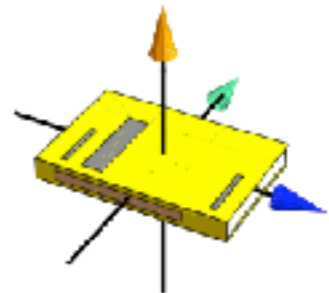
(a) No rotation.



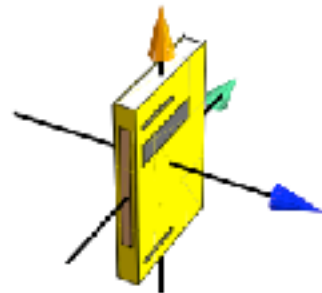
(b) Rotation of 90° about $\hat{z}$.



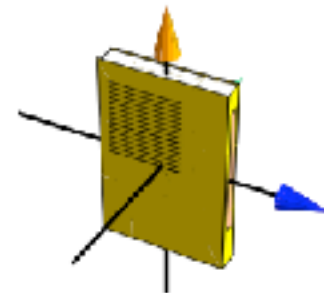
(c) Rotation of 90° about $\hat{y}$.



(d) No rotation.



(e) Rotation of 90° about $\hat{y}$.



(f) Rotation of 90° about $\hat{z}$.

$$\mathbf{R}(\theta_1)\mathbf{R}(\theta_2) \neq \mathbf{R}(\theta_2)\mathbf{R}(\theta_1)$$

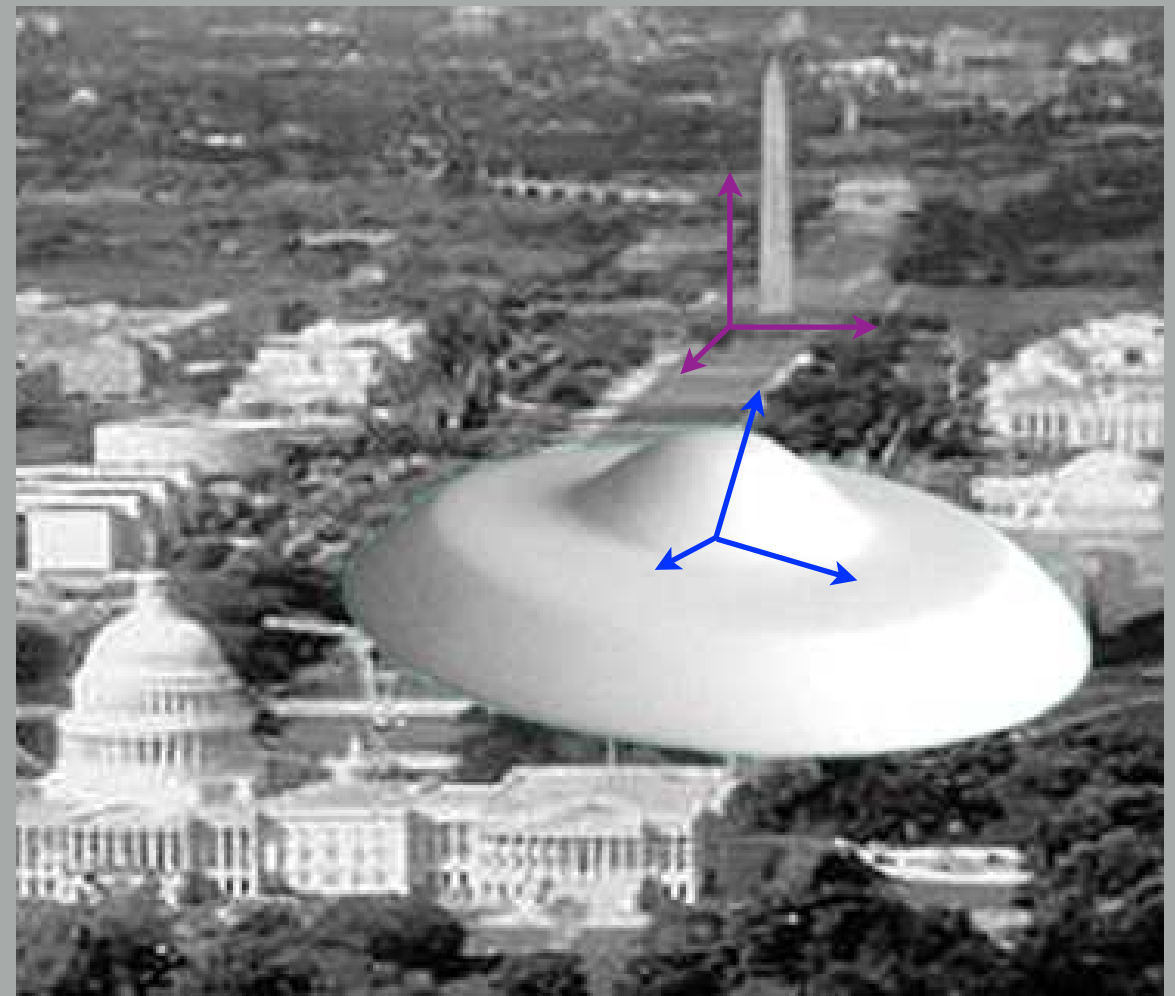
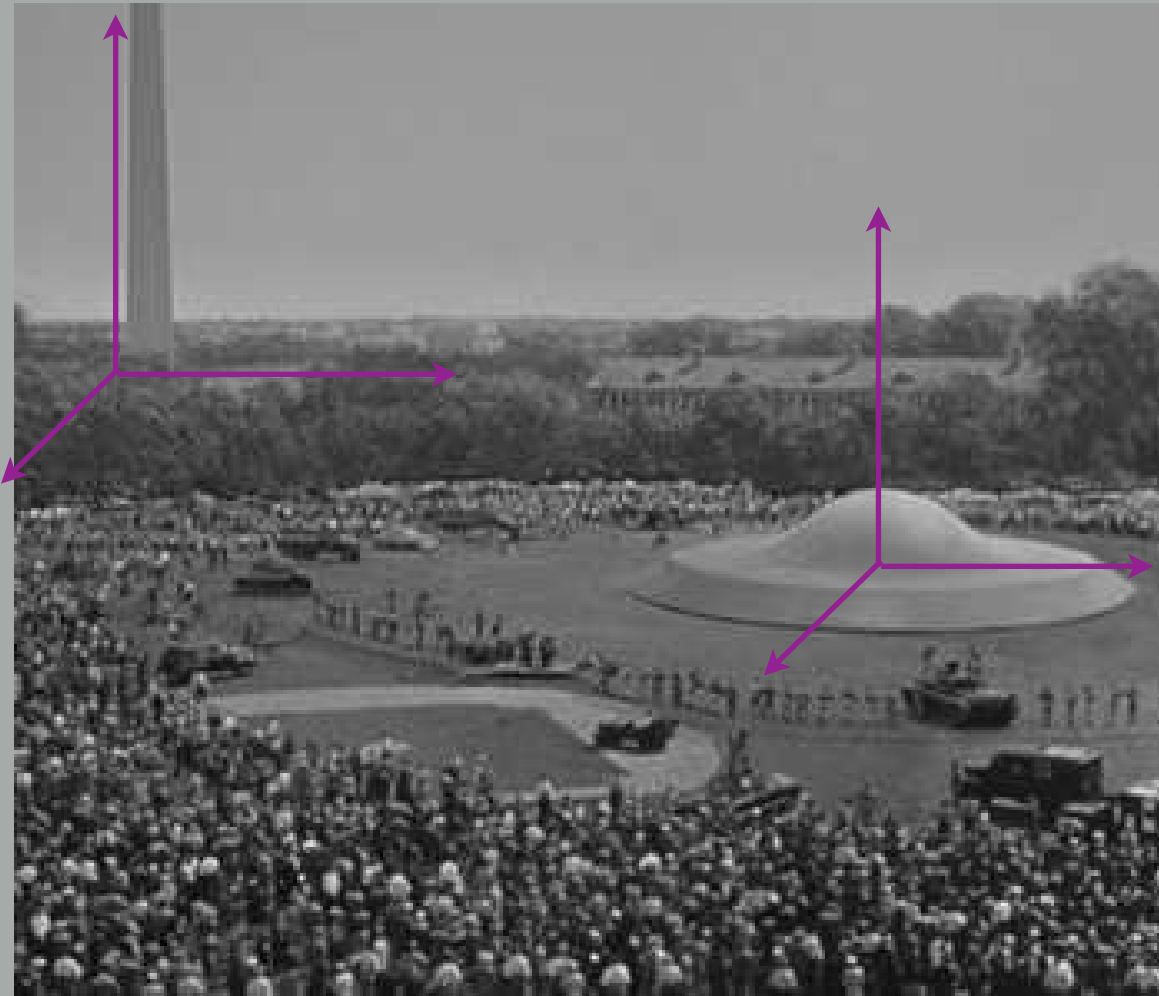
Order of Rotations in 3D

Order of rotations in 3D *does* matter
i.e.,

rotations in 3D do *not* ^{necessarily} commute

$$\mathbf{R}(\theta_1)\mathbf{R}(\theta_2) \neq \mathbf{R}(\theta_2)\mathbf{R}(\theta_1)$$

Describing Rotations in 3D



What's different about 3D?



