

# CS 565 Computer Vision

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Lectures 6 and 7: Fourier Transform

# Disclaimer

- Any unreferenced image is taken from the following web-page
  - <http://betterexplained.com/articles/an-interactive-guide-to-the-fourier-transform/>

# Note

- If a hammer is the only tool you have, you will look at every problem as a nail.
- The more tools you have, the more problems you can tackle.
- Our foray into the “Fourier world” is an attempt to gather as many tools as we can.

# Fourier Transform

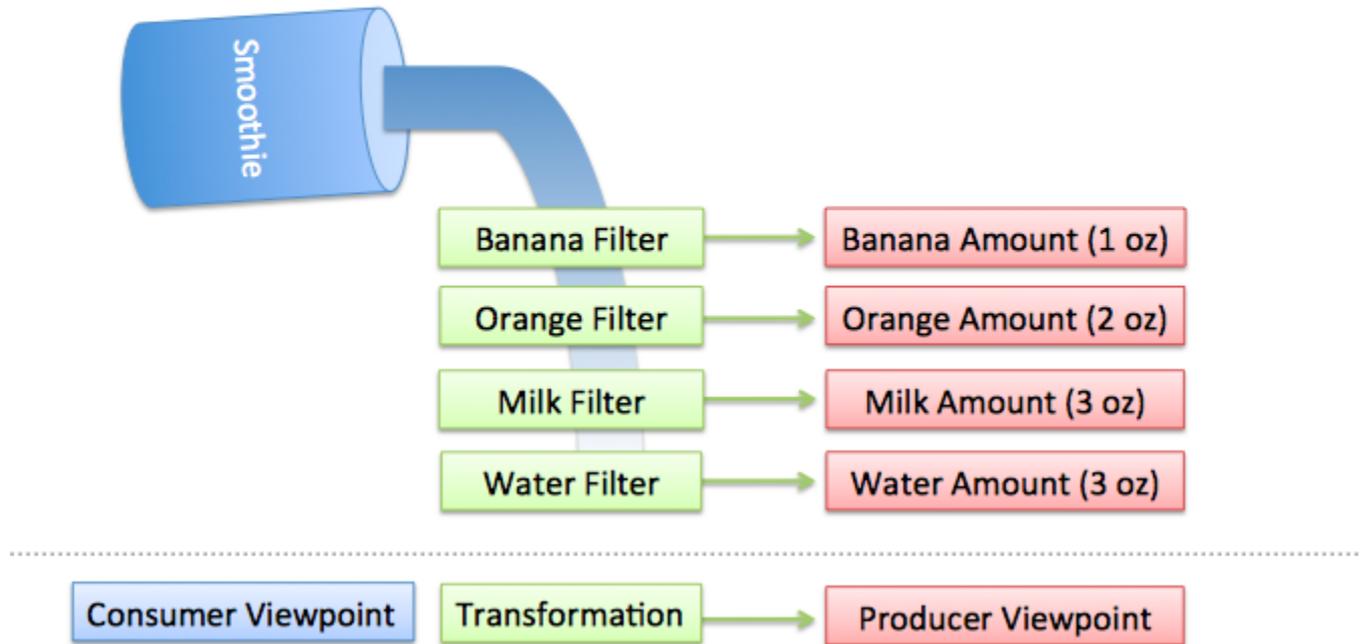
- One of the deepest mathematical insights.
- For any signal, it extracts its “ingredients”.
  - This is a **very powerful** idea.
  - Given an observation, it gives you the causes.
  - Given an image, it gives you its constituents.
- Understanding the Fourier Transform requires some of the **most beautiful mathematics** ever invented.

# Fourier Transform

- The mathematics can become (more than) a little bit overwhelming.
- So we'll break it down into smaller, easier steps.

