

A4 Eryk Halicki

1. a) i) False.

$x = \bar{x}$ in the case where b is in $\text{Col}(A)$, since $\text{proj}_{\text{Col}(A)} b = b$.

ii) if b is orthogonal to $\text{Col}(A)$ then we are solving for

$$A\hat{x} = 0, \text{ since } \text{proj}_{\text{Col}(A)}(b) = 0.$$

So, \hat{x} is in $\text{Null}(A)$.

$\therefore \text{Null}(A) = \text{Row}(A)^\perp$, \hat{x} is orthogonal to $\text{Row}(A)$.

True

iii) take the case where A has L.I columns but is not square. Then A^{-1} does not exist, so the equation doesn't make sense.

False

iv) if a system has infinite solutions, we know $Ax=b$ has a solution, by definition. so $b \in \text{Col}(A)$, and the least squares solution is the same as the normal solution, I.E. $\hat{x}=x$. And since there are infinite choices of x , there are infinite LS solutions.

True

v) If \hat{x}_1 and \hat{x}_2 are ^{LS} solutions to $Ax=b$, then $A\hat{x}_1=\hat{b}$ and $A\hat{x}_2=\hat{b}$.
if $\hat{x}_1+\hat{x}_2$ is also an LS solution, then $A(\hat{x}_1+\hat{x}_2)=\hat{b}=A\hat{x}_1+A\hat{x}_2=2\hat{b}$
but, $\hat{b}=2\hat{b}$ is only true if $\hat{b}=0$, so this is False

$$b) \quad i) A = \begin{bmatrix} 1 & -1 \\ 3 & 2 \\ -2 & 4 \end{bmatrix} \quad b = \begin{bmatrix} 4 \\ 1 \\ 3 \end{bmatrix}$$

$$A^T A x = A^T b$$

$$\begin{pmatrix} 1 & 3 & -2 \\ -1 & 2 & 4 \end{pmatrix} \begin{bmatrix} 1 & -1 \\ 3 & 2 \\ -2 & 4 \end{bmatrix} x = A^T b$$

$$\begin{bmatrix} 14 & -3 \\ -3 & 21 \end{bmatrix} x = \begin{bmatrix} 1 \\ 10 \end{bmatrix}$$

$$14x_1 - 3x_2 = 1$$

$$-3x_1 + 21x_2 = 10$$

$$-98x_1 + 21x_2 = -7$$

$$-3x_1 + 21x_2 = 10$$

$$-95x_1 = -17$$

$$x_1 = \frac{17}{95} \approx 0.1789$$

$$x_2 \approx 0.50$$

$Ax=b$ is overconstrained,
with 3 equations
and 2 variables,
and has linearly
independent columns,
so we know LS
will only have one
solution.

$$x = \begin{bmatrix} 0.1789 \\ 0.5017 \end{bmatrix}$$

$$ii) \begin{bmatrix} 1 \\ 3 \\ -2 \end{bmatrix} = v_1$$

$$a_2 = \begin{bmatrix} -1 \\ 2 \\ 4 \end{bmatrix}$$

$$v_2 = a_2 - \text{proj}_{v_1}(a_2)$$

$$= a_2 - \frac{v_1 \cdot a_2}{\|v_1\|^2} v_1$$

$$= a_2 - \frac{[-1+6-8]}{14} \begin{bmatrix} 1 \\ 3 \\ -2 \end{bmatrix}$$

$$= \begin{bmatrix} -1 \\ 2 \\ 4 \end{bmatrix} - \begin{bmatrix} -3/14 \\ -9/14 \\ 6/14 \end{bmatrix}$$

$$\approx \begin{bmatrix} -0.786 \\ 2.643 \\ 3.571 \end{bmatrix}$$

$$\left\{ \begin{bmatrix} 1 \\ 3 \\ -2 \end{bmatrix}, \begin{bmatrix} -0.786 \\ 2.643 \\ 3.571 \end{bmatrix} \right\} \text{ basis for } \text{Col}(A)$$

$$\text{proj}_{\text{col}(A)}(b) = \frac{v_1 \cdot b}{v_1 \cdot v_1} v_1 + \frac{v_2 \cdot b}{v_2 \cdot v_2} v_2$$

