

# PSET 4 Solutions

1

a)  $X = \text{linspace}(-1, 1, 50)$

$$C_{\text{computed}} = \text{Vander}(X) \setminus X$$

$$C_{\text{actual}} = \begin{pmatrix} 0 \\ 0 \\ 0 \\ \vdots \\ 0 \\ 0 \\ 0 \end{pmatrix} = \text{zeros}(50, 1); \quad C_{\text{actual}}(49) = 1$$

$$\text{fractional error} = \frac{\text{norm}(C_{\text{computed}} - C_{\text{actual}})}{\text{norm}(C_{\text{actual}})}$$

On my machine, this is 0

b)  $C_{\text{computed}} = \text{Vander}(X) \setminus (\frac{1}{3} * X)$

$$C_{\text{actual}} = \text{zeros}(50, 1) \quad C_{\text{actual}}(49) = 1/3$$

$$\text{fractional error} = 1.96 \cdot 10^{-5} \quad (\text{that's a large fraction!})$$

c)  $\text{Cond}(\text{Vander}(X)) = 2.4 \cdot 10^{18}$

Rounding error on order of  $10^{-16}$

This condition number would lead us to expect fractional error of costs on order of  $10^2$ .

There must be additional sources of error

2

a) The rows add up to 1 means

$$A \begin{pmatrix} 1 \\ 1 \end{pmatrix} = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$

Hence  $\begin{pmatrix} 1 \\ 1 \end{pmatrix}$  is an eigenvector of eigenvalue 1

Because  $A$  is symmetric, it has an orthonormal basis of eigenvectors.

Thus  $\begin{pmatrix} -1 \\ 1 \end{pmatrix}$  must be an eigenvector

$$A \begin{pmatrix} -1 \\ 1 \end{pmatrix} = \begin{pmatrix} -1/2 \\ 1/2 \end{pmatrix}$$

Hence  $\begin{pmatrix} -1 \\ 1 \end{pmatrix}$  is eigenvector of eigenvalue  $1/2$ .

Normalizing and putting in form of eigenvalue decomposition

$$A = \begin{pmatrix} 1/\sqrt{2} & -1/\sqrt{2} \\ 1/\sqrt{2} & 1/\sqrt{2} \end{pmatrix} \begin{pmatrix} 1 & 0 \\ 0 & 1/2 \end{pmatrix} \begin{pmatrix} 1/\sqrt{2} & 1/\sqrt{2} \\ -1/\sqrt{2} & 1/\sqrt{2} \end{pmatrix}$$

b)  $A = Q \Lambda Q^*$

$$A^n = \underbrace{Q \Lambda Q^*}_{\text{orthogonal}} \underbrace{Q \Lambda Q^*}_{\text{orthogonal}} \dots \underbrace{Q \Lambda Q^*}_{\text{orthogonal}}$$

Because  $Q$  is orthogonal,  $Q^* Q = I$

$$A^n = Q \Lambda^n Q^*$$

$$\text{Note } \Lambda^n = \begin{pmatrix} 1^n & 0 \\ 0 & (1/2)^n \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 0 & 1/2^n \end{pmatrix}$$

$$\text{As } n \rightarrow \infty, A^n \rightarrow \begin{pmatrix} 1/\sqrt{2} & -1/\sqrt{2} \\ 1/\sqrt{2} & 1/\sqrt{2} \end{pmatrix} \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} 1/\sqrt{2} & 1/\sqrt{2} \\ -1/\sqrt{2} & 1/\sqrt{2} \end{pmatrix} = \begin{pmatrix} 1/2 & 1/2 \\ 1/2 & 1/2 \end{pmatrix}$$

3) Rank of  $A=1$

Thus reduced SVD of  $A$  is

$$A = \begin{pmatrix} 1 \\ U_1 \\ 1 \end{pmatrix} (\sigma_1) (-V_1^*) \quad \text{where } U_1 \text{ is } 3 \times 1 \\ V_1 \text{ is } 2 \times 1$$

The range of  $A$  is the span of  $U_1$ .

Observing from  $A = \begin{pmatrix} 0 & 4 \\ 0 & 0 \\ 0 & 0 \end{pmatrix}$ , the range of  $A$  is span of  $\begin{pmatrix} 4 \\ 0 \\ 0 \end{pmatrix}$ .

Normalizing gives  $U_1 = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}$

To find the right singular vector, we first find the null space of  $A$ , which is any multiple of  $\begin{pmatrix} 1 \\ 0 \end{pmatrix}$ .

The first right singular vector must be perpendicular to  $\begin{pmatrix} 1 \\ 0 \end{pmatrix}$  and normal, thus, it is  $\begin{pmatrix} 0 \\ 1 \end{pmatrix}$ .

Observing  $A \begin{pmatrix} 0 \\ 1 \end{pmatrix} = 4 \cdot \begin{pmatrix} 1 \\ 0 \end{pmatrix}$  we see  $\sigma_1 = 4$

Thus  $A = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} (4) \begin{pmatrix} 0 & 1 \end{pmatrix}$  is reduced SVD.

