

Correct.

①

a)

~~some~~ ~~Sometimes~~

always

$$\|x\|_2 = \sigma_{\max}(x) \leq \sqrt{\|x\|_F^2} = \sqrt{\sum \sigma_i^2}$$

if  $\sigma_{\max} = \sqrt{\sum \sigma_i^2}$  rank 1 so only one non-0  $\sigma$

8.5/12

b)

Sometimes

$$A: (q, \epsilon)\text{-RIR} \text{ if } \|x\|_b \leq q \Rightarrow (1-\epsilon)\|x\|_2^2 \leq \|Ax\|_2^2 \leq (1+\epsilon)\|x\|_2^2$$

$x$  soln to  $\{\min \|z\| \mid Az = b\}$

Correct.

HW 4 had another example

c)

$$L = D - A$$

$$\text{suppose } Lv = \lambda v$$

$$= (D - A)v$$

if  $D = I$  then

~~Sometimes~~

Never

only true sometimes  
in expectation.

Sometimes. True whenever D is a scaling of the identity. 1.5/3

d)

Never

$$\|Vx\|_2^2 = \|x\|_2^2 \neq \|V^T x\|_2^2$$

Sometimes. True when  $n = k$ . 1/3

)

a) 3 - C, sharp drop off between 1st + 2nd

1 - a, smoother color gradient so less gap between  $\sigma_i$

flipped a and b. 3/4

2 - b, by default, BW then a bunch of grey-tones

b) ~~WAA~~ image C would yield the cleanest picture although you could very easily approximate this image with a much lower rank approximation.  $\|X\|_F^2 = \sum \sigma_i^2$   
error  $\downarrow$  as  $\sigma_i \downarrow$

correct

c) power method converges fastest when the gap between  $\sigma_i$  is large.  
 $\Sigma^3$  has the largest gap, so it'll converge the fastest.  
 $\Sigma^1$  has the smallest gap  $\Rightarrow$  slow converg.

$\sigma_2^2$  actually has the smallest gap, but correct reasoning. 1.5/2

3

7/8

a)

$$\begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} v = \lambda v$$

no self-loops,  $\text{diag}_{ii} = 0$

$A$  is symmetric so by hint

$$\text{tr}(A) = \sum_{i=1}^n \lambda_i = 0$$

since all entries along the diagonal ~~are~~ are 0.

since  $\sum_{i=1}^n \lambda_i = 0$ , there must exist at least

one  $\lambda_j > 0$  and  $\lambda_k < 0$ .

**what prevents all eigenvalues from being 0? 3/4**

b) Laplacian  $L = B^T B$  is PSD

meaning all  $\lambda_i \geq 0$

hence adjacency matrix  $A$  must have  
at least some eigenvalues which are different.

**good!**

4

3/10

a) by JL we have that

$$\|\Pi A v\|_2^2 \leq (1+\epsilon) \|A v\|_2^2$$

$$\text{and } \|\Pi v\|_2^2 \leq (1+\epsilon) \|v\|_2^2$$

$$\Pi A = U_d S_d V_d^T$$

need to use subspace  
embedding theorem. 2/6

$$\Pi A \tilde{v}_d = \lambda_d \tilde{v}_d$$

if  $\lambda$  is an eigenvalue for  $A$  then  $f(\lambda)$  is an eigenvalue for  $f(A)$

$$V = [v_1, \dots, v_d]$$

$$\text{Since } \|\Pi v\|_2^2 \leq (1+\epsilon) \|v\|_2^2$$

$$\text{and } \|\Pi A v\|_2^2 \leq (1+\epsilon) \|A v\|_2^2$$

$$\text{then } \|A \tilde{v}_d\|_2^2 \leq (1+\epsilon) \|v_d\|_2^2$$

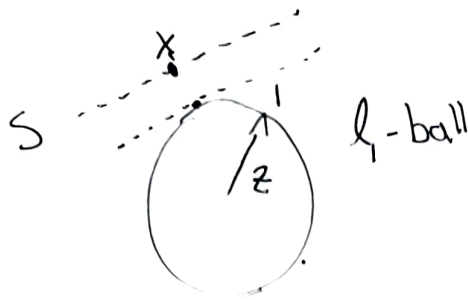
b)  $O(d^3)$  vs  $O(nd^2)$  so yes, faster  
using an iterative technique such as  
the Power Method or Krylov

then we're in  $O(d^2)(I)$  for  $I$  iterations

above method is  $O(mnd)$  time,  
so actually not faster. Can be  
speed up with subsampled  
randomized hadamard. 1/4

5

a)



correct answer, but incomplete reasoning. 4/6

$$C = 1$$

$$H = \{y : \text{sign}(y) \cdot y = 1\}$$

for a given point  $x$ , if  $\|x\|_1 \leq 1$

then ~~sign(x)~~ we're in our set

if  $\|x\|_1 > 1$  then either  $x > 1$  or  $-x > 1$

$\text{sign}(x) x = 1$  will always be within our point  $x$  and the set  $S$ .

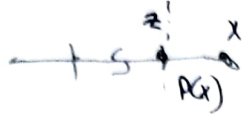
$H$  is a tangent line to  $S$  so will always serve as a <sup>separating</sup> hyperplane.

5)

b)



$$a = x - p(x) = z \in S \text{ closest to } x$$



$$(x - p(x)) \cdot y = c$$

direction is  $(x - p(x))$

want magnitude is  $\|z\|_2$

$$\|z - x\|_2^2 = \|z\|_2^2 - \|x\|_2^2 \quad \sqrt{p(x)^2 + \|x\|_2^2} = c$$

$$\text{gives you } H = \{y : (x - p(x)) \cdot y = \sqrt{p(x)^2 + \|x\|_2^2}\}$$

what is y here?  $(x - p(x))^\top p(x)$   
works. 3/6