# Fourier Transform – An Analogy

## Smoothie to Recipe



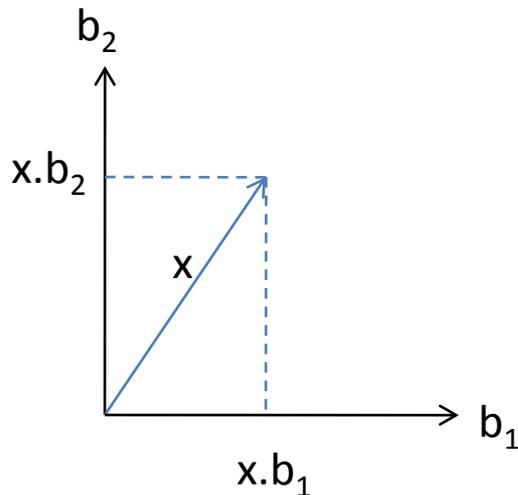
Source: <http://betterexplained.com/articles/an-interactive-guide-to-the-fourier-transform/>

# Fourier Transform

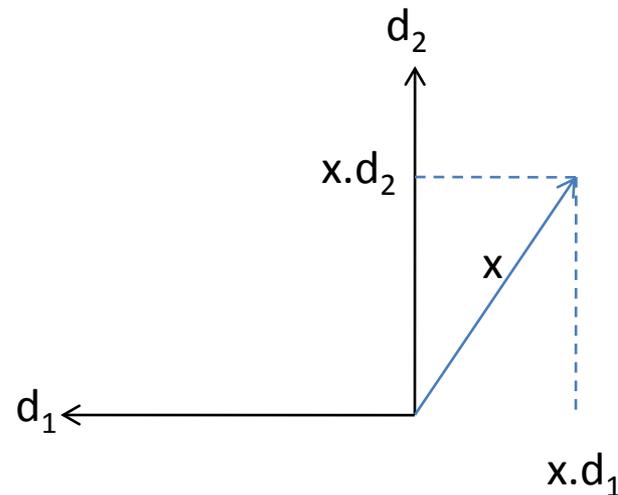
- We start with some pre-requisite mathematics.
  - Remember, math is not magic!
  - You **can** understand it if you take the correct perspective.

# Projection

- A 2D vector  $x$  can be represented in an orthonormal basis  $\{b_1, b_2\}$  by the formula  $x = (x \cdot b_1)b_1 + (x \cdot b_2)b_2$ .
  - Coefficient for basis vector  $k$  is the projection  $(x \cdot b_k)$ .



$$x = (x \cdot b_1)b_1 + (x \cdot b_2)b_2$$



$$x = (x \cdot d_1)d_1 + (x \cdot d_2)d_2$$

# Mathematical Background

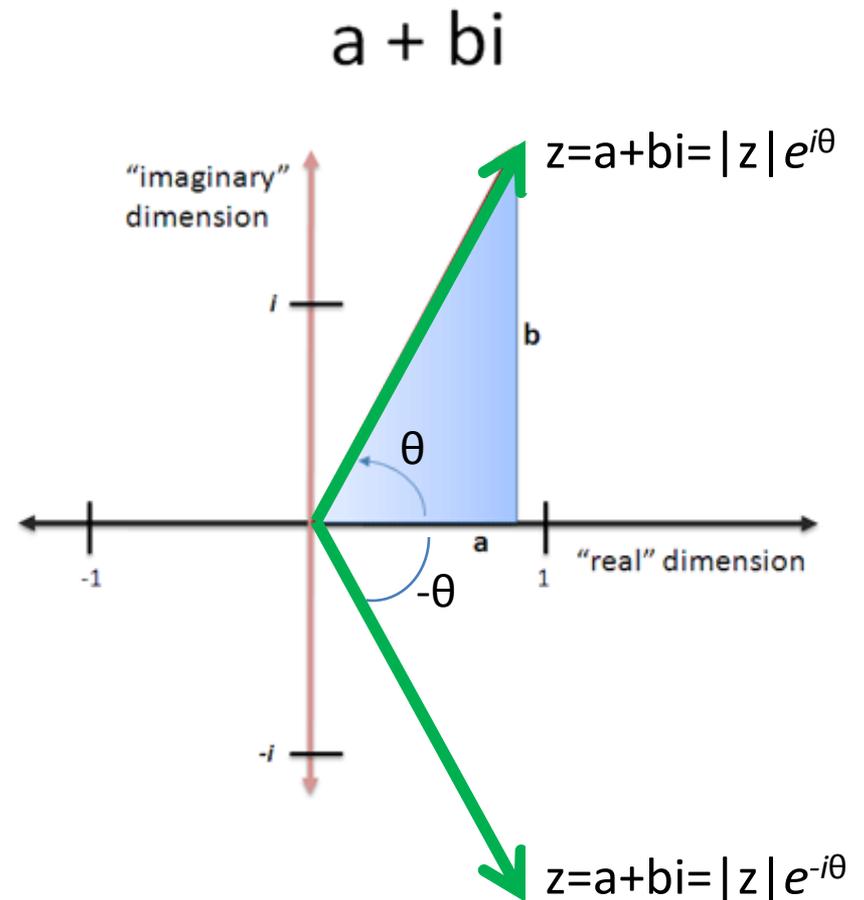
- $\pi$ 
  - circumference/diameter of **any** circle.
  - universal constant ( $\pi = 3.14159265\dots$ )
- $e$ 
  - Euler's number ( $e = 2.71828182\dots$ )
- $i$ 
  - non-existent, imaginary number (what!!!!)
  - makes analysis and computations easier ( $i^2 = -1$ )