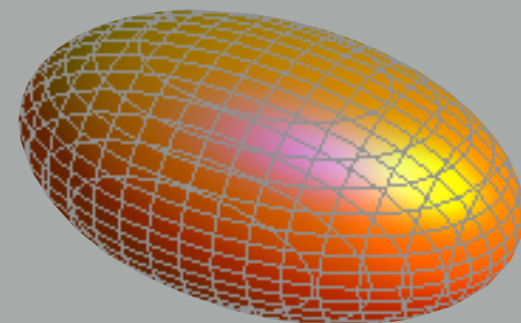
Lab (field) coordinates



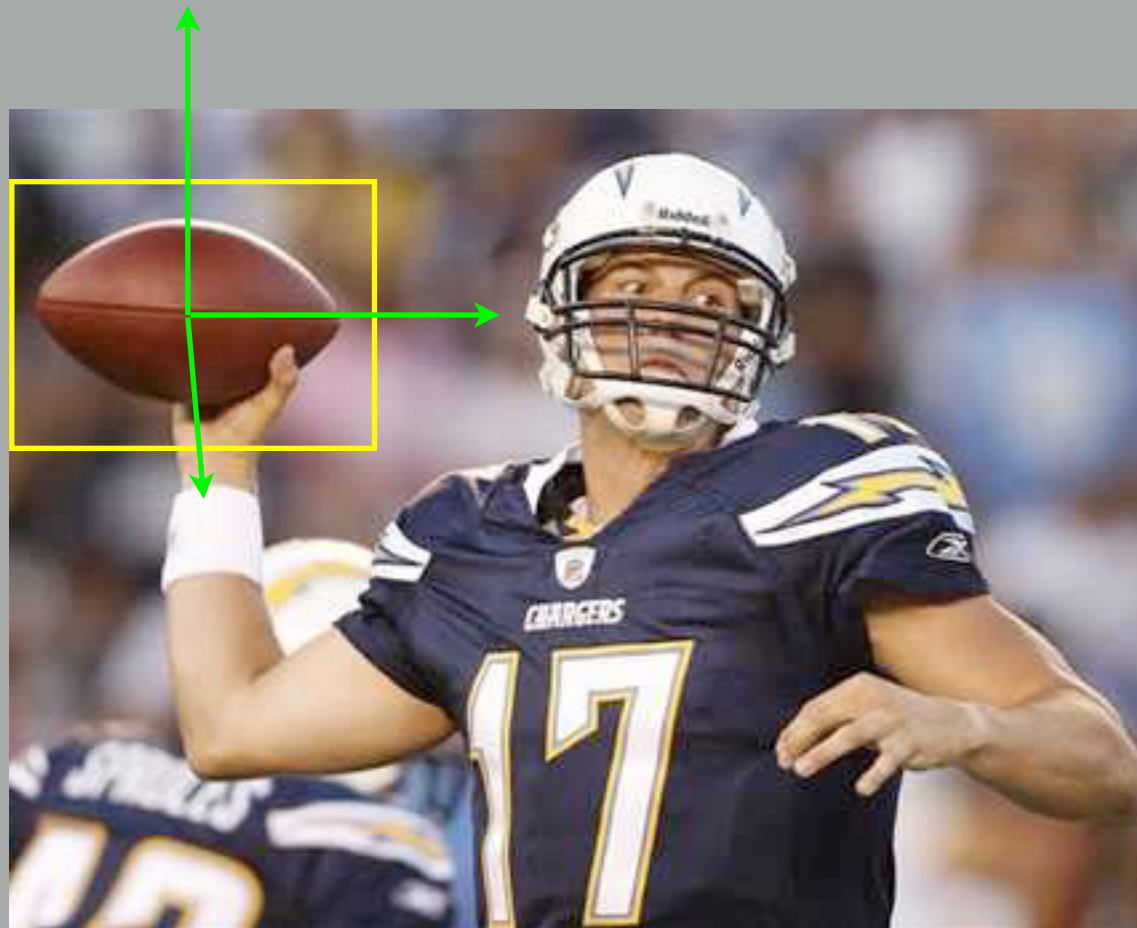
What's different about 3D?



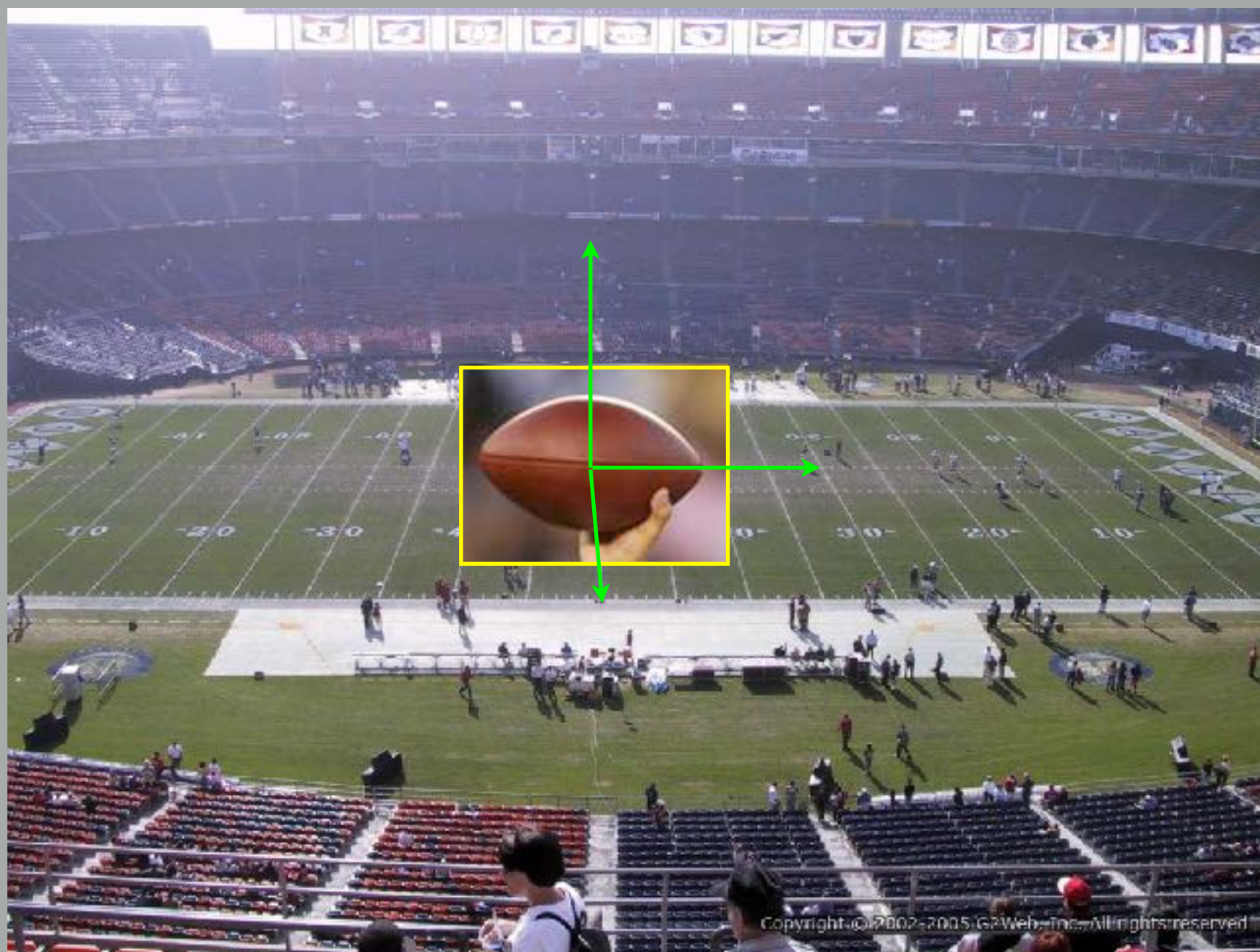
$\approx$



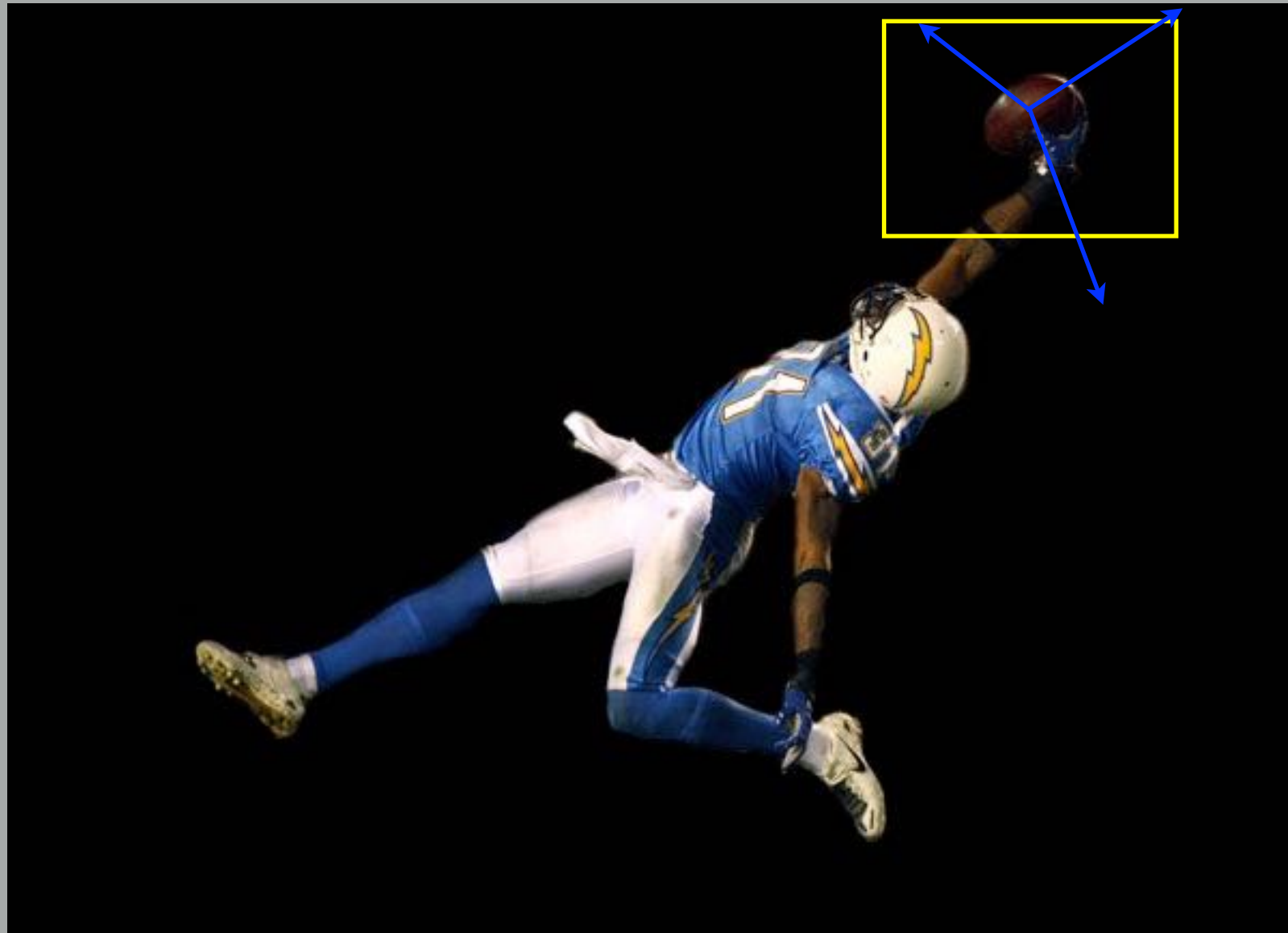
What's different about 3D?