$$\approx \begin{bmatrix} -0.323 \\ 1.540 \\ 1.649 \end{bmatrix} = \hat{b}$$

$$A\hat{x} = \hat{b}$$

$$x_1 - x_2 = -0.323$$

$$3x_1 + 2x_2 = 1.540$$

$$-2x_1 + 4x_2 = 1.649$$

$$x_1 = -0.323 + x_2$$

$$3(-0.323 + x_2) + 2x_2 = 1.54$$

$$-0.969 + 3x_2 + 2x_2 = 1.54$$

$$5x_2 = 2.509$$

$$x_2 = 0.5018$$

$$x_1 = 0.1788$$

$$\hat{x} \approx \begin{bmatrix} 0.1788 \\ 0.5018 \end{bmatrix}$$

$$\text{iii) } \|A\hat{x} - b\| = \left\| \begin{bmatrix} -0.323 \\ 1.540 \\ 1.649 \end{bmatrix} - \begin{bmatrix} 4 \\ 1 \\ 3 \end{bmatrix} \right\| = 4.56$$

A4 cont.

$$1. c) -x_1 + 2x_2 = 7$$

$$3x_1 - 6x_2 = -1$$

$$A = \begin{bmatrix} -1 & 2 \\ 3 & -6 \end{bmatrix} \quad b = \begin{bmatrix} 7 \\ -1 \end{bmatrix}$$

$$A^T A \hat{x} = A^T b$$

$$\begin{bmatrix} 10 & -20 \\ -20 & 40 \end{bmatrix} \hat{x} = \begin{bmatrix} -10 \\ 20 \end{bmatrix}$$

$$\begin{bmatrix} 10 & -20 \\ 0 & 0 \end{bmatrix} \hat{x} = \begin{bmatrix} -10 \\ 0 \end{bmatrix}$$

$$10x_1 - 20x_2 = -10$$

$$x_1 - 2x_2 = -1$$

$$x_1 = -1 + 2x_2 \quad + = x_2$$

$$\hat{x} = \begin{bmatrix} -1 \\ 0 \end{bmatrix} + \begin{bmatrix} 2 \\ 1 \end{bmatrix} +$$

2. a)

$$Q = [q_1, q_2, q_3, \dots, q_n]$$

$$q_1 = \frac{\begin{bmatrix} 1 \\ -1 \\ 1 \\ -1 \end{bmatrix}}{\|a_1\|}$$

$$= \frac{\begin{bmatrix} 1/\sqrt{4} \\ -1/\sqrt{4} \\ 1/\sqrt{4} \\ -1/\sqrt{4} \end{bmatrix}}{\|a_1\|} = \begin{bmatrix} 1/2 \\ -1/2 \\ 1/2 \\ -1/2 \end{bmatrix}$$

$$q_2 = \frac{a_2 - \text{proj}_{W_1}(a_2)}{\|a_2 - \dots\|}$$

$$= a_2 - \frac{a_2 \cdot q_1}{\|q_1\|^2} q_1$$

$$= \begin{bmatrix} 1/2 \\ 1/2 \\ 1/2 \\ 1/2 \end{bmatrix}$$

$$q_3 = \frac{a_3 - \text{proj}_{W_2}(a_3)}{\|a_3 - \dots\|}$$

$$= a_3 - \left(\frac{a_3 \cdot q_1}{q_1 \cdot q_1} q_1 + \frac{a_3 \cdot q_2}{q_2 \cdot q_2} q_2 \right)$$

$$= \begin{bmatrix} 0 \\ -0.7071 \\ 0 \\ 0.7071 \end{bmatrix} \text{ norm}(\dots)$$

$$Q = \begin{bmatrix} 1/2 & 1/2 & 0 \\ -1/2 & 1/2 & -0.7071 \\ 1/2 & 1/2 & 0 \\ -1/2 & 1/2 & 0.7071 \end{bmatrix}$$

$$A = QR$$

$$Q^T A = R$$

$$Q^T A = \begin{bmatrix} 2 & -1 & -0.5 \\ 0 & 1 & 2.5 \\ 0 & 0 & 0.7071 \end{bmatrix} = R$$

b) True. If M has LI columns, so does R .

$$M = QR$$

$$M\hat{x} = b$$

$$M^T M \hat{x} = M^T b$$

$$(QR)^T (QR) \hat{x} = (QR)^T b$$

$$R^T Q^T Q R \hat{x} = R^T Q^T b$$

$$R^T R \hat{x} = R^T Q^T b$$

$$\cancel{(R^T)^{-1} R^T R \hat{x}} = \cancel{(R^T)^{-1} R^T Q^T b}$$

$$R \hat{x} = Q^T b$$

since R is square and has LI columns, it is invertible.