# Complex Numbers

- Real numbers are represented by  $\mathbb{R}^1$ .
- We can write any real number  $x$  as  $x+0i$ .
- Therefore,  $\mathbb{R}^1$  is contained within the space of complex numbers  $\mathbb{C}^1$ .
  - Complex numbers  $z$  have a real part  $\text{Re}(z)$  and an imaginary part  $\text{Im}(z)$ .
- Basis vector for  $\mathbb{R}^1$  is the scalar 1.
- Basis vectors for  $\mathbb{C}^1$  are  $\{(1,0),(0,i)\}$ .

# Complex Numbers

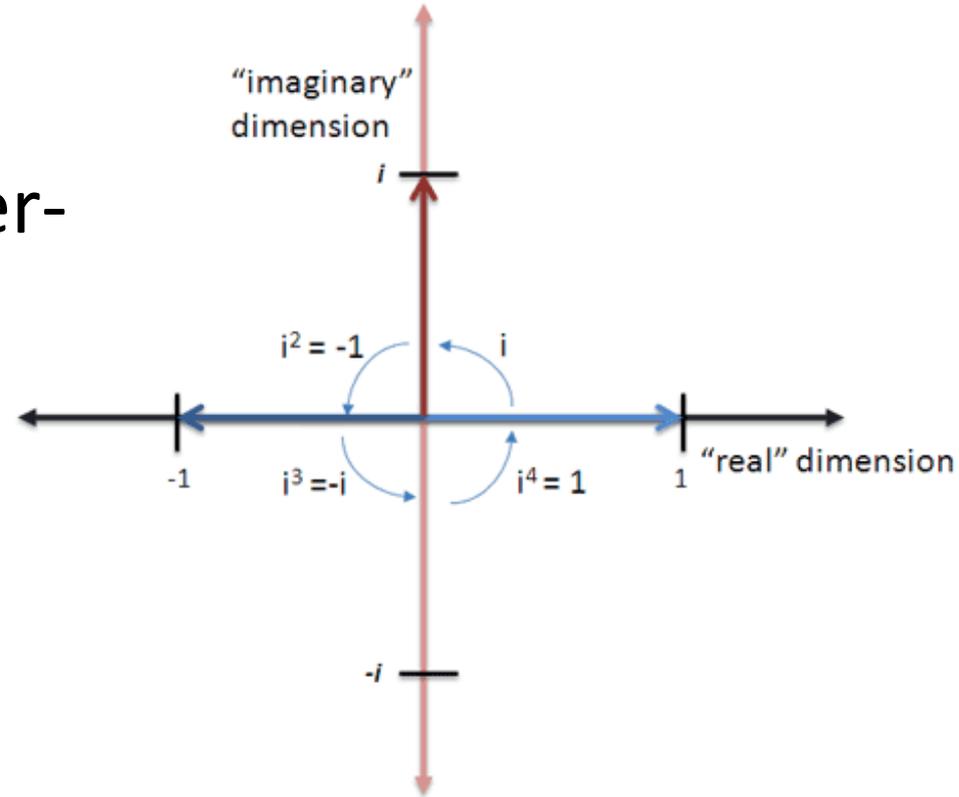
- Norm (magnitude, modulus) of  $z$  is given by  $|z| = \sqrt{a^2 + b^2}$ .
- Phase is the angle  $\theta = \arctan(b/a)$ .
- A complex number can also be represented in Polar form  $z = a + bi = |z| e^{i\theta}$ .
- Conjugate of  $z$  is given by  $\text{conj}(z) = a - bi = |z| e^{-i\theta}$ .
- **HW: Compute the values of  $\sqrt{z \cdot z}$  and  $\sqrt{z \cdot \text{conj}(z)}$ . Which one yields the norm of  $z$ ?**



