Please print your name:

Bonus challenge. Let me know about any typos you spot in the posted solutions (or lecture sketches). Any typo, that is not yet fixed by the time you send it to me, is worth a bonus point.

Problem 1. The final exam will be comprehensive, that is, it will cover the material of the whole semester.

- Make sure that you have completed all homework.
- Review the practice problems for both midterms (for the material up to Midterm #2).
- The problems below cover the material since Midterm #2.

Problem 2.

- (a) Determine the SVD of $A = \begin{bmatrix} 2 & 0 \\ 0 & 2 \\ 1 & 1 \end{bmatrix}$.
- (b) Determine the best rank 1 approximation of $A = \begin{bmatrix} 2 & 0 \\ 0 & 2 \\ 1 & 1 \end{bmatrix}$.
- (c) Determine the SVD of $A = \begin{bmatrix} 2 & 1 \\ 2 & 1 \end{bmatrix}$.

Solution.

(a) $A^T A = \begin{bmatrix} 2 & 0 & 1 \\ 0 & 2 & 1 \end{bmatrix} \begin{bmatrix} 2 & 0 \\ 0 & 2 \\ 1 & 1 \end{bmatrix} = \begin{bmatrix} 5 & 1 \\ 1 & 5 \end{bmatrix}$ has 6-eigenvector $\begin{bmatrix} 1 \\ 1 \end{bmatrix}$ and 4-eigenvector $\begin{bmatrix} -1 \\ 1 \end{bmatrix}$.

We conclude that $V = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix}$ and $\Sigma = \begin{bmatrix} \sqrt{6} & \\ & 2 \end{bmatrix}$.

$$\boldsymbol{u}_1 = \frac{1}{\sigma_1} A \boldsymbol{v}_1 = \frac{1}{\sqrt{6}} \begin{bmatrix} 2 & 0 \\ 0 & 2 \\ 1 & 1 \end{bmatrix} \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 1 \end{bmatrix} = \frac{1}{\sqrt{12}} \begin{bmatrix} 2 \\ 2 \\ 2 \end{bmatrix} = \frac{1}{\sqrt{3}} \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$$

$$m{u}_2 = rac{1}{\sigma_2} A m{v}_2 = rac{1}{2} \left[egin{array}{cc} 2 & 0 \\ 0 & 2 \\ 1 & 1 \end{array}
ight] rac{1}{\sqrt{2}} \left[egin{array}{cc} -1 \\ 1 \end{array}
ight] = rac{1}{2\sqrt{2}} \left[egin{array}{cc} -2 \\ 2 \\ 0 \end{array}
ight] = rac{1}{\sqrt{2}} \left[egin{array}{cc} -1 \\ 1 \\ 0 \end{array}
ight]$$

 $m{u}_3$ needs to be chosen so that the matrix U is orthogonal. To find such a vector, we can start with a random vector like $\left[egin{array}{c} 0 \\ 0 \\ 1 \end{array} \right]$ and then apply a step of Gram–Schmidt to produce a vector that is orthogonal to $\left[egin{array}{c} 1 \\ 1 \\ 1 \end{array} \right]$ and $\left[egin{array}{c} -1 \\ 1 \\ 0 \end{array} \right]$:

$$\begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} - \begin{bmatrix} \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \cdot \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} - \begin{bmatrix} \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \cdot \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix} \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix} \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} - \frac{1}{3} \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} = \frac{1}{3} \begin{bmatrix} -1 \\ -1 \\ 2 \end{bmatrix}.$$
 We normalize this to $\frac{1}{\sqrt{6}} \begin{bmatrix} -1 \\ -1 \\ 2 \end{bmatrix}.$

Hence,
$$A = U\Sigma V^T$$
 with $U = \begin{bmatrix} 1/\sqrt{3} & -1/\sqrt{2} & -1/\sqrt{6} \\ 1/\sqrt{3} & 1/\sqrt{2} & -1/\sqrt{6} \\ 1/\sqrt{3} & 0 & 2/\sqrt{6} \end{bmatrix}$, $\Sigma = \begin{bmatrix} \sqrt{6} & 0 \\ 0 & 2 \\ 0 & 0 \end{bmatrix}$, $V = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix}$.