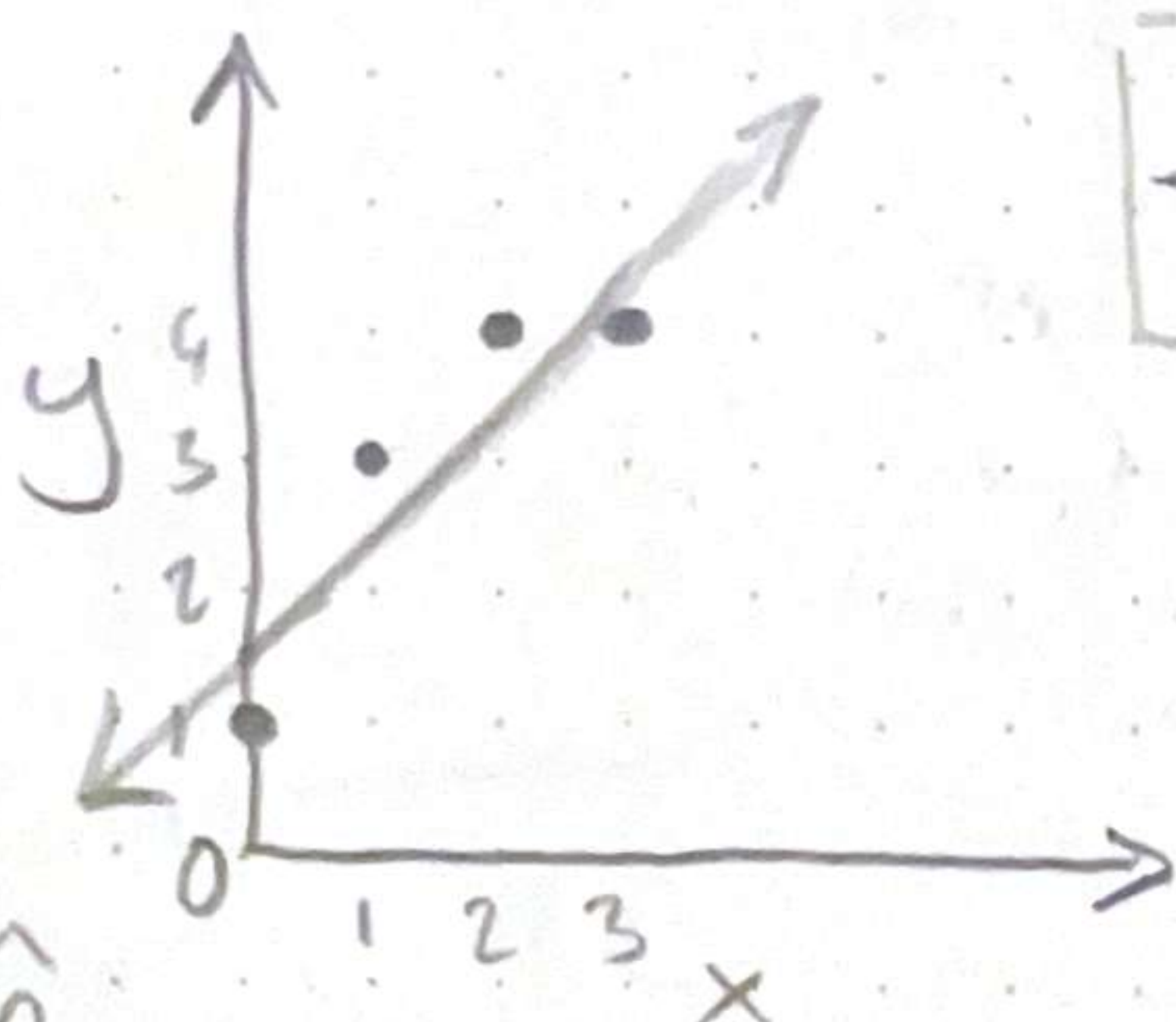
$$c) Ax=b$$

$$R\hat{x} = Q^T b$$

$$\begin{bmatrix} 2 & -1 & -0.5 \\ 0 & 1 & 2.5 \\ 0 & 0 & 0.7071 \end{bmatrix} \hat{x} = \begin{bmatrix} 0.5 & -0.5 & 0.5 & -0.5 \\ 0.5 & 0.5 & 0.5 & 0.5 \\ 0 & -0.7071 & 0 & 0.7071 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix}$$