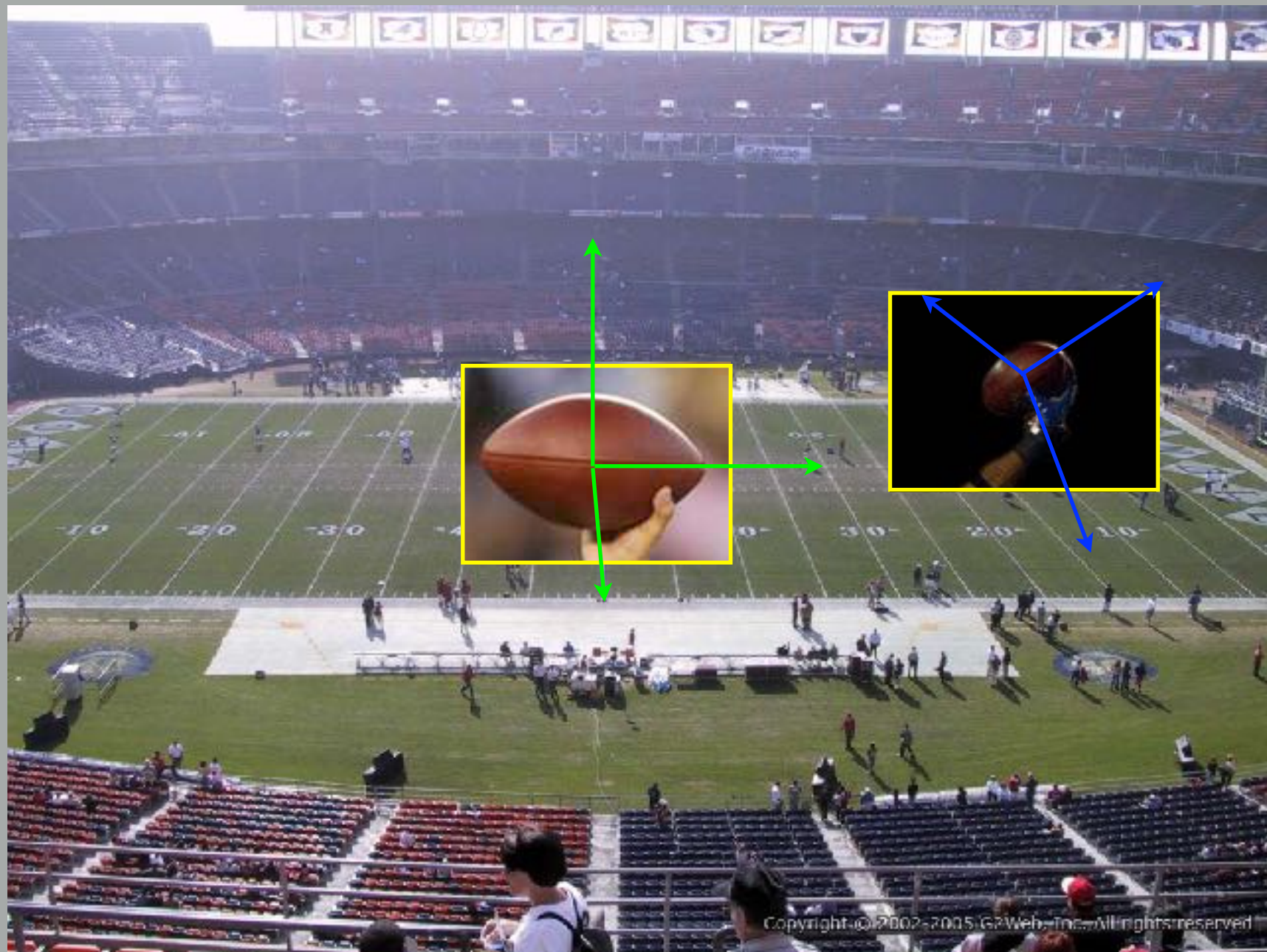
Lab (field) coordinates



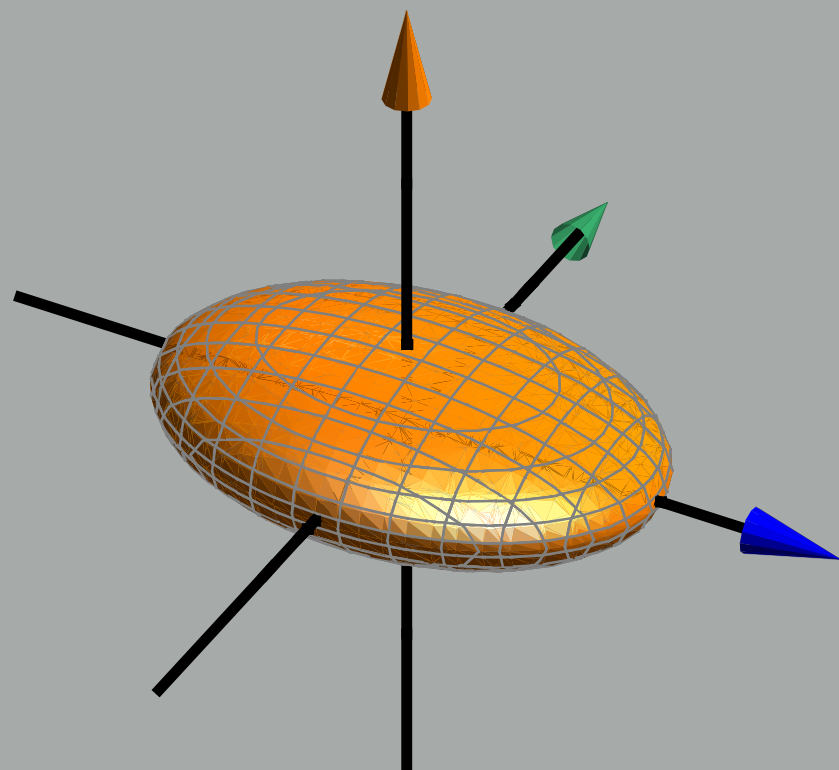
What's different about 3D?



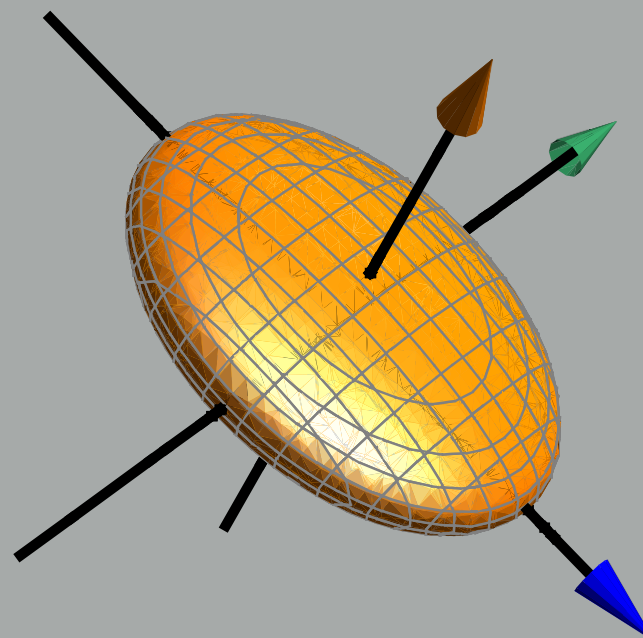
Player coordinates



Describing Rotations in 3D

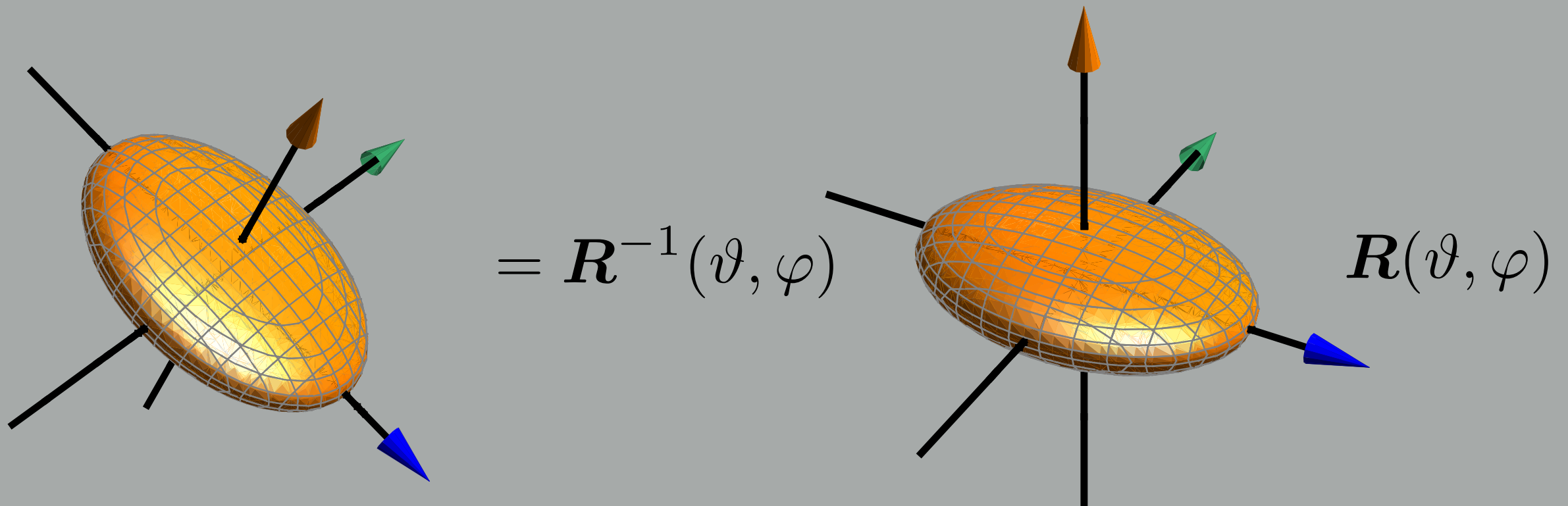


S



S'