(b) From the SVD we just computed it follows that the best rank 1 approximation of A is (that is, we keep 1 singular value only) is

$$\frac{1}{\sqrt{3}} \begin{bmatrix} 1\\1\\1 \end{bmatrix} \begin{bmatrix} \sqrt{6} \end{bmatrix} \frac{1}{\sqrt{2}} \begin{bmatrix} 1\\1 \end{bmatrix}^T = \begin{bmatrix} 1&1\\1&1\\1&1 \end{bmatrix}.$$

(c) $A^TA = \begin{bmatrix} 2 & 2 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} 2 & 1 \\ 2 & 1 \end{bmatrix} = \begin{bmatrix} 8 & 4 \\ 4 & 2 \end{bmatrix}$ has 10-eigenvector $\begin{bmatrix} 2 \\ 1 \end{bmatrix}$ and 0-eigenvector $\begin{bmatrix} -1 \\ 2 \end{bmatrix}$.

(Hint: We can immediately read off the 0-eigenvector (make sure that's obvious!). It then follows from the spectral theorem that the vector orthogonal to it must be another eigenvector.)

Normalizing, we conclude that $V = \frac{1}{\sqrt{5}} \begin{bmatrix} 2 & -1 \\ 1 & 2 \end{bmatrix}$ and $\Sigma = \begin{bmatrix} \sqrt{10} & 0 \end{bmatrix}$.

$$\boldsymbol{u}_1 = \frac{1}{\sigma_1} A \boldsymbol{v}_1 = \frac{1}{\sqrt{10}} \left[\begin{array}{cc} 2 & 1 \\ 2 & 1 \end{array} \right] \frac{1}{\sqrt{5}} \left[\begin{array}{c} 2 \\ 1 \end{array} \right] = \frac{1}{\sqrt{50}} \left[\begin{array}{c} 5 \\ 5 \end{array} \right] = \frac{1}{\sqrt{2}} \left[\begin{array}{c} 1 \\ 1 \end{array} \right]$$

We cannot obtain u_2 in the same way because $\sigma_2 = 0$. Since for every vector u_2 , $Av_2 = \sigma_2 u_2$, we can choose u_2 as we wish, as long as the columns of U are orthonormal in the end.

Let's choose $u_2 = \frac{1}{\sqrt{2}} \begin{bmatrix} -1 \\ 1 \end{bmatrix}$ (the only other choice is $u_2 = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ -1 \end{bmatrix}$). Then, $U = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix}$.

In summary, $A = U\Sigma V^T$ with $U = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix}$, $\Sigma = \begin{bmatrix} \sqrt{10} & \\ 0 & \end{bmatrix}$, $V = \frac{1}{\sqrt{5}} \begin{bmatrix} 2 & -1 \\ 1 & 2 \end{bmatrix}$.

Problem 3. Consider $A = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 0 & -1 \end{bmatrix}$.

- (a) Determine the SVD of A.
- (b) Determine the best rank 1 approximation of A.
- (c) Determine the pseudoinverse of A.
- (d) Find the smallest solution to $\begin{bmatrix} 1 & 1 & 1 \\ 1 & 0 & -1 \end{bmatrix} x = \begin{bmatrix} 2 \\ 1 \end{bmatrix}$.

(Then, as a mild check, compare its norm to the obvious solution $\boldsymbol{x} = [\ 1 \ \ 1 \ \ 0\]^T.)$

Solution.

(a) $A^T A = \begin{bmatrix} 2 & 1 & 0 \\ 1 & 1 & 1 \\ 0 & 1 & 2 \end{bmatrix}$ has characteristic polynomial

$$\det\left(\begin{bmatrix}2-\lambda & 1 & 0\\ 1 & 1-\lambda & 1\\ \hline 0 & 1 & 2-\lambda\end{bmatrix}\right) = 0 - 1 \cdot \det\left(\begin{bmatrix}2-\lambda & 0\\ 1 & 1\end{bmatrix}\right) + (2-\lambda)\det\left(\begin{bmatrix}2-\lambda & 1\\ 1 & 1-\lambda\end{bmatrix}\right)$$
$$= -(2-\lambda) + (2-\lambda)\underbrace{((2-\lambda)(1-\lambda) - 1)}_{=\lambda^2 - 3\lambda + 1}$$
$$= (2-\lambda)(\lambda^2 - 3\lambda) = (2-\lambda)\lambda(\lambda - 3).$$

Hence, the eigenvalues are 0, 2, 3.