To make this a full SVD, we add the remaining orthonormal columns to  $U$  and  $V$

$$A = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 4 & 0 \\ 0 & 0 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}$$

b) The null space is spanned by the right singular vectors of singular value zero.

$$AV_i = \sigma_i U_i \Rightarrow V_i \text{ in Null}(A) \text{ if } \sigma_i = 0$$

These right singular vectors are  $\begin{pmatrix} -1/\sqrt{6} \\ 2/\sqrt{6} \\ 1/\sqrt{6} \end{pmatrix}$  and  $\begin{pmatrix} 1/\sqrt{3} \\ 1/\sqrt{3} \\ -1/\sqrt{3} \end{pmatrix}$ .

$$4) \quad V_j(k) = e^{2\pi i j k / N}$$

To show  $V_j$  is an eigenvector,

$$\text{show } (AV_j)(k) = \lambda_j V_j(k)$$

$$AV_j(k) = \sum_m A_{km} V_j(m)$$

$$= -1 V_j(k-1) + 2 V_j(k) - 1 V_j(k+1)$$

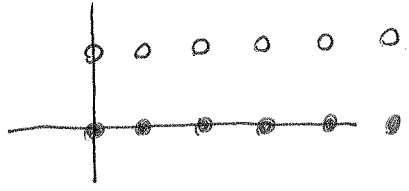
$$= -e^{2\pi i j (k-1)/N} + 2e^{2\pi i j k / N} - e^{2\pi i j (k+1)/N}$$

$$= e^{2\pi i j k / N} \left[ -1 e^{-2\pi i j / N} + 2 - e^{2\pi i j / N} \right]$$

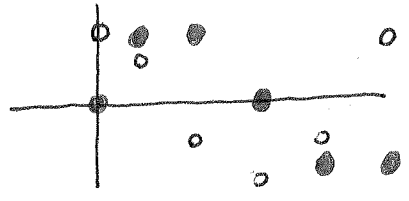
$$= e^{2\pi i j k / N} \left[ 2 - 2 \cos 2\pi j / N \right]$$

Hence  $V_j$  is eigenvector of eigenvalue  $2 - 2 \cos 2\pi j / N$

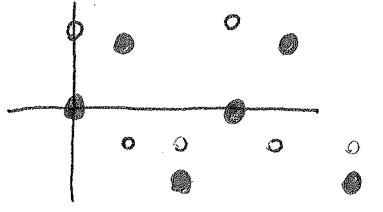
5a



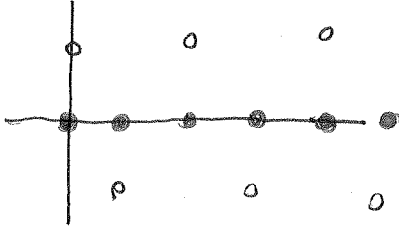
$j=0$



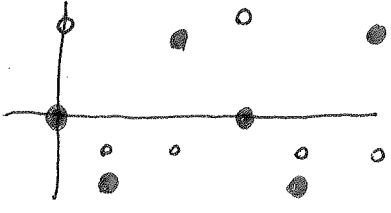
$j=1$



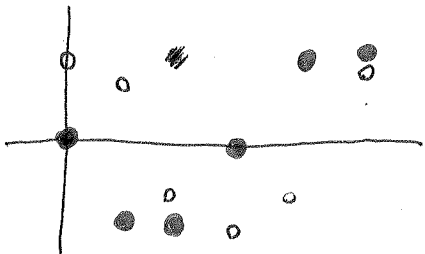
$j=2$



$j=3$



$j=4$

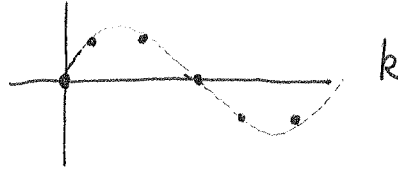


$j=5$

— real part ○  
— imaginary part ●

5b)

$$X = \begin{pmatrix} 0 \\ 1 \\ 1 \\ 0 \\ -1 \\ -1 \end{pmatrix}$$



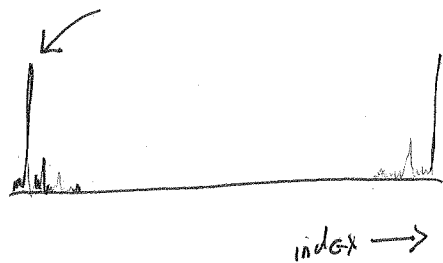
$$\begin{aligned} X(k) &= \frac{\sin 2\pi \frac{k \cdot 1}{6}}{\sin 2\pi/6} = \frac{2}{\sqrt{3}} \sin \frac{2\pi k}{6} \\ &= \frac{2}{\sqrt{3}} \left[ \frac{e^{2\pi i k/6}}{2i} - \frac{e^{-2\pi i k/6}}{2i} \right] \end{aligned}$$

$$F_N\left(\frac{1}{N}\hat{X}\right) = X.$$

$$\text{So } \frac{1}{6}\hat{X} = \begin{pmatrix} 0 \\ \frac{1}{\sqrt{3}i} \\ 0 \\ 0 \\ 0 \\ 0 \\ -\frac{1}{\sqrt{3}i} \end{pmatrix}$$

$$\hat{X} = \begin{pmatrix} 0 \\ -2\sqrt{3}i \\ 0 \\ 0 \\ 0 \\ 2\sqrt{3}i \end{pmatrix}$$

6) fft of signal looks like  $\circ$   
at (2-indexed) position 294.



```
x = wavread('single_note.wav')  
plot(abs(fft(X)))
```

This corresponds to wave that oscillates  
293 times in 66150 index positions.

The sampling rate is  $44100 \frac{\text{sample}}{\text{sec}}$ ,

So the wave oscillates at  
293 times in 1.5 sec  
or 195.3 Hz.

A piano table gives this as  $G_3$   
(though  $F$  actually hit a  $G$  of neighboring octave)