Similarity Transform

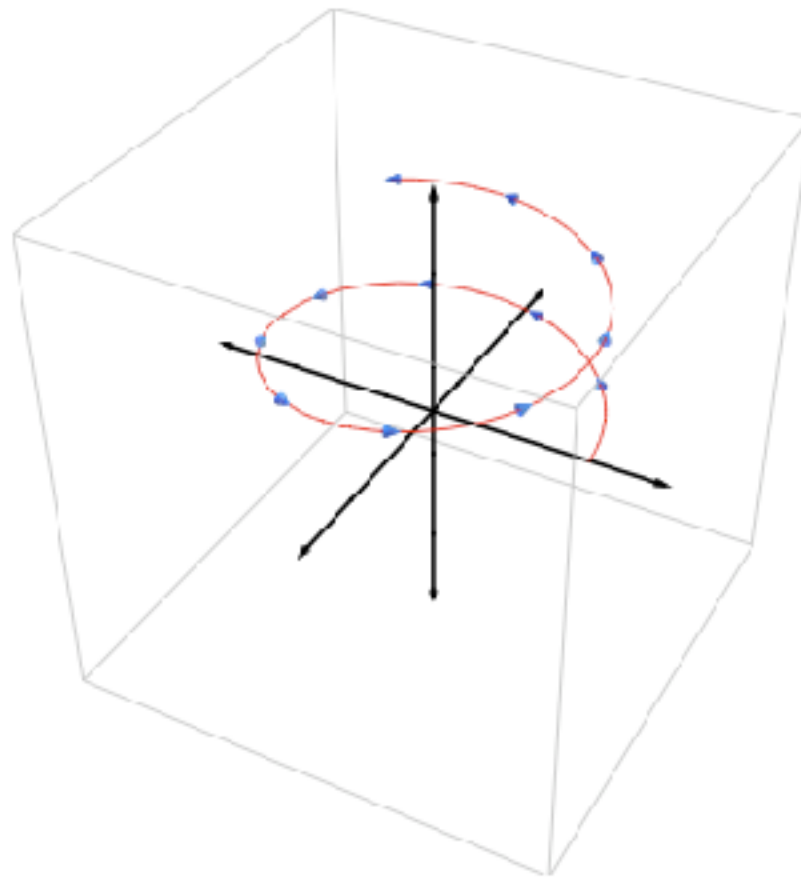


$$S' = R^{-1} S R$$

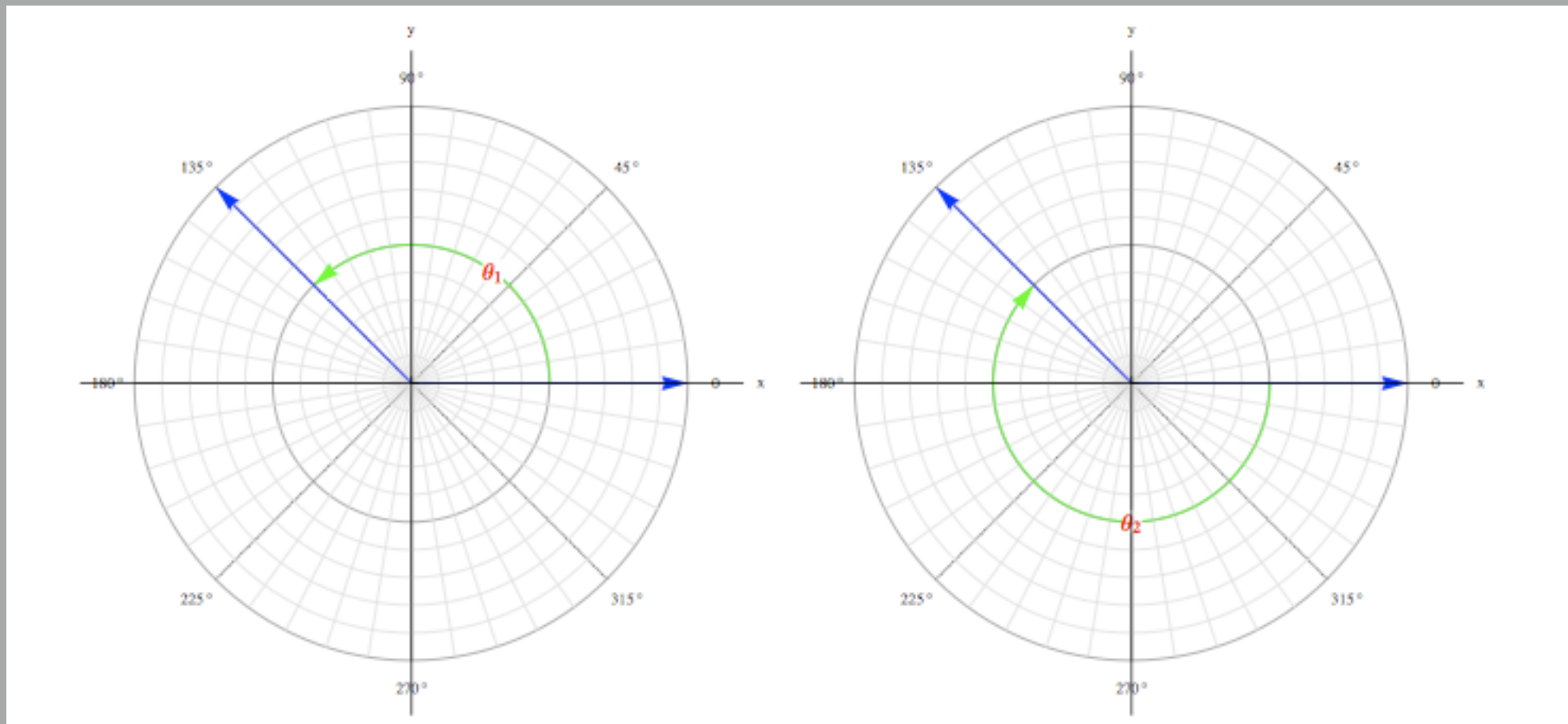
Looks just the same as in 2D!