# Multiplication by $i$ Represents $90^\circ$ Rotation in $\mathbb{C}$

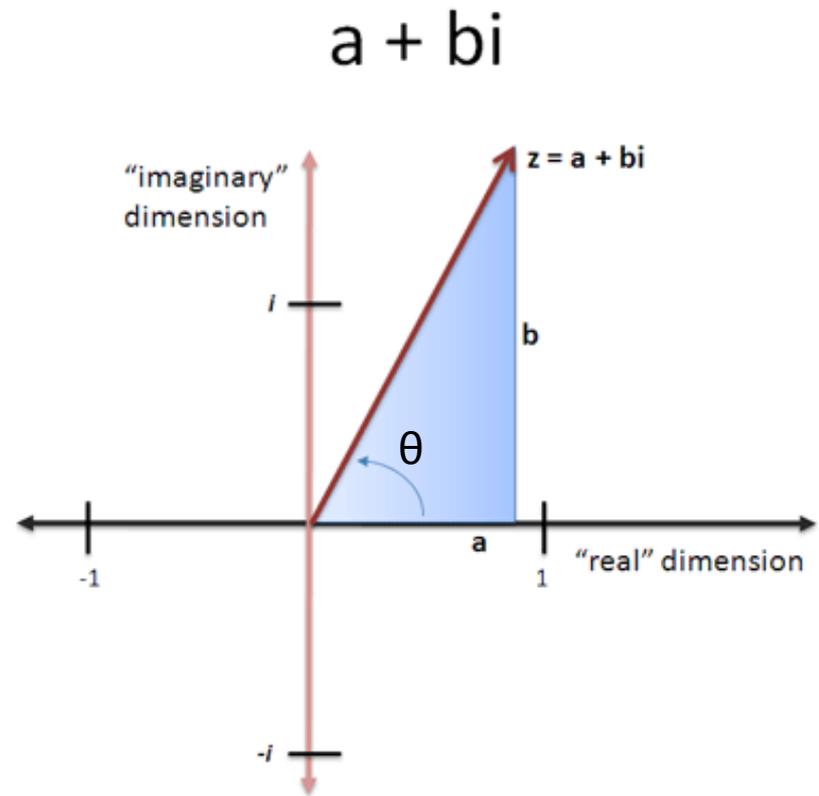
- Multiplication by  $i$  is a **rotation** by  $90^\circ$  counter-clockwise in  $\mathbb{C}$ .

- $1 * i = i$
- $1 * i * i = -1$
- $1 * i * i * i = -i$
- $1 * i * i * i * i = 1$



# Multiplication by Complex Number Represents Rotation in $\mathbb{C}$

- Multiplication by any complex number  $z = a+bi$  causes **rotation** by its **angle**  $\theta = \arctan(b/a)$



# The Bigger Picture

- The complex space  $\mathbb{C}$  is just a generalization of the real space  $\mathbb{R}$  where rotation amounts to multiplication.
- We don't care about  $\mathbb{C}$  itself but we care about the fact that in  $\mathbb{C}$  complicated rotations can be represented as simply as multiplications.
  - We don't care whether –ve numbers actually exist or not, we care that they make calculations of profit/loss or debit/credit easier.