- The 0-eigenspace null $\left(\begin{bmatrix} 2 & 1 & 0 \\ 1 & 1 & 1 \\ 0 & 1 & 2 \end{bmatrix}\right)$ has basis $\begin{bmatrix} 1 \\ -2 \\ 1 \end{bmatrix}$. Normalized: $\frac{1}{\sqrt{6}}\begin{bmatrix} 1 \\ -2 \\ 1 \end{bmatrix}$
- The 2-eigenspace null $\left(\begin{bmatrix}0&1&0\\1&-1&1\\0&1&0\end{bmatrix}\right)$ has basis $\begin{bmatrix}-1\\0\\1\end{bmatrix}$. Normalized: $\frac{1}{\sqrt{2}}\begin{bmatrix}-1\\0\\1\end{bmatrix}$

• The 3-eigenspace null
$$\left(\begin{bmatrix} -1 & 1 & 0 \\ 1 & -2 & 1 \\ 0 & 1 & -1 \end{bmatrix}\right)$$
 has basis $\begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$. Normalized: $\frac{1}{\sqrt{3}}\begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$

Therefore,
$$V = \begin{bmatrix} 1/\sqrt{3} & -1/\sqrt{2} & 1/\sqrt{6} \\ 1/\sqrt{3} & 0 & -2/\sqrt{6} \\ 1/\sqrt{3} & 1/\sqrt{2} & 1/\sqrt{6} \end{bmatrix}$$
 and $\Sigma = \begin{bmatrix} \sqrt{3} & 0 & 0 \\ 0 & \sqrt{2} & 0 \end{bmatrix}$.

Next,
$$u_1 = \frac{1}{\sigma_1} A v_1 = \frac{1}{\sqrt{3}} \begin{bmatrix} 1 & 1 & 1 \\ 1 & 0 & -1 \end{bmatrix} \frac{1}{\sqrt{3}} \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$
 and $u_2 = \frac{1}{\sigma_2} A v_2 = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 1 & 1 \\ 1 & 0 & -1 \end{bmatrix} \frac{1}{\sqrt{2}} \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ -1 \end{bmatrix}$.

Hence,
$$U = \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}$$
.

In summary,
$$A = U\Sigma V^T$$
 with $U = \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}$, $\Sigma = \begin{bmatrix} \sqrt{3} & 0 & 0 \\ 0 & \sqrt{2} & 0 \end{bmatrix}$, $V = \begin{bmatrix} 1/\sqrt{3} & -1/\sqrt{2} & 1/\sqrt{6} \\ 1/\sqrt{3} & 0 & -2/\sqrt{6} \\ 1/\sqrt{3} & 1/\sqrt{2} & 1/\sqrt{6} \end{bmatrix}$.

Advanced comment. The computations can be considerably simplified if we start by computing AA^T instead (and then U before V). Can you fill in the details?

(b) From the SVD we just computed it follows that the best rank 1 approximation of A is (that is, we keep 1 singular value only) is

$$\begin{bmatrix} 1 \\ 0 \end{bmatrix} \begin{bmatrix} \sqrt{3} \end{bmatrix} \frac{1}{\sqrt{3}} \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}^T = \begin{bmatrix} 1 & 1 & 1 \\ 0 & 0 & 0 \end{bmatrix}.$$

(c) The pseudoinverse of A is

$$A^{+} = V \Sigma^{+} U^{T} = \begin{bmatrix} 1/\sqrt{3} & -1/\sqrt{2} & 1/\sqrt{6} \\ 1/\sqrt{3} & 0 & -2/\sqrt{6} \\ 1/\sqrt{3} & 1/\sqrt{2} & 1/\sqrt{6} \end{bmatrix} \begin{bmatrix} 1/\sqrt{3} & 0 \\ 0 & 1/\sqrt{2} \\ 0 & 0 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix} = \begin{bmatrix} 1/3 & 1/2 \\ 1/3 & 0 \\ 1/3 & -1/2 \end{bmatrix}.$$

(d) The smallest solution to $Ax = \begin{bmatrix} 2 \\ 1 \end{bmatrix}$ is

$$x = A^{+} \begin{bmatrix} 2 \\ 1 \end{bmatrix} = \begin{bmatrix} 1/3 & 1/2 \\ 1/3 & 0 \\ 1/3 & -1/2 \end{bmatrix} \begin{bmatrix} 2 \\ 1 \end{bmatrix} = \begin{bmatrix} 7/6 \\ 2/3 \\ 1/6 \end{bmatrix}.$$

(For comparison, $\|x\| = \sqrt{11/6} \approx 1.354$ is indeed less than $\|[1 \ 1 \ 0]^T\| = \sqrt{2} \approx 1.414$.)