Extra slides

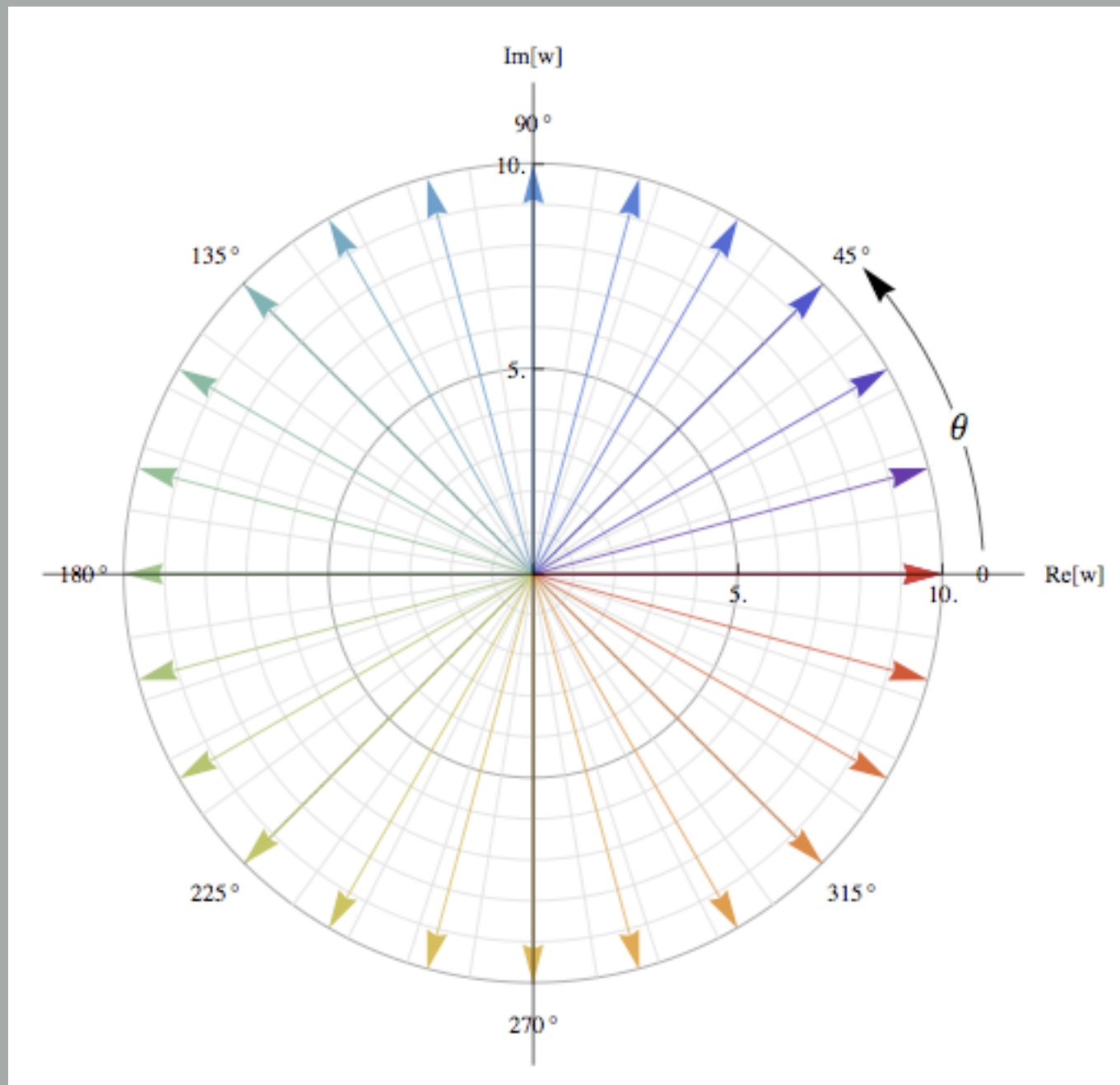
Phase Wrapping



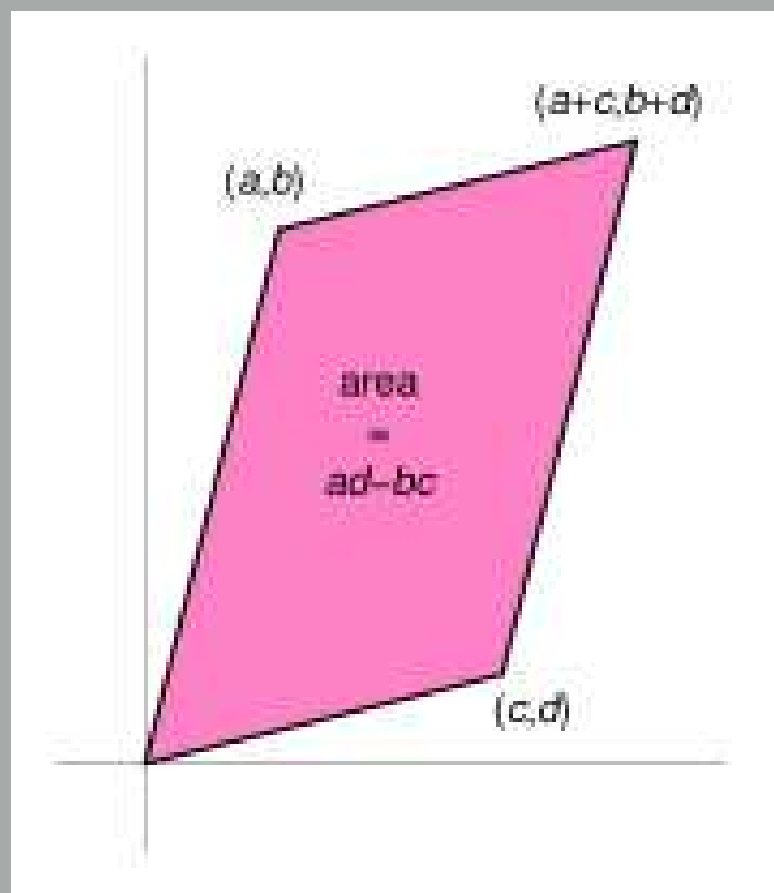
Aliasing



The Complex Plane

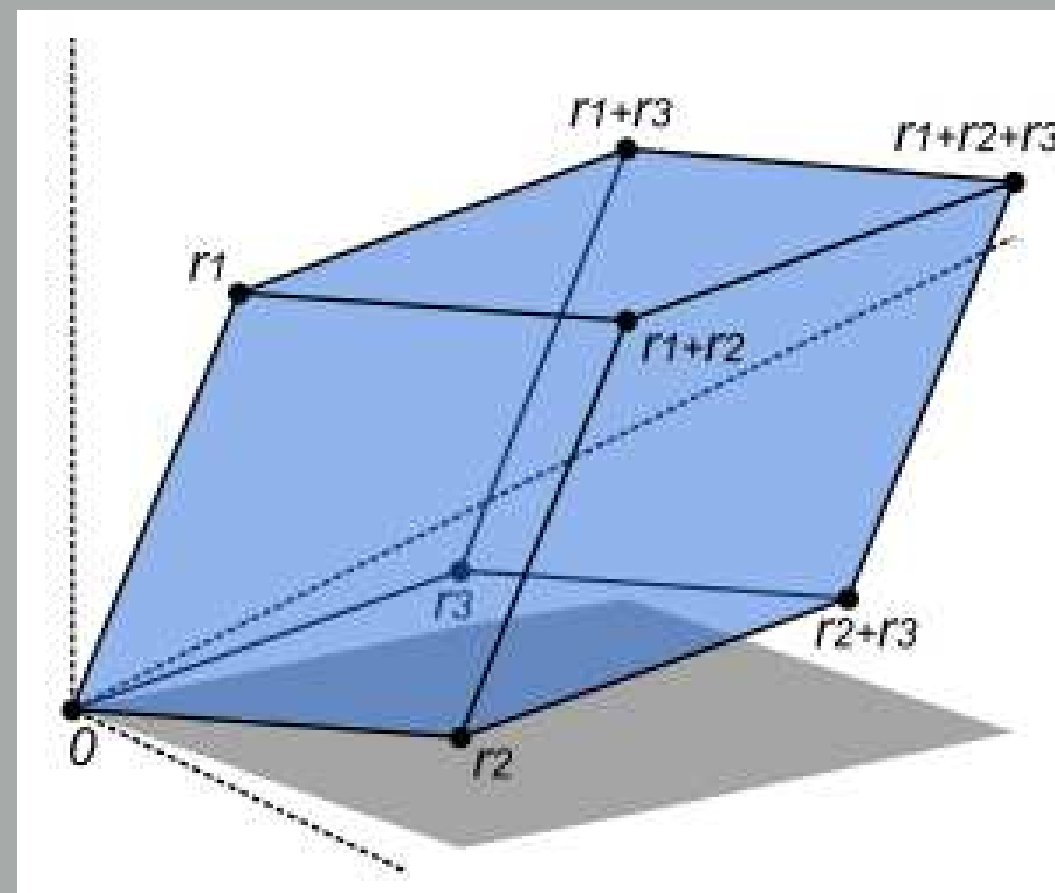


Matrix Determinant



$$|\mathbf{A}| = \text{area}$$

$\mathbf{A} = 2 \times 2$ real matrix



$$|\mathbf{A}| = \text{volume}$$

$\mathbf{A} = 3 \times 3$ real matrix

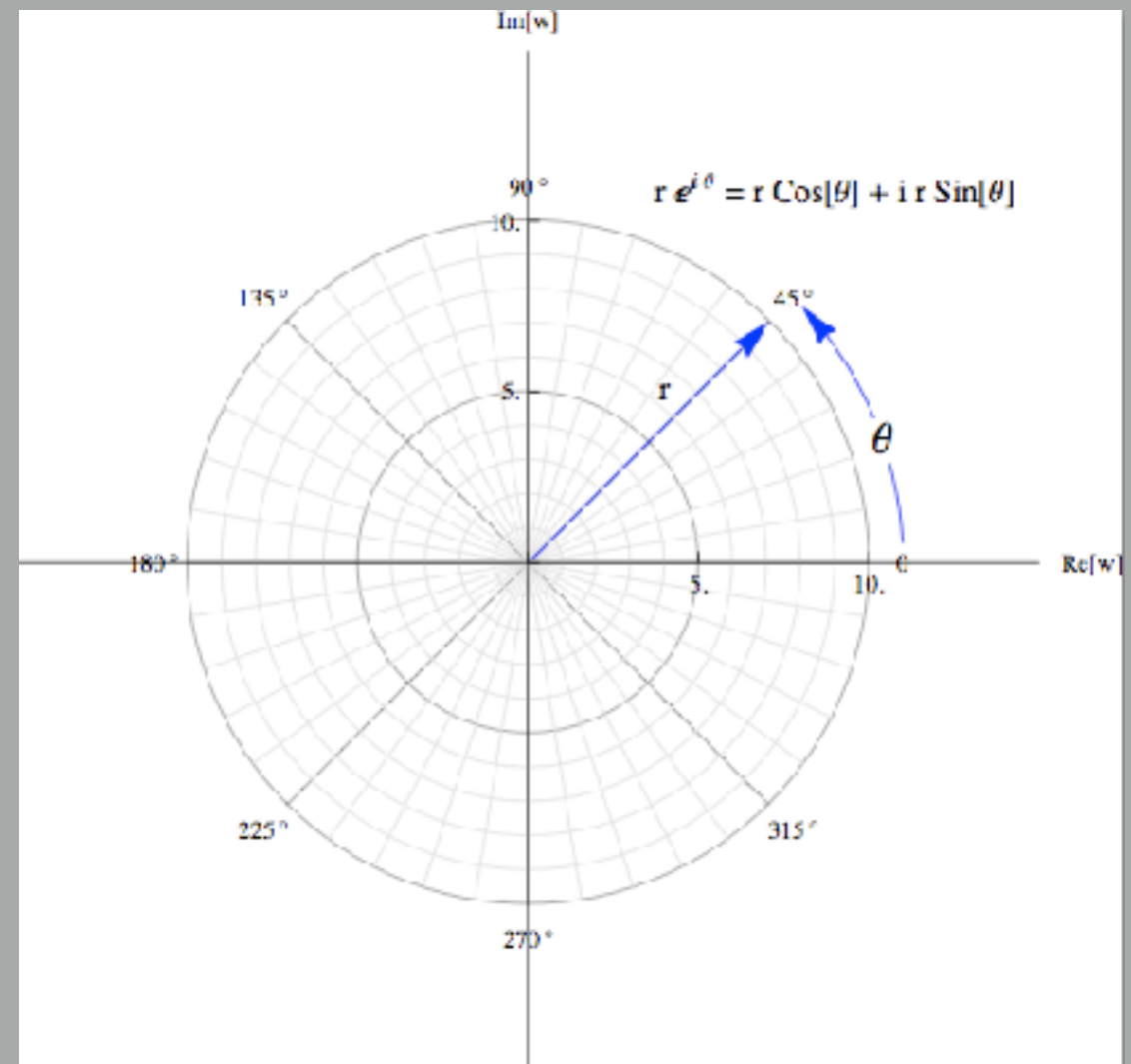
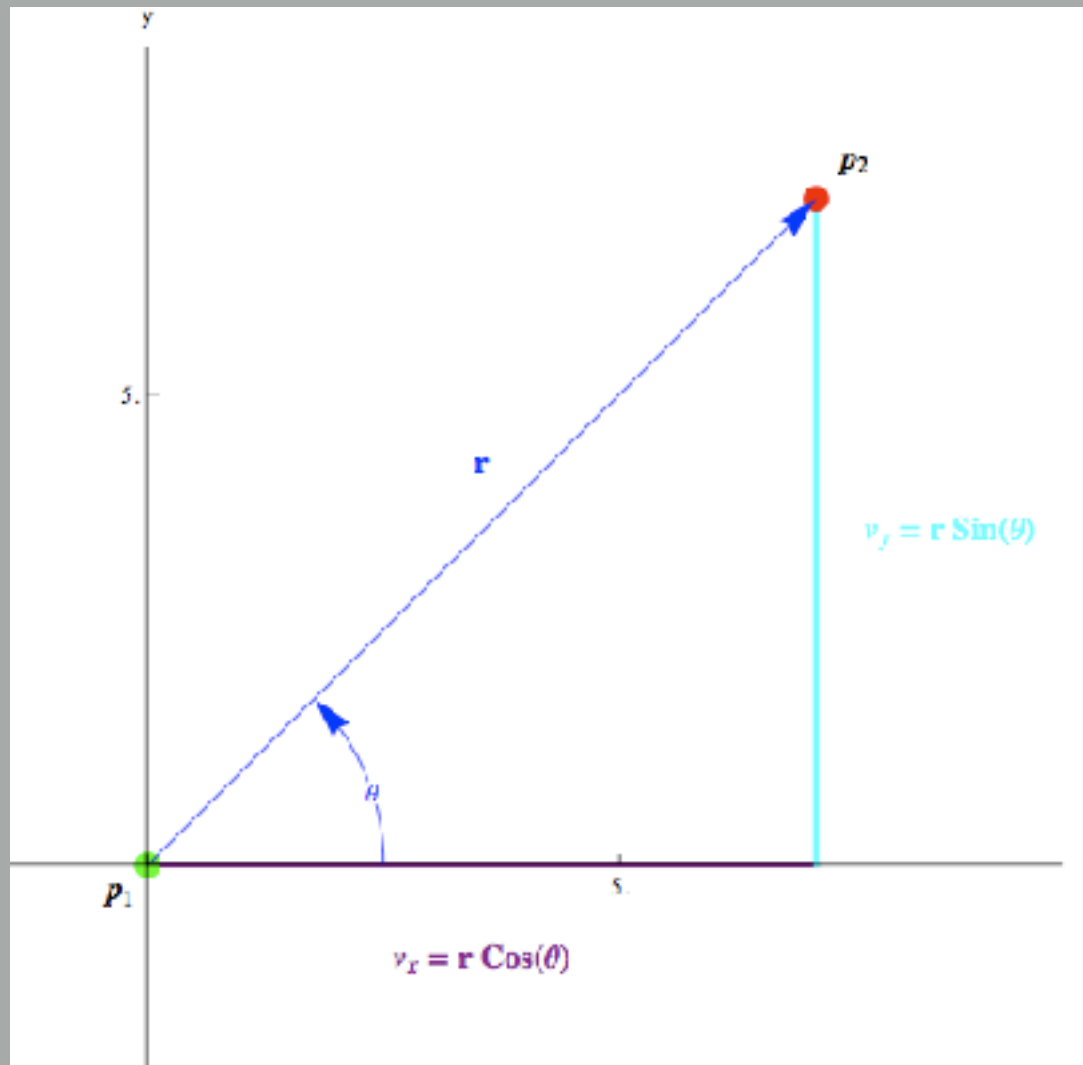
Matrix Multiplication