# Euler

- One of the greatest mathematicians ever.
- Fundamental contributions in calculus, graph theory, optics, fluid dynamics, mechanics, astronomy and even music theory.
- Almost totally blind for the last 20 years of his life.
  - Yet did the most productive work during this time.

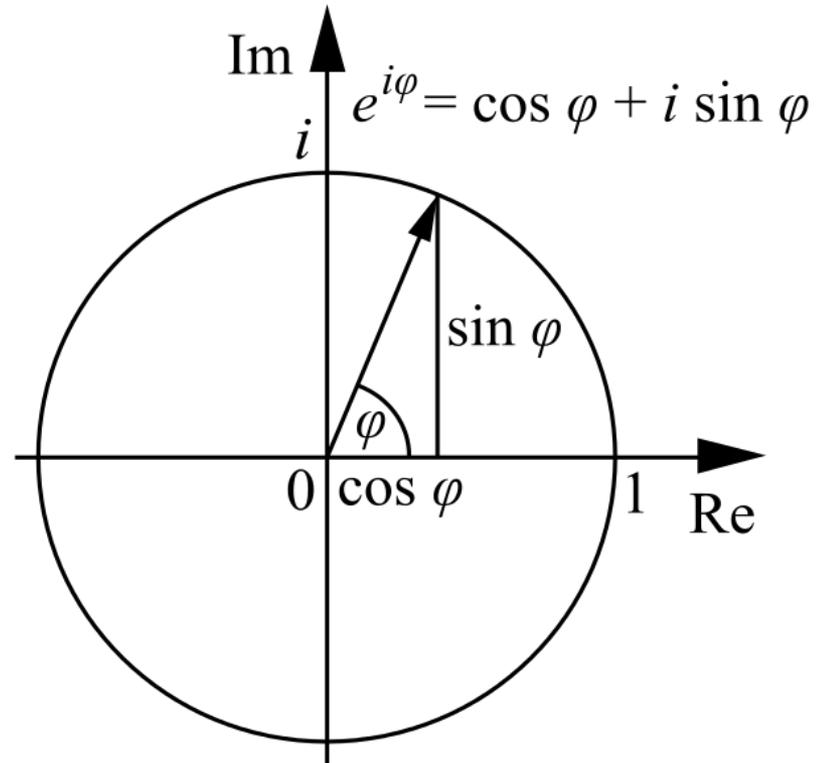


Source:  
[http://en.wikipedia.org/wiki/Leonhard\\_Euler](http://en.wikipedia.org/wiki/Leonhard_Euler)

# Euler's Formula

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

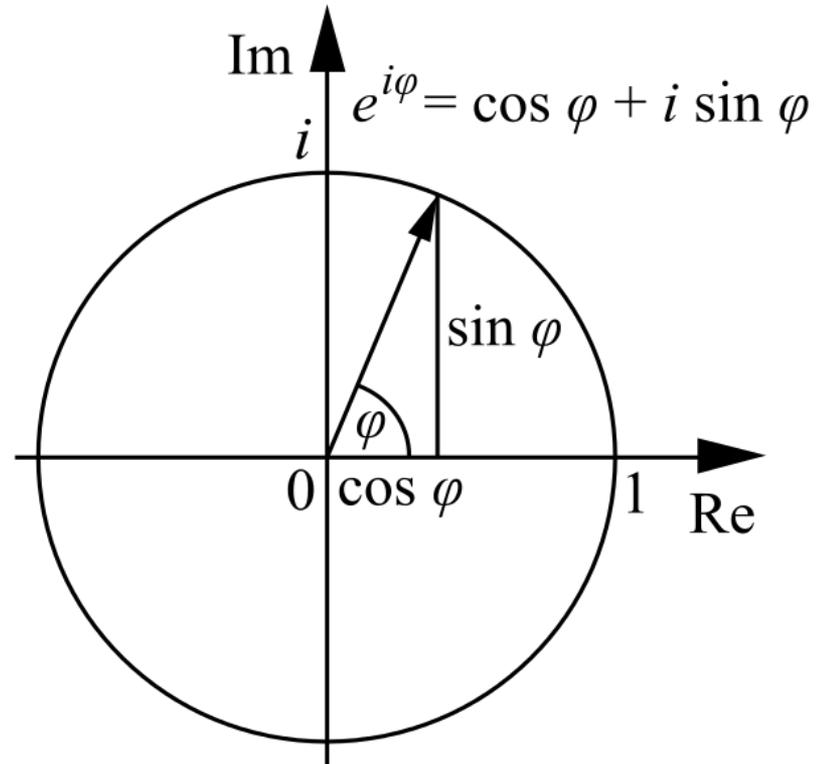
- **Mathematics does not get more beautiful than this equation.**
- What you can describe using sinusoids, you can describe using the numbers  $e=2.71828182\dots$  and  $i=\sqrt{-1}$



