Midterm

(due: 2020-03-13)

You may (and should) consult any resources you wish *except* for people (outside the course staff). Provide attribution for any good ideas you might get. Your final write-up should be your own.

True/false For each of the following statements, either give a brief argument that it is true or provide an example to show it is false.

- 1 pt The eigenvalues of a real symmetric positive definite matrix are all positive.
- 1 pt Every square nonsingular matrix A can be decomposed as A = LUwhere L is unit lower triangular and U is upper triangular.
- 1 pt The eigenvalues of a 2-by-2 matrix are differentiable functions of the matrix entries.
- 1 pt For every square matrix, power iteration eventually converges to a dominant eigenvector.
- 1 pt If $A \in \mathbb{R}^{n \times n}$ has condition number $\kappa_2(A) = 1$, then $A = \alpha Q$ for some real $\alpha \neq 0$ and orthogonal matrix $Q \in \mathbb{R}^{n \times n}$.
- 1 pt If $A \in \mathbb{R}^{m \times n}$ has full column rank, then $A^{\dagger}A = I$.

Speedy and stable Consider each of the following:

- 2 pts Rewrite the following code for better speed and numerical stability assuming PA = LU is already computed.
 - $dx = -inv(A)^* dA^* inv(A)^* b$
- 2 pts For x > 1, the equation $x = \cosh(y)$ can be solved as

$$y = -\ln\left(x - \sqrt{x^2 - 1}\right).$$

What happens when $x = 10^8$? How can we fix it?

2 pts Explain the output of the following code fragment

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 $\begin{array}{lll} & x = 2; \\ & & \text{for } k = 1{:}100, \, x = \mathrm{sqrt}(x); \, \, \mathrm{end} \\ & & & \text{for } k = 1{:}100, \, x = x^2; \quad \, \mathrm{end} \\ & & & & & & & \\ & & & & & & & \\ & & & & & & & \\ & & & & & & & \\ & & & & & & & \\ \end{array}$

Polynomial fitting Consider fitting a polynomial of degree at most d to n data points, i.e.

minimize
$$\sum_{i=1}^{n} (p(x_i) - y_i)^2$$

- 2 pts Complete the following code to find $p(x) = \sum_{j=0}^{d} c_j x^j$ that solves the minimization problem by a standard linear least squares formulation:
 - function $[c] = poly_least_squares(x, y, d)$

You may write equivalent code in Julia or Python.

2 pts Modify the code to solve the problem

minimize
$$\sum_{i=1}^{n} (p(x_i) - y_i)^2 + \mu ||c||^2$$

In MATLAB, your routine should have the interface

function $[c] = poly_least_squares_tik(x, y, d, mu)$

You may write equivalent code in Julia or Python.

2 pts Modify the code to solve the problem

minimize
$$\sum_{i=1}^{n} (p(x_i) - y_i)^2 + \mu \int_{-1}^{1} p'(x)^2 dx$$

In MATLAB, your routine should have the interface

function $[c] = poly_least_squares_smooth(x, y, d, mu)$

You may write equivalent code in Julia or Python.

Sensitive pseudoinverse Consider the problem

 $x(s) = (A + sE)^{\dagger}b$

where $A \in \mathbb{R}^{m \times n}$ has full column rank. Suppose the economy QR factorization A = QR is given.

- 2 pts Compute x(0). Your code should take at most O(mn) time. In MAT-LAB, this means filling in the second line in this fragment:
 - [Q,R] = qr(A,0); % This is $O(mn^2)$ 2 x0 = ???; % This should be O(mn)
- 4 pts Compute x'(0) (worth two points). Your code should take at most O(mn) time (worth two points). You can sanity check your code by a finite difference test:

 $\begin{array}{ll} h = 1e{-}6; \\ xp = (A{+}h{}^{*}E) \backslash b; \\ xm = (A{-}h{}^{*}E) \backslash b; \\ dx_{fd} = (xp{-}xm)/2/h; \\ dx = \% \ \text{your computation} \end{array}$

 $_{6}$ relerr = norm(dx fd-dx)/norm(dx)

Conditioning

2 pts Argue that if $A \in \mathbb{R}^{n \times n}$ is symmetric and positive definite and $A_{11} \in \mathbb{R}^{k \times k}$ is a leading principal submatrix, then $\kappa_2(A_{11}) \leq \kappa(A)$, where κ denotes the usual 2-norm condition number for solving linear systems.

Hint: You may use the *interlace theorem*, which tells us that if $\mu_1 \leq \mu_2 \leq \ldots \leq \mu_k$ are the eigenvalues of A_{11} and $\lambda_1 \leq \lambda_2 \leq \ldots \leq \lambda_n$ are the eigenvalues of A, then

$$\lambda_j \le \mu_j \le \lambda_{n-k+j}.$$

- 2 pts Show by example that the hypothesis that A is positive definite is necessary in the above statement.
- 2 pts Show that if $A \in \mathbb{R}^{m \times n} = \begin{bmatrix} A_1 & A_2 \end{bmatrix}$, then $\kappa(A_1) \leq \kappa(A)$, where κ here denotes the usual condition number for least squares.