Dot Product

$$\begin{pmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{pmatrix} \begin{pmatrix} b_{11} & b_{12} & b_{13} \\ b_{21} & b_{22} & b_{23} \\ b_{31} & b_{32} & b_{33} \end{pmatrix} = \begin{pmatrix} c_{11} & c_{12} & c_{13} \\ c_{21} & c_{22} & c_{23} \\ c_{31} & c_{32} & c_{33} \end{pmatrix}$$

$$c_{12} = \begin{pmatrix} a_{11} & a_{12} & a_{13} \end{pmatrix} \begin{pmatrix} b_{12} \\ b_{22} \\ b_{32} \end{pmatrix} = a_{11}b_{12} + a_{12}b_{22} + a_{13}b_{32}$$

Complex Numbers



Vector Multiplication

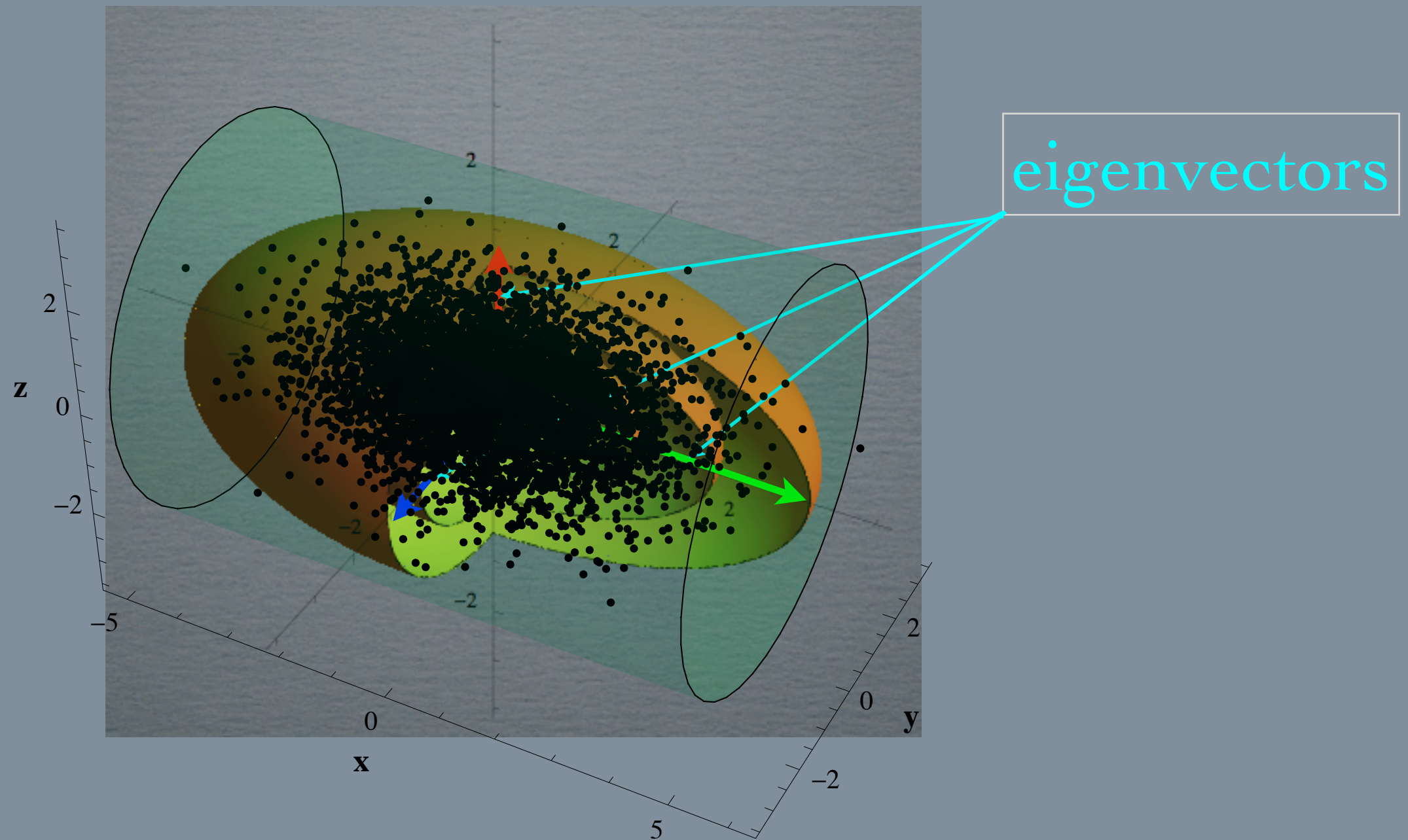
$$\mathbf{q} = (a, b) \quad \text{columns here}$$

$$\mathbf{r} = \begin{pmatrix} x \\ y \end{pmatrix} \quad \text{must equal rows here}$$

$$\mathbf{qr} = (\textcircled{a}, \textcircled{b}) \begin{pmatrix} \textcircled{x} \\ \textcircled{y} \end{pmatrix} = \textcircled{a}\textcircled{x} + \textcircled{b}\textcircled{y}$$