In 1988, readers of the [\*Mathematical Intelligencer\*](#) voted it "the Most Beautiful Mathematical Formula Ever". In total, Euler was responsible for three of the top five formulae in that poll.

# Euler's Formula

- What can we describe using  $\cos(\theta)$  and  $\sin(\theta)$ ?
  - Positions on a circle.
- The formula says that that position is  $2.7182818284^{\theta\sqrt{-1}}$  or simply  $e^{i\theta}$ .



In Matlab:

```
>> [exp(sqrt(-1)*pi/4); cos(pi/4)+i*sin(pi/4)]
```

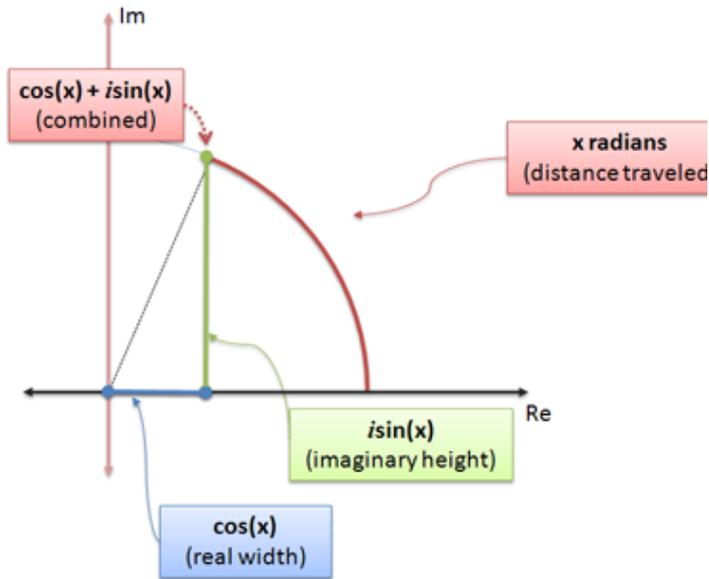
```
ans =
```

```
0.7071 + 0.7071i
```

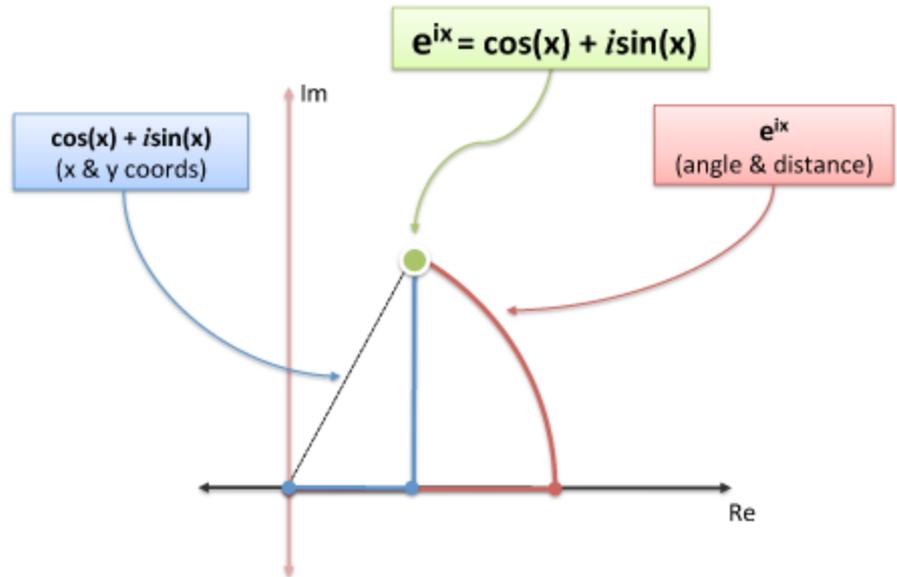
```
0.7071 + 0.7071i
```

# Euler's Formula

## Traversing A Circle



## Two Paths, Same Result



# Euler's Formula – The Bigger Picture

- **Describes circular motion.**
- Two ways to describe motion
  - Cartesian: Go 3 units east and 4 units north
  - Polar: Go 5 units at an angle of 71.56 degrees
- Depending on the problem, polar or Cartesian coordinates are more useful.
- **Euler's formula lets us convert between polar and Cartesian representation to use the best tool for the job.**

# The link between Euler's Formula and the Fourier Transform

- **Fourier's claim: Any signal can be made from circular motion.**
- Euler's formula generates all circular motions.
- So Euler's formula is the tool that the Fourier Transform needs to decompose signals into circular motions.

# Fourier Transform

- In the Fourier Transform, we factorise the angular distance  $\theta$  into angular speed  $\omega$  and time  $t$ .
  - $\theta$  is angular distance along the circle ( $0-2\pi$ ).