$$\hat{x} = \begin{bmatrix} 1 \\ 2 \\ 0 \end{bmatrix}$$

3. a)



$$\text{---} = 1.5 + x$$

$$\begin{bmatrix} 1 & 0 \\ 1 & 1 \\ 1 & 2 \\ 1 & 3 \end{bmatrix} \begin{bmatrix} \hat{\beta}_0 \\ \hat{\beta}_1 \end{bmatrix} = \begin{bmatrix} 1 \\ 2 \\ 3 \\ 4 \end{bmatrix}$$

$$\hat{\beta} = \begin{bmatrix} 1.5 \\ 1 \end{bmatrix}$$

$$X^T X \hat{\beta} = X^T Y$$

$$\begin{bmatrix} 1 & 1 & 1 & 1 \\ 0 & 1 & 2 & 3 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \\ 1 \\ 2 \\ 3 \end{bmatrix} \hat{\beta} = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 0 & 1 & 2 & 3 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \\ 3 \\ 4 \end{bmatrix}$$

b) c)
digital

4. a) ~~the~~ In order to be symmetric,
 $A^T = A$.

in the case $B = A^T A$

$$C = A A^T$$

We can show both B and C are symmetric matrices by the following.

$$B^T = B$$

$$(A^T A)^T = A^T A$$

$$A^T A^T A = A^T A \checkmark$$

$$C^T = C$$

$$(A A^T)^T = A A^T$$

$$A^T A^T A = A A^T \checkmark$$

b) i) True.

As per part a), $A^T A$ and $A A^T$ are both symmetric. And according to the spectral theorem, if a matrix is symmetric, it is orthogonally diagonalizable.

ii) False

A matrix is symmetric IFF it is orthogonally diagonalizable.

~~But~~ Not all orthogonal matrices are symmetric, eg. $\begin{bmatrix} 0 & 0 & 1 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}$

So not all orthogonal matrices are orthogonally diagonalizable.

iii) True

If A is orthogonally diagonalizable, we automatically know it is symmetric and Real.

Additionally, if a symmetric matrix has all real entries, all its eigenvalues are real, so we know, A has all real eigenvalues.

iv) True

If we use Gram-Schmidt, we can find a real, orthogonal basis for W , such that

$Q = [q_1, q_2, \dots, q_k]$, where q_i are the basis vectors for W .

Once we have Q , we can normalize each column, and have a matrix B , containing orthonormal basis vectors for W .

It can be shown that projection matrix $P = BB^T$.

~~BBT~~

BB^T is automatically symmetric

since $(BB^T)^T = BB^T$.

Since B is real, BB^T is real and symmetric, and has all real eigenvalues.

c) $A = \begin{bmatrix} 6 & -2 \\ -2 & 9 \end{bmatrix}$

i) A is symmetric, so it is orthogonally diagonalizable. As such, in PDP^{-1} , P is orthogonal, ~~and~~ ~~it~~ meaning all eigenvectors of A corresponding to different eigenvalues are orthogonal!

ii) $\begin{bmatrix} 6 & -2 \\ -2 & 9 \end{bmatrix}$ $\det(A - \lambda I) = (6 - \lambda)(9 - \lambda) - 4$

$$= 54 - 6\lambda - 9\lambda + \lambda^2 - 4$$

$$= \lambda^2 - 15\lambda + 50$$

$$= (\lambda - 10)(\lambda - 5) = 0$$

$$\lambda = 10, 5$$

$$(A - 10I)x = 0$$

$$\begin{bmatrix} -4 & -2 \\ -2 & -1 \end{bmatrix} x = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$-4x_1 = 2x_2$$

$$x_1 = \frac{-x_2}{2}$$

$$\begin{bmatrix} -1/2 \\ 1 \end{bmatrix} = x / \|x\| = x = \begin{bmatrix} -0.447 \\ 0.894 \end{bmatrix}$$

$$(A - 5I)x = 0$$

$$\begin{bmatrix} 1 & -2 \\ -2 & 4 \end{bmatrix} x = 0$$

$$x_1 - 2x_2 = 0$$

$$x_1 = 2x_2$$

$$x = \begin{bmatrix} 2 \\ 1 \end{bmatrix} / \|x\| = x = \begin{bmatrix} 0.894 \\ 0.447 \end{bmatrix}$$