Slides from previous lecture to add here
START

DIFFUSION ANISOTROPY IN 3D

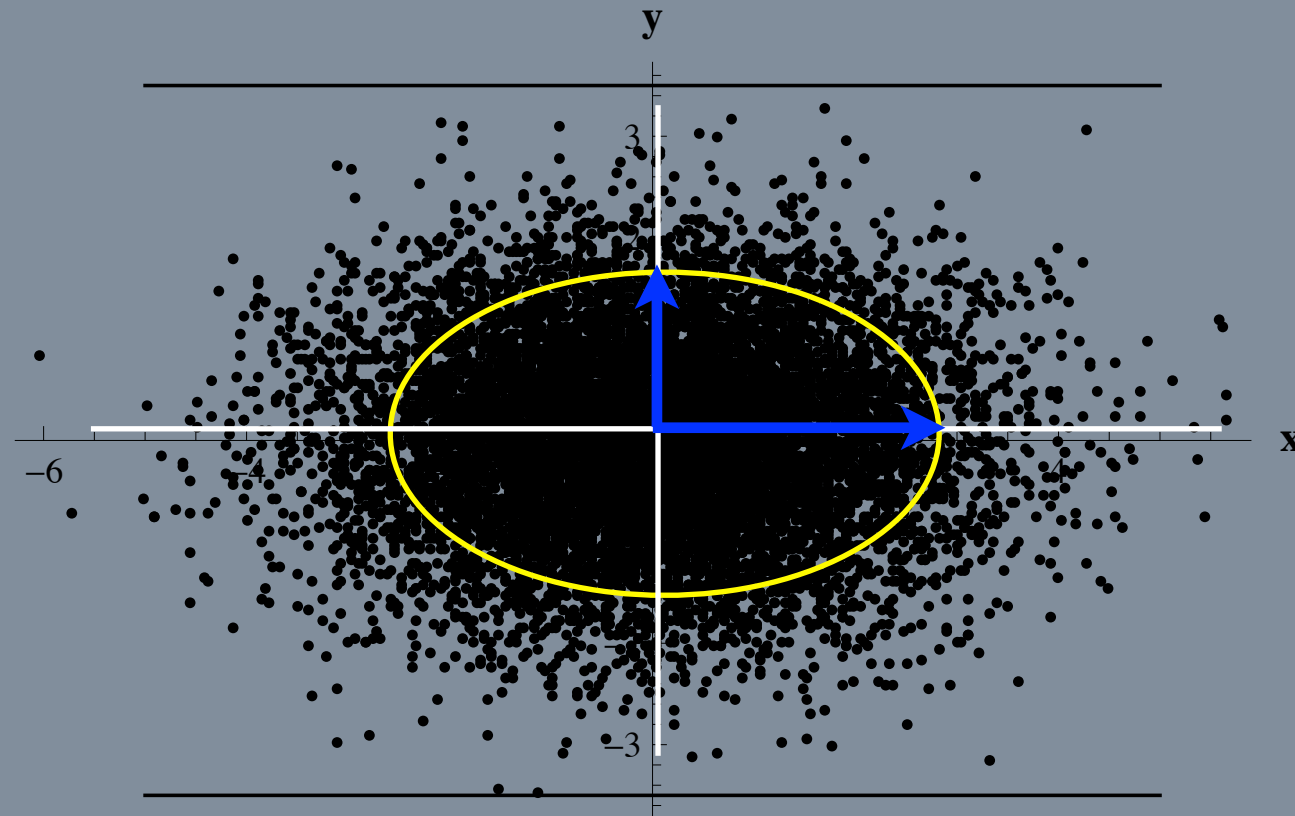


probability contours in 3D

THE 2D GAUSSIAN DISTRIBUTION

$$P(\mathbf{r}|\mathbf{r}_0, \tau) \sim N(\mathbf{r}_0, \mathbf{\Sigma})$$

$$\mathbf{r} = \{x, y\}$$



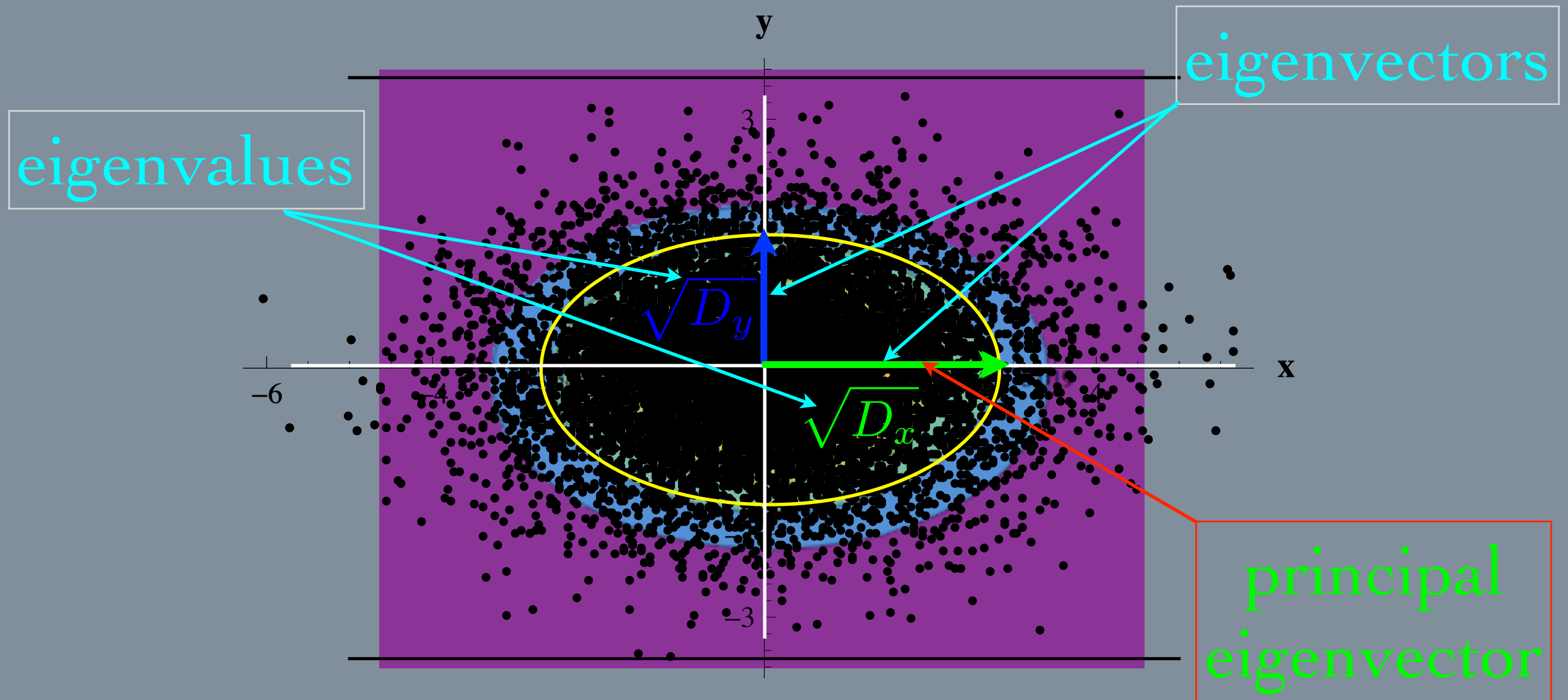
Covariance matrix

Diffusion Tensor

$$\mathbf{\Sigma} = \begin{pmatrix} \sigma_x^2 & 0 \\ 0 & \sigma_y^2 \end{pmatrix} = 4\tau \begin{pmatrix} D_x & 0 \\ 0 & D_y \end{pmatrix}$$

$\{D_x, D_y\}$ are the *principal diffusivities*

ANISOTROPIC GAUSSIAN DIFFUSION



1. The relative dimensions of the contours tells us about local structure
2. The orientation of the largest dimension is related to the orientation of the structure

Covariance Matrix

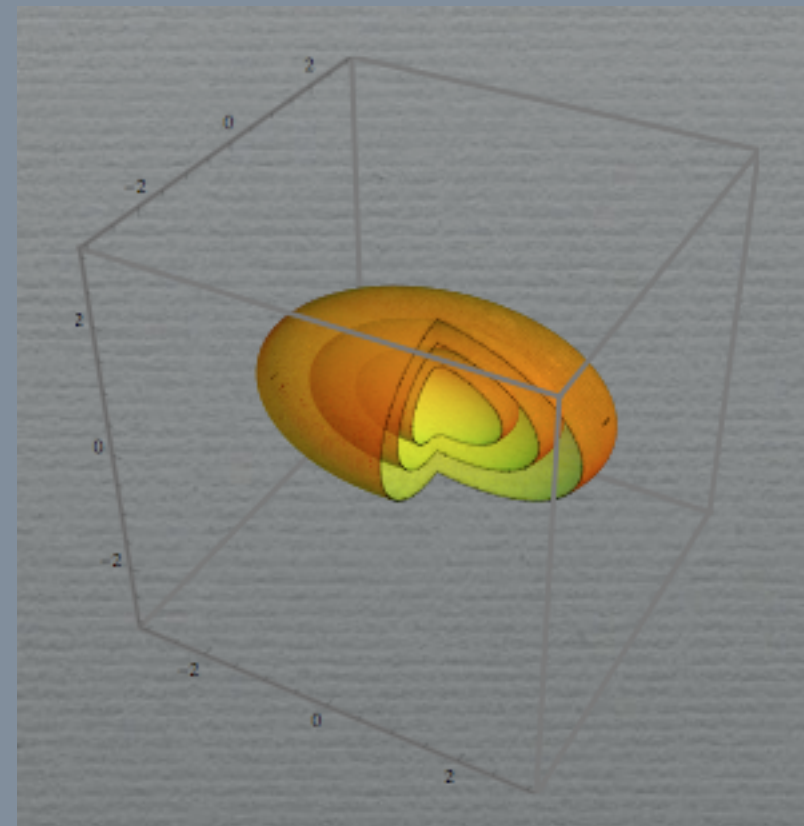
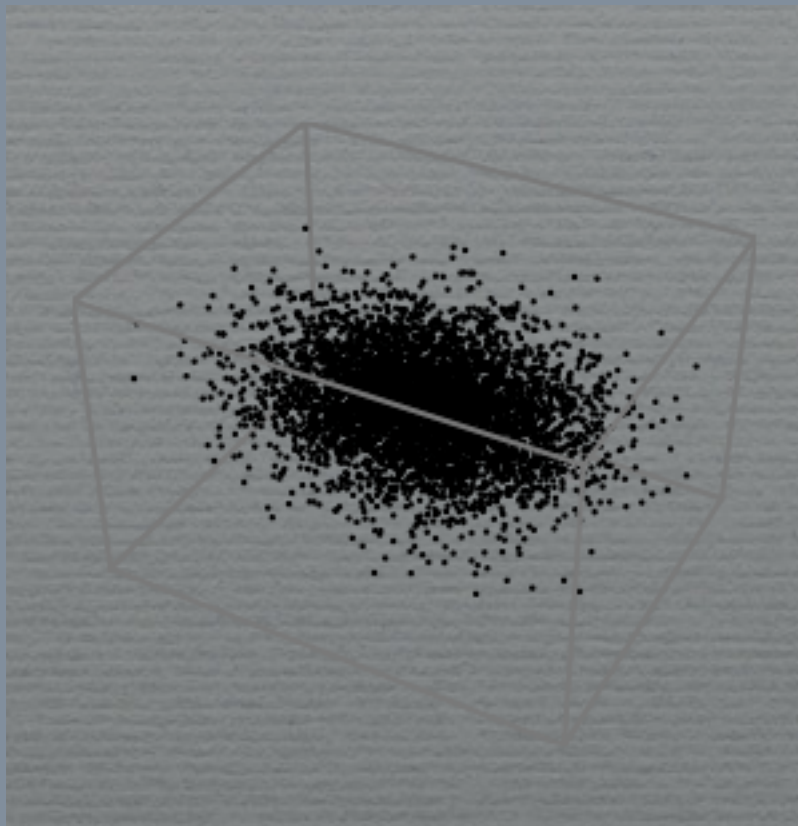
$$\Sigma = \begin{pmatrix} \sigma_{xx}^2 & \sigma_{xy}^2 \\ \sigma_{yx}^2 & \sigma_{yy}^2 \end{pmatrix} = 4\tau \underbrace{\begin{pmatrix} D_{xx} & D_{xy} \\ D_{yx} & D_{yy} \end{pmatrix}}_D$$

diffusion tensor

$$D = \begin{pmatrix} D_{xx} & D_{xy} \\ D_{yx} & D_{yy} \end{pmatrix}$$

3D Gaussian Distribution

$$P(\mathbf{x}, t) = \frac{1}{(2\pi)^{3/2} |\Sigma|^{1/2}} \exp \left[-\frac{1}{2} (\mathbf{x} - \mu)^t \Sigma^{-1} (\mathbf{x} - \mu) \right]$$



Covariance Matrix

$$\Sigma = \begin{pmatrix} \sigma_{xx}^2 & \sigma_{xy}^2 & \sigma_{xz}^2 \\ \sigma_{yx}^2 & \sigma_{yy}^2 & \sigma_{yz}^2 \\ \sigma_{zx}^2 & \sigma_{zy}^2 & \sigma_{zz}^2 \end{pmatrix} = 6\tau \underbrace{\begin{pmatrix} D_{xx} & D_{xy} & D_{xz} \\ D_{yx} & D_{yy} & D_{yz} \\ D_{zx} & D_{zy} & D_{zz} \end{pmatrix}}_D$$

diffusion tensor
symmetric and real

$$D = \begin{pmatrix} D_{xx} & D_{xy} & D_{xz} \\ D_{yx} & D_{yy} & D_{yz} \\ D_{zx} & D_{zy} & D_{zz} \end{pmatrix}$$

The Inverse Covariance Matrix

$$\mathbf{\Lambda} = \begin{pmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{pmatrix} \quad \mathbf{\Lambda}^{-1} = \begin{pmatrix} \lambda_1^{-1} & 0 \\ 0 & \lambda_2^{-1} \end{pmatrix}$$

$$\text{where} \quad \lambda_i = \frac{1}{\sigma_i^2} \quad \text{and} \quad \sigma_i^2 = 2D_i\tau$$

eigenvalues of the inverse
covariance matrix are

THE MEANING OF THE EIGENSYSTEM

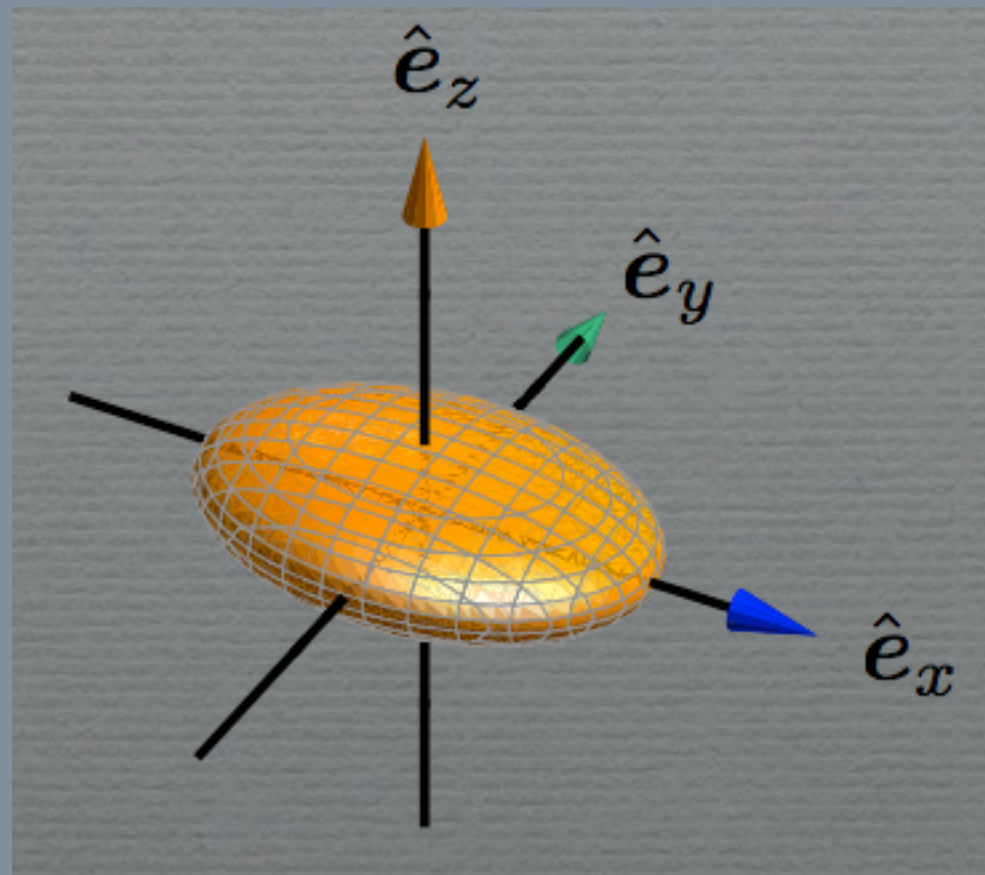
The eigenvectors of $\mathbf{D}$ are the unique directions along which the molecular displacements are uncorrelated