- Since  $\theta = \omega t$ , we can write  $e^{i\theta} = e^{i\omega t}$ 
  - So  $e^{i\omega t}$  determines how far we have moved along the circle in time  $t$  travelling at speed  $\omega$ .
- By varying  $\omega$  and  $t$ , we can compute how far a circular motion with speed  $\omega$  will be at time  $t$ .

# Fourier Transform

- Angular speed  $\omega = 2\pi f$  where  $f$  is the frequency in cycles per unit time. (**HW: Verify this. Hint: Just look at the definitions and/or units of  $\omega$  and  $f$ .**)
- So we can write  $e^{i\theta} = e^{i\omega t} = e^{i2\pi f t}$ 
  - So  $e^{i2\pi f t}$  determines how far we have moved along the circle in time  $t$  travelling with a frequency  $f$ .
- **By varying  $f$  and  $t$ , we can compute how far a circular motion with frequency  $f$  will be at time  $t$ .**

# Fourier Transform

- Let  $\mathbf{x} \in \mathbb{R}^N$ . That is  $\mathbf{x} = (x_0, x_1, \dots, x_{N-1})^T$ . Obviously,  $\mathbf{x} \in \mathbb{C}^N$  too.
- Assume signal repeats after every 1 second.
- Divide this 1 second into  $N$  slices.
- Assume value  $x_n$  occurs at time  $t=n/N$  seconds.
- Position on the circle at time  $t=n/N$  is given by  $e^{i\theta} = e^{i\omega t} = e^{i2\pi f t} = e^{i2\pi f n/N}$
- This gives us  $N$  positions along a circular motion with frequency  $f$ .
  - Let us denote them by the vector  $\mathbf{u}_f = (e^{i2\pi f 0/N}, e^{i2\pi f 1/N}, \dots, e^{i2\pi f (N-1)/N})^T$ .
- Similarity of  $\mathbf{x}$  and  $\mathbf{u}_f$  can be computed by ...
  - Inner-product  $\mathbf{x}^T \mathbf{u}_f$ .

