Review

- Basic graph algorithms - Reachability - BFS/DFS/applications

Greedy algorithms - Shortest paths, MST, clustering, Huffman - Divide and conquer interval scheduling under

integer mult, matrix mult; DFT, closest pair, into to merge sort, polynomial mult.

DFT - review.

. Given $\vec{\alpha} = (a_0, \dots, a_{n-1})$; we want DFT($\vec{\alpha}$).

$$\begin{bmatrix} b_0 \\ b_{n1} \end{bmatrix} = i \begin{bmatrix} a_0 \\ b_{n1} \end{bmatrix} \begin{bmatrix} a_0 \\ a_{n-1} \end{bmatrix}$$

w is promotive not not of whote.

w=1 and + ke[1,n-1], w=1.

 $\subseteq_{\mathbf{K}^{\circ}}$ (Olo, α_{1}) w is principle and noof of ningth. w=-1

$$\begin{bmatrix} \mathcal{W}^{000} & \mathcal{W}^{001} \\ \mathcal{W}^{000} & \mathcal{W}^{001} \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}.$$

DET
$$((\alpha_0, \alpha_1)) = \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix} \begin{bmatrix} \alpha_0 \\ \alpha_n \end{bmatrix} = \begin{bmatrix} \alpha_0 + \alpha_1 \\ \alpha_0 - \alpha_1 \end{bmatrix}$$

$$P(w^0) = P(1) = a_0 + a_1$$

 $P(w^1) = P(-1) = a_0 - a_1$

$$(A^{-1})_{i,j} = \frac{\omega}{\omega}$$

$$\begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix} \begin{bmatrix} 1/2 & 1/2 \\ 1/2 & -1/2 \end{bmatrix}^2 \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

Vandermonde matrices

If (do, do,..., or,)

are distinct points

then this is a

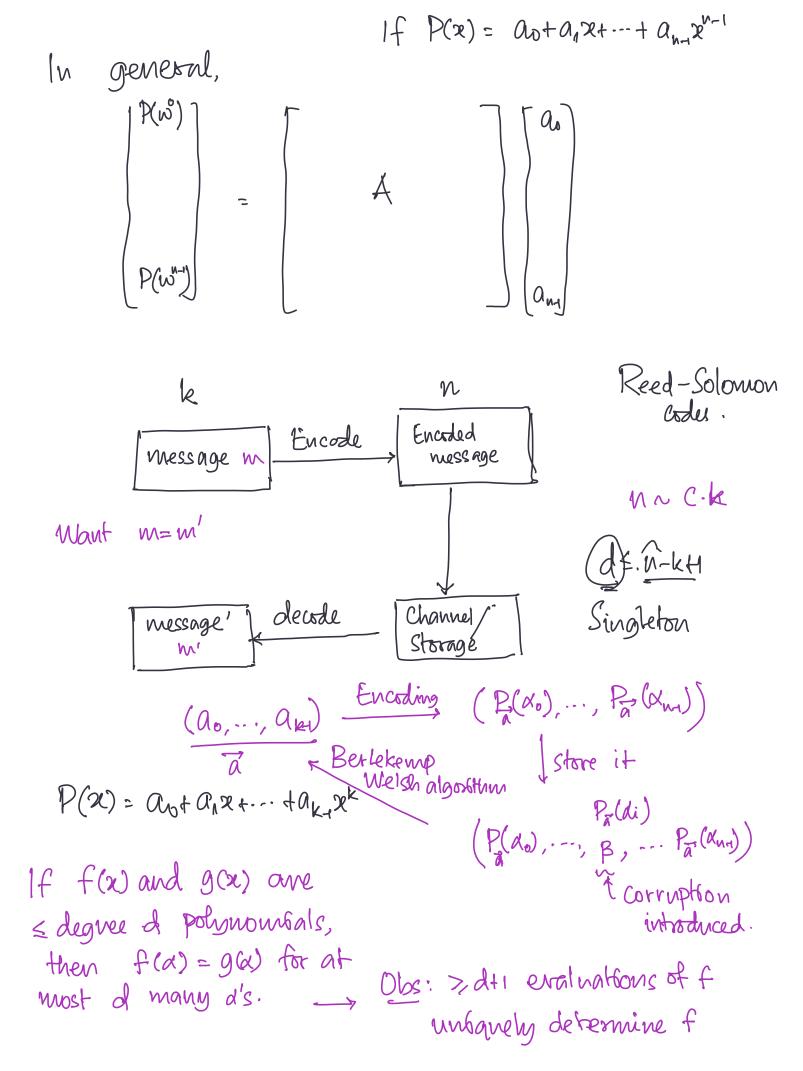
full sk watsox

and inverses are

easy to compute.

$$\prod_{i\neq j} (\alpha_i - \alpha_j).$$

$$\begin{bmatrix} \frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & -\frac{1}{2} \end{bmatrix} \begin{bmatrix} a_0 + a_1 \\ a_1 - a_1 \end{bmatrix} = \begin{bmatrix} \frac{1}{2} (a_0 + a_1) + \frac{1}{2} (a_0 - a_1) \\ \frac{1}{2} (a_0 + a_1) - \frac{1}{2} (a_0 - a_1) \end{bmatrix} = \begin{bmatrix} a_0 \\ a_1 \end{bmatrix}$$



Interpolation: $f(x) = G_0 + G_1 \times + \cdots + G_k \times d$ $f(x_0), \dots, f(x_d) \cdot \begin{cases} \begin{cases} \frac{d}{2} & G(x_0) \\ \frac{d}{2} & G(x_0) \end{cases} = f(x_0) \end{cases}$ $\begin{cases} \begin{cases} \frac{d}{2} & G(x_0) \\ \frac{d}{2} & G(x_0) \end{cases} = \begin{cases} \frac{d}{2} & G(x_0) \\ \frac{d}{2} & G(x_0) \end{cases} = \begin{cases} \frac{d}{2} & G(x_0) \\ \frac{d}{2} & G(x_0) \end{cases}$