Problem 4. Find the best approximation (in the L^2 sense) of f(x) = x on the interval [0, 4] using a function of the form $y = a + b\sqrt{x}$.

Solution. The best approximation we are looking for is the orthogonal projection of f(x) onto span $\{1, \sqrt{x}\}$, where the dot product of functions is

$$\langle f, g \rangle = \int_0^4 f(t)g(t)dt.$$

To find an orthogonal basis for span $\{1, \sqrt{x}\}$, following Gram-Schmidt, we compute

$$\sqrt{x} - \left(\begin{array}{c} \text{projection of} \\ \sqrt{x} \text{ onto } 1 \end{array} \right) = \sqrt{x} - \frac{\langle \sqrt{x}, 1 \rangle}{\langle 1, 1 \rangle} 1 = \sqrt{x} - \frac{4}{3}.$$

In the last step, we used that

$$\langle 1, 1 \rangle = \int_0^4 1 dt = 4, \quad \langle \sqrt{x}, 1 \rangle = \int_0^4 \sqrt{t} dt = \left[\frac{1}{3/2} t^{3/2} \right]_0^4 = \frac{16}{3}.$$

Hence, $1, \sqrt{x} - \frac{4}{3}$ is an orthogonal basis for span $\{1, \sqrt{x}\}$.

The orthogonal projection of $f: [0,4] \to \mathbb{R}$ onto span $\{1,\sqrt{x}\} = \text{span}\{1,\sqrt{x}-\frac{4}{3}\}$ therefore is

$$\frac{\langle f,1\rangle}{\langle 1,1\rangle} 1 + \frac{\langle f,\sqrt{x}-\frac{4}{3}\rangle}{\langle\sqrt{x}-\frac{4}{3},\sqrt{x}-\frac{4}{3}\rangle} \bigg(\sqrt{x}-\frac{4}{3}\bigg) \ = \ \frac{1}{4} \int_0^4 f(t)\mathrm{d}t + \frac{9}{8} \bigg(\sqrt{x}-\frac{4}{3}\bigg) \int_0^4 f(t) \bigg(\sqrt{t}-\frac{4}{3}\bigg) \mathrm{d}t.$$

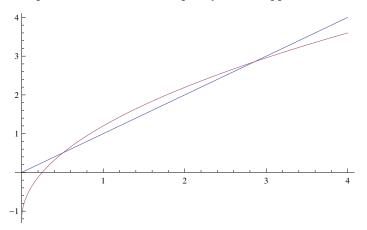
Here, we used that

$$\left\langle \sqrt{x} - \frac{4}{3}, \sqrt{x} - \frac{4}{3} \right\rangle = \int_0^4 \left(\sqrt{t} - \frac{4}{3} \right)^2 \mathrm{d}t = \int_0^4 \left(t - \frac{8}{3} \sqrt{t} + \frac{16}{9} \right) \mathrm{d}t = \left[\frac{t^2}{2} - \frac{16}{9} t^{3/2} + \frac{16}{9} t \right]_0^4 = 8 - \frac{128}{9} + \frac{64}{9} = \frac{8}{9} + \frac{8}{9} + \frac{1}{9} + \frac{$$

In our case, this best approximation is

$$\begin{split} &\frac{1}{4} \int_0^4 t \mathrm{d}t + \frac{9}{8} \bigg(\sqrt{x} - \frac{4}{3} \bigg) \int_0^4 t \bigg(\sqrt{t} - \frac{4}{3} \bigg) \mathrm{d}t \\ &= \frac{1}{4} \bigg[\frac{t^2}{2} \bigg]_0^4 + \frac{9}{8} \bigg(\sqrt{x} - \frac{4}{3} \bigg) \bigg[\frac{2}{5} t^{5/2} - \frac{2}{3} t^2 \bigg]_0^4 = 2 + \frac{12}{5} \bigg(\sqrt{x} - \frac{4}{3} \bigg) = \frac{12}{5} \sqrt{x} - \frac{6}{5}. \end{split}$$

The plot below confirms the quality of this approximation:



Problem 5. Suppose that f(x) is 7-periodic. Write down the first few terms of the Fourier series for f(x) with undetermined coefficients. For one the coefficients, spell out how to compute it.