# Fourier Transform – The Bigger Picture

- Project signal  $\mathbf{x}$  onto circular motions  $\mathbf{u}_f$  of different frequencies  $f = 0, 1, \dots, N-1$ .
  - Fourier coefficient for frequency  $f$  is  $X_f = \mathbf{x}^T \mathbf{u}_f$

# Fourier Transform – Projection onto Circular Motion

- For the Fourier transform, the  $N$  dimensional signal vector  $\mathbf{x}$  is **projected** onto the circular basis vectors  $\mathbf{e}^{i2\pi f}$ .
  - Coefficient for basis vector with frequency  $f$  is the projection ( $\mathbf{x}^T \mathbf{e}^{-i2\pi f}$ ).

$$\begin{aligned} X_f &= \mathbf{x}^T \mathbf{e}^{-i2\pi f} \\ &= x_0 e^{-i2\pi f \frac{0}{N}} + \dots + x_{N-1} e^{-i2\pi f \frac{N-1}{N}} \end{aligned}$$

- Do you notice something strange in the projection?

# Fourier Transform – Projection onto Circular Motion

- Why the negative sign in the exponent?
- In order to measure lengths in any number space, a norm must be defined such that  $|x| = \sqrt{x \cdot x} = \text{length of vector } x$ .
- In the space of Complex numbers, inner product is defined as  $x \cdot y = x^* \text{conj}(y)$  where  $\text{conj}(y) = \text{Re}(y) - \text{Im}(y)i = |y|e^{-i\theta}$ .
- **HW: For a complex vector  $f = (f_1, \dots, f_N)$ , compute  $f \cdot f$  and  $f \cdot \text{conj}(f)$ . Which one yields the squared norm of  $f$  (given by  $|f|^2 = |f_1|^2 + \dots + |f_N|^2$ )?**
- The negative sign signifies conjugation of  $e^{i2\pi f}$ . So that the norm can be properly defined in Complex space.

# Fourier Transform

$$X_f = \frac{1}{\sqrt{N}} \sum_{n=0}^{N-1} x_n e^{-i2\pi f n/N} = \frac{1}{\sqrt{N}} \mathbf{x}^T e^{-i2\pi f}$$

Decompose the signal into its constituent frequencies from  $f=0$  to  $N-1$ .

# Inverse Fourier Transform

$$x_n = \frac{1}{\sqrt{N}} \sum_{f=0}^{N-1} X_f e^{i2\pi f n/N}$$

Synthesize the signal from its constituent frequencies.

# Orthonormality of the Fourier Basis

- The basis vectors for different frequencies  $f$  are orthonormal.

$$\begin{array}{c} \text{Orthogonality Condition} \\ \left[ \begin{array}{c} e^{i2\pi f_p \frac{0}{N}} \\ \vdots \\ e^{i2\pi f_p \frac{N-1}{N}} \end{array} \right] \cdot \left[ \begin{array}{c} e^{i2\pi f_q \frac{0}{N}} \\ \vdots \\ e^{i2\pi f_q \frac{N-1}{N}} \end{array} \right] = 0 \quad \forall p \neq q \end{array}$$

$$\begin{array}{c} \text{Normality Condition} \\ \left[ \begin{array}{c} e^{i2\pi f_p \frac{0}{N}} \\ \vdots \\ e^{i2\pi f_p \frac{N-1}{N}} \end{array} \right] \cdot \left[ \begin{array}{c} e^{i2\pi f_p \frac{0}{N}} \\ \vdots \\ e^{i2\pi f_p \frac{N-1}{N}} \end{array} \right] = 1 \quad \forall p \end{array}$$

- So the different frequencies do not interfere with each other in representing the signal.
- HW: Prove orthonormality of Fourier basis.**

# Frequency Domain Filtering Pipeline



# Frequency Domain Low-Pass Filtering (Smoothing)



a b c

**FIGURE 4.50** (a) Original image ( $784 \times 732$  pixels). (b) Result of filtering using a GLPF with  $D_0 = 100$ . (c) Result of filtering using a GLPF with  $D_0 = 80$ . Note the reduction in fine skin lines in the magnified sections in (b) and (c).

# Frequency Domain High-Pass Filtering (Sharpening)



# Frequency Domain Band-Pass Filtering