The eigenvalues $\{D_x, D_y\}$ are the principal diffusivities along these directions

THE 3D GAUSSIAN DISTRIBUTION:

$$P(\mathbf{r}|\mathbf{r}_0, \tau) \sim N(\mathbf{r}_0, \mathbf{\Sigma})$$

$$\mathbf{r} = \{x, y, z\}$$



Covariance matrix

Diffusion Tensor

$$\mathbf{\Sigma} = \begin{pmatrix} \sigma_{xx}^2 & 0 & 0 \\ 0 & \sigma_{yy}^2 & 0 \\ 0 & 0 & \sigma_{zz}^2 \end{pmatrix} = 6\tau \begin{pmatrix} D_x & 0 & 0 \\ 0 & D_y & 0 \\ 0 & 0 & D_z \end{pmatrix}$$

THE DIFFUSION TENSOR

The three eigenvectors of $\mathbf{D}$

$$\{\vec{e}_1, \vec{e}_2, \vec{e}_3\}$$

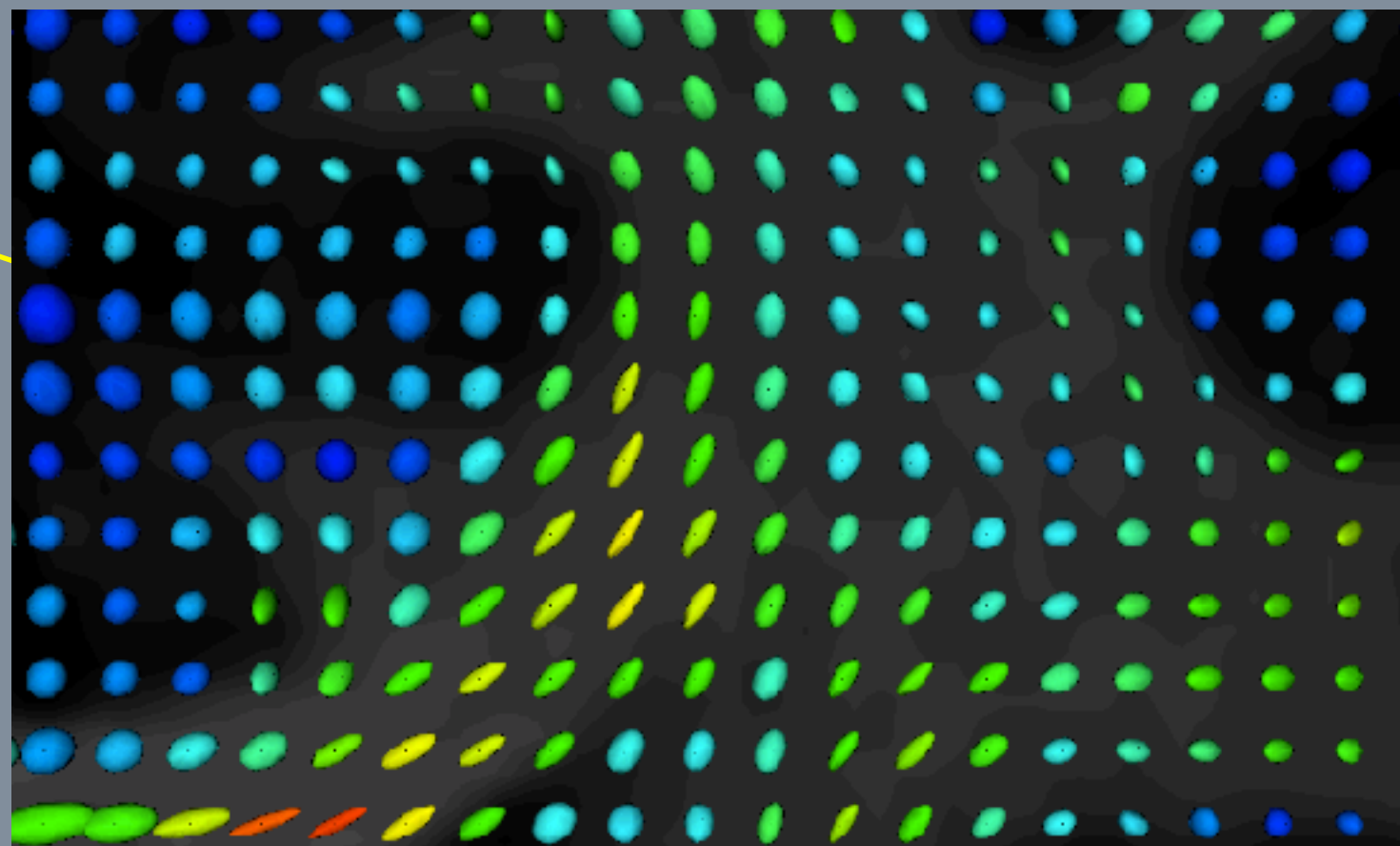
are the three unique directions along which the molecular displacements are uncorrelated

The three eigenvalues of $\mathbf{D}$

$$\{D_x, D_y, D_z\}$$

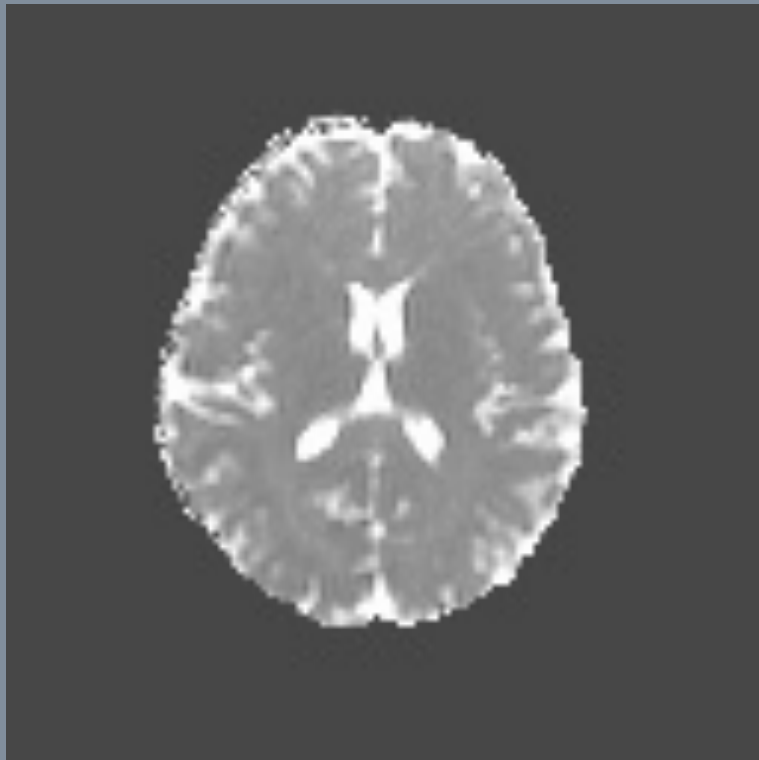
are the ADC values along these directions

RECONSTRUCTED DIFFUSION TENSOR IS A VERY SIMPLIFIED MODEL!

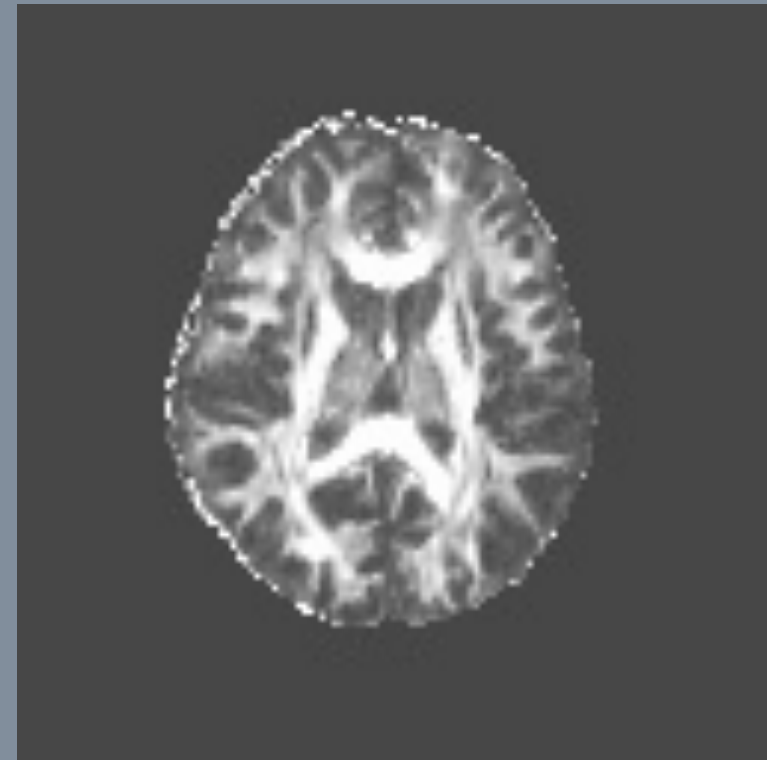


diffusion ellipsoids

MEASURED DTI PARAMETERS
ARE AVERAGING OVER A LOT OF PHYSIOLOGY!



Mean diffusion



Diffusion anisotropy

Slides from previous lecture to add here
END