Solution. The Fourier series for f(x) is

$$f(x) = a_0 + a_1 \cos\left(\frac{2\pi}{7}x\right) + b_1 \sin\left(\frac{2\pi}{7}x\right) + a_2 \cos\left(\frac{4\pi}{7}x\right) + b_2 \sin\left(\frac{4\pi}{7}x\right) + a_3 \cos\left(\frac{6\pi}{7}x\right) + \dots$$

The coefficient a_3 can be computed as

$$a_3 = \frac{\left\langle f(x), \cos\left(\frac{6\pi}{7}x\right)\right\rangle}{\left\langle \cos\left(\frac{6\pi}{7}x\right), \cos\left(\frac{6\pi}{7}x\right)\right\rangle} = \frac{\int_0^7 f(t)\cos\left(\frac{6\pi}{7}t\right)dt}{\int_0^7 \cos^2\left(\frac{6\pi}{7}t\right)dt} = \frac{2}{7}\int_0^7 f(t)\cos\left(\frac{6\pi}{7}t\right)dt.$$

For the final (optional) equality, we used that $\int_0^7 \cos^2(\frac{6\pi}{7}t) dt = \int_0^7 \sin^2(\frac{6\pi}{7}t) dt$ combined with $\cos^2 + \sin^2 = 1$ to conclude that the integral in the denominator must be $\frac{7}{2}$.

Problem 6.

(a) If A has λ -eigenvalue \boldsymbol{v} , then A^3 has

- (b) A is singular if and only if $\dim \text{null}(A)$
- (c) If $A = \begin{bmatrix} i & 1+2i \\ 3 & 4 \\ 5i & 6-i \end{bmatrix}$, then its conjugate transpose is $A^* = \begin{bmatrix} 1 & 1+2i \\ 3 & 4 \\ 5i & 6-i \end{bmatrix}$
- (d) The norm of the vector $\boldsymbol{v} = \left[\begin{array}{c} 1-i \\ 2i \end{array} \right]$ is $\| \boldsymbol{v} \| =$
- (e) By Euler's identity, $e^{ix} =$
- (f) What exactly does it mean for a matrix A to have full column rank?
- (g) The pseudoinverse of $A = \begin{bmatrix} 3 & 0 & 0 \\ 0 & -7 & 0 \end{bmatrix}$ is $A^+ = \begin{bmatrix} 3 & 0 & 0 \\ 0 & -7 & 0 \end{bmatrix}$.
- (h) If A is invertible then its pseudoinverse is $A^+ =$
- (i) If A has full column rank then its pseudoinverse is $A^+=$
- (j) Suppose the linear system Ax = b has infinitely many solutions x.

Which of these solutions is produced by A^+b ?

(k) Write down the 2×2 rotation matrix by angle θ .

Solution.

- (a) If A has λ -eigenvalue \boldsymbol{v} , then A^3 has λ^3 -eigenvalue \boldsymbol{v} .
- (b) A is singular (i.e. not invertible) if and only if dim null(A) > 0.
- (c) If $A = \begin{bmatrix} i & 1+2i \\ 3 & 4 \\ 5i & 6-i \end{bmatrix}$, then its conjugate transpose is $A^* = \begin{bmatrix} -i & 3 & -5i \\ 1-2i & 4 & 6+i \end{bmatrix}$.
- (d) The norm of the vector $\mathbf{v} = \begin{bmatrix} 1-i \\ 2i \end{bmatrix}$ is $\|\mathbf{v}\| = \sqrt{|1-i|^2 + 2^2} = \sqrt{6}$.
- (e) By Euler's identity, $e^{ix} = \cos(x) + i\sin(x)$.
- (f) A matrix A has full column rank if its rank equals the number of columns.
- (g) The pseudoinverse of $A = \begin{bmatrix} 3 & 0 & 0 \\ 0 & -7 & 0 \end{bmatrix}$ is $A^+ = \begin{bmatrix} 1/3 & 0 \\ 0 & -1/7 \\ 0 & 0 \end{bmatrix}$.
- (h) If A is invertible, then $A^+ = A^{-1}$.
- (i) If A has full column rank then its pseudoinverse is $A^+ = (A^T A)^{-1} A^T$.
- (j) The one of smallest norm.
- (k) The 2×2 rotation matrix by angle θ is $\begin{bmatrix} \cos(\theta) & -\sin(\theta) \\ \sin(\theta) & \cos(\theta) \end{bmatrix}$.