

Lecture 8: Recap:

Definition: (Discrete Fourier Transform) Given  $f_0, f_1, \dots, f_{n-1} \in \mathbb{C}$ , then the discrete Fourier Transform (DFT) is defined as:

$$\vec{c} = \begin{pmatrix} c_0 \\ c_1 \\ \vdots \\ c_{n-1} \end{pmatrix} \in \mathbb{C}^n \quad \text{where} \quad c_k = \frac{1}{n} \sum_{j=0}^{n-1} f_j e^{-i\left(\frac{2jk\pi}{n}\right)} \quad \text{for } k=0, 1, 2, \dots, n-1$$

The inverse discrete Fourier Transform recovers the original signal:

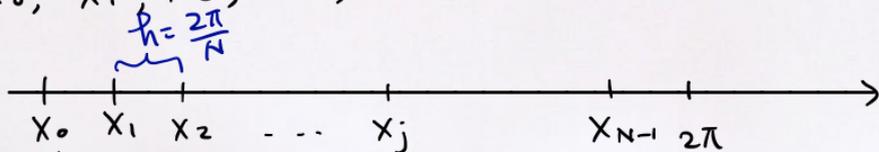
$$f_j = \sum_{k=0}^{n-1} c_k e^{i\left(\frac{2jk\pi}{n}\right)} \quad \text{for } j=0, 1, 2, \dots, n-1$$

## Recall:

Consider:  $\frac{d^2 u}{dx^2} = f$  for  $x \in [0, 2\pi]$  with periodic boundary condition.  
 $u(0) = u(2\pi)$

Suppose  $f$  is measured only at  $N$  discrete points =

$$x_0, x_1, x_2, \dots, x_{N-1}$$



Let  $\vec{f} = \begin{pmatrix} f(x_0) \\ f(x_1) \\ \vdots \\ f(x_{N-1}) \end{pmatrix} \stackrel{\text{def}}{=} \begin{pmatrix} f_0 \\ f_1 \\ \vdots \\ f_{N-1} \end{pmatrix} \in \mathbb{R}^N$  and  $\vec{u} = \begin{pmatrix} u(x_0) \\ u(x_1) \\ \vdots \\ u(x_{N-1}) \end{pmatrix} = \begin{pmatrix} u_0 \\ u_1 \\ \vdots \\ u_{N-1} \end{pmatrix} \in \mathbb{R}^N$  (unknown)



Claim:  $\{e^{ikx}\}_{k=0}^{N-1}$  is a basis of  $\mathbb{C}^N$  (consisting of eigenvectors)

Pf:  $\begin{pmatrix} | & | & & | \\ e^{i0x} & e^{i1x} & \dots & e^{i(N-1)x} \\ | & | & & | \end{pmatrix} = A\omega = \begin{pmatrix} 1 & 1 & \dots & 1 \\ \vdots & \omega & \dots & \omega^{N-1} \\ \vdots & \vdots & \dots & \vdots \\ 1 & \omega^{N-1} & \dots & \omega^{(N-1)^2} \end{pmatrix}; \omega = e^{i\frac{2\pi}{N}}$

Claim:  $\text{Rank}(\tilde{D}) = N-1$  and null space of  $D$  is =  
 $N(\tilde{D}) = \text{span}\left\{\begin{pmatrix} 1 \\ \vdots \\ 1 \end{pmatrix}\right\}$

Pf:  $D^2$  has  $N-1$  distinct non-zero eigenvalues.  
 $\therefore \text{Rank}(\tilde{D}) = N-1$ .

$N(\tilde{D}) =$  eigenspace associated to zero eigenvalue  $= \lambda_0$ .  
The only eigenvector of  $\tilde{D}$  with eigenvalue  $= 0$  is  $\vec{e}^{i0x} = \begin{pmatrix} 1 \\ \vdots \\ 1 \end{pmatrix}$ .  
 $\therefore N(\tilde{D}) = \text{span}\left\{\begin{pmatrix} 1 \\ \vdots \\ 1 \end{pmatrix}\right\}$ .

Claim: If  $\vec{u}_1$  and  $\vec{u}_2$  are both solutions of  $\tilde{D} \vec{u} = \vec{f}$ , then:  
$$\vec{u}_1 = \vec{u}_2 + c \begin{pmatrix} 1 \\ \vdots \\ 1 \end{pmatrix} \text{ for some constant } c.$$

Proof:  $\tilde{D} \vec{u}_1 = \vec{f}$  ;  $\tilde{D} \vec{u}_2 = \vec{f}$   $\therefore D(\vec{u}_1 - \vec{u}_2) = 0$

$\therefore \vec{u}_1 - \vec{u}_2 \in N(\tilde{D}) = \text{span} \left\{ \begin{pmatrix} 1 \\ \vdots \\ 1 \end{pmatrix} \right\}.$

Thus,  $\vec{u}_1 = \vec{u}_2 + c \begin{pmatrix} 1 \\ \vdots \\ 1 \end{pmatrix}.$  for some  $c.$

## Numerical Spectral method

Since  $\{ \overrightarrow{e^{ikx}} \}_{k=0}^{N-1}$  is a basis. We can write:

$$\vec{u} = \sum_{k=0}^{N-1} \hat{u}_k \overrightarrow{e^{ikx}} \quad \text{and} \quad \vec{f} = \sum_{k=0}^{N-1} \hat{f}_k \overrightarrow{e^{ikx}}$$

↑  
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In other words, for each  $j$ ,  $f_j = f(x_j) = \sum_{k=0}^{N-1} \hat{f}_k (e^{ikx})_j$   
 $= \sum_{k=0}^{N-1} \hat{f}_k e^{ikx_j}$   
(DFT!)

∴  $\hat{f}_k$  can be determined by DFT.

To solve  $\frac{d^2 u}{dx^2} = f$ , we approximate it by

$$\tilde{D} \vec{u} = \vec{f}.$$

Now,  $\tilde{D}\vec{u} = \vec{f}$  becomes:

$$\tilde{D} \left( \sum_{k=0}^{N-1} \hat{u}_k \vec{e}^{ikx} \right) = \sum_{k=0}^{N-1} \hat{f}_k \vec{e}^{ikx}$$

$$\Leftrightarrow \sum_{k=0}^{N-1} \hat{u}_k \tilde{D} \vec{e}^{ikx} = \sum_{k=0}^{N-1} \hat{f}_k \vec{e}^{ikx}$$

$\parallel$   
 $(-\lambda_k^2) \vec{e}^{ikx}$

$$\Leftrightarrow \sum_{k=0}^{N-1} \hat{u}_k (-\lambda_k^2) \vec{e}^{ikx} = \sum_{k=0}^{N-1} \hat{f}_k \vec{e}^{ikx}$$

Comparing coefficients, we get

$$\underbrace{-\lambda_k^2}_{\text{known}} \underbrace{\hat{u}_k}_{\text{unknown}} = \underbrace{\hat{f}_k}_{\text{known}} \quad \text{for } k=0, 1, 2, \dots, N-1$$

(algebraic equation)