$$P = \begin{bmatrix} 0.894 & -0.447 \\ 0.447 & 0.894 \end{bmatrix} \quad D = \begin{bmatrix} 5 & 0 \\ 0 & 10 \end{bmatrix}$$

$$C^T$$

i) by spectral theorem,

$$uu^T = \cancel{Q} D \cancel{Q}^T \text{ Since } uu^T \text{ is symmetric.}$$

Through spectral decomposition,

$$uu^T = \lambda_1 q_1 q_1^T + \lambda_2 q_2 q_2^T \dots + \lambda_n q_n q_n^T$$

Since $(uu^T)u = u(u^T u)$, u is an eigenvector of uu^T with eigenvalue $u^T u \cdot (\|u\|^2)$

$$\text{let } q_1 = \frac{u}{\|u\|} \text{ and } \lambda_1 = \|u\|^2$$

$$\text{then, } uu^T = \|u\|^2 q_1 q_1^T + \lambda_2 q_2 q_2^T \dots$$

$$\text{but } q_1 q_1^T = u \frac{1}{\|u\|} u^T \frac{1}{\|u\|}$$

$$= \frac{uu^T}{\|u\|^2}$$

so the equation becomes

$$uu^T = \cancel{\|u\|^2} \frac{uu^T}{\cancel{\|u\|^2}}$$

implying uu^T only has one non zero

eigenvalue. This means $\text{rank}(D) = 1$,

and by extension, $\text{rank}(Q D Q^T) = 1 = \text{rank}(uu^T)$

$$\text{ii) } A = 5 \cdot \begin{bmatrix} 0.894 \\ 0.447 \end{bmatrix} \begin{bmatrix} 0.894 & 0.447 \end{bmatrix} \\ + 10 \cdot \begin{bmatrix} -0.447 \\ 0.894 \end{bmatrix} \begin{bmatrix} -0.447 & 0.894 \end{bmatrix}$$

~~th~~


```

1 import numpy as np
2 import math
3
4 w = np.array([[44,61,81,113,131]]).T
5 p = np.array([[91,98,103,110,112]]).T
6
7 ln_w = np.log(w)
8
9 W = np.hstack((np.ones((w.shape[0],1)), ln_w))
10 print(W)
11
12 B = np.linalg.solve(W.T@W, W.T@p)
13 print(B)
14
15 child_estimate = B[0,0] + B[1,0] * math.log(100)
16 print(child_estimate)

```

```

3b.py
81 [[1.          3.78418963]
82  [1.          4.11087386]
83  [1.          4.39444915]
84  [1.          4.72738782]
85  [1.          4.87519732]]
86 [[17.92434012]
87  [19.38499925]]
88 107.19556073585245

```

12,


```

1 import numpy as np
2
3 t = np.atleast_2d(np.arange(13)).T
4 p = np.array([[0, 8.8, 29.9, 62.0, 104.7, 159.1, 222.0, 294.5, 380.4, 471.1
5 , 571.7, 686.8, 809.2]]).T
6
7 T = np.hstack((np.ones((t.shape[0],1)), t, t*t, t*t*t))
8 print(T)
9
10 B = np.linalg.solve(T.T@T, T.T@p)
11 print(B)
12
13 velocity_estimate = B.T@np.array([[1, 4.5, 4.5**2, 4.5**3]]).T
14 print(velocity_estimate)
15
16 # It was predictable that there would be a unique least squares solution
17 # because the system is over determined (more equations than variables)
18 # since there are 13 equations and 4 variables (B0, B1, B2, B3)
19

```

3c.py

14,24

All

```

283 [1.000e+00 6.000e+00 3.600e+01 2.160e+02]
284 [1.000e+00 7.000e+00 4.900e+01 3.430e+02]
285 [1.000e+00 8.000e+00 6.400e+01 5.120e+02]
286 [1.000e+00 9.000e+00 8.100e+01 7.290e+02]
287 [1.000e+00 1.000e+01 1.000e+02 1.000e+03]
288 [1.000e+00 1.100e+01 1.210e+02 1.331e+03]
289 [1.000e+00 1.200e+01 1.440e+02 1.728e+03]]
290 [[-0.85576923]
291 [ 4.70248501]
292 [ 5.55536963]
293 [-0.02736014]]
294 [[130.30845561]]

```