

# ONLA

## INSTRUCTIONS FOR QUALIFYING EXAMS

Start each problem on a new sheet of paper.

Write your university identification number at the top of each sheet of paper.

### **DO NOT WRITE YOUR NAME!**

Complete this sheet and staple to your answers. Read the directions of the exam very carefully.

STUDENT ID NUMBER \_\_\_\_\_

DATE: \_\_\_\_\_

### EXAMINEES: DO NOT WRITE BELOW THIS LINE

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**Pass/fail recommend on this form.**

Total score: \_\_\_\_\_

DO NOT FORGET TO WRITE YOUR SID NO. ON YOUR EXAM. PLEASE USE BLANK PAGES AT END FOR ADDITIONAL SPACE.

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1. (10 points) Let  $A$  be a unitary matrix.
  - a) Prove that the condition number of  $A$  is equal to 1.
  - b) Prove that  $A$  is orthogonally diagonalizable.

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2. (10 points) Let  $A$  be a real square matrix that has eigendecomposition  $A = V\Lambda V^{-1}$  (here  $\Lambda$  is the diagonal matrix of eigenvalues and  $V$  is the nonsingular eigenvector matrix). Suppose that a perturbation  $A + \delta A$  has eigenvalue  $\mu$ . Prove that there exists some eigenvalue  $\lambda$  of  $A$  such that  $|\lambda - \mu| \leq \kappa(V)\|\delta A\|$ , where  $\kappa(V)$  denotes the condition number of  $V$  and  $\|\cdot\|$  the spectral norm.

Hint: You may wish to first prove that if  $\mu$  is not an eigenvalue of  $A$ , then  $-1$  is an eigenvalue of  $(\Lambda - \mu I)^{-1}V^{-1}\delta AV$ .

3. (10 points) Assume the fundamental axiom of floating point arithmetic is in place.
- a) Prove or disprove that backwards substitution is backward stable for a  $2 \times 2$  upper triangular system.
  - b) Prove or disprove that the addition of 1 is backward stable (i.e. the algorithm defined by  $\tilde{f}(x) = \text{fl}(x) \oplus 1$  where  $\oplus$  denotes floating point addition and  $\text{fl}(x)$  denotes the floating point representation of  $x$ ).

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4. (10 points) Consider  $Ax = b$  with  $A \in \mathbb{R}^{2 \times 2}$  solved with Gauss-Seidel and Jacobi iterations.
- Derive the spectral radius for both methods.
  - Prove that Gauss-Seidel converges if and only if Jacobi converges.

5. (10 points) Consider  $Ax = b$ :

$$\begin{pmatrix} 2 & -1 & 0 \\ -1 & 2 & -1 \\ 0 & -1 & 2 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 4 \\ 0 \\ 0 \end{pmatrix}$$

and the Conjugate Gradient (CG) algorithm

$$r_0 = b - Ax_0, p_0 = r_0,$$

for  $i = 0, 1, 2, \dots$

$$\alpha_i = (r_i^T r_i) / (p_i^T A p_i)$$

$$x_{i+1} = x_i + \alpha_i p_i$$

$$r_{i+1} = r_i - \alpha_i A p_i$$

$$\beta_i = (r_{i+1}^T r_{i+1}) / (r_i^T r_i)$$

$$p_{i+1} = r_{i+1} + \beta_i p_i$$

- a) With  $k$ th Krylov subspace  $\mathcal{K}_k = \text{span}\{r_0, Ar_0, \dots, A^{k-1}r_0\}$ , determine the vectors defining the Krylov spaces for  $k \leq 3$ , taking initial approximation  $x_0 = 0$ .
- b) Solve  $Ax = b$  with CG using zero initial guess  $x_0 = 0$ .
- c) Verify that  $r_0, \dots, r_{k-1}$  form an orthogonal basis for  $\mathcal{K}_k$  for  $k = 1, 2, 3$ .
- d) Verify that  $p_0, \dots, p_{k-1}$  form an A-orthogonal basis for  $\mathcal{K}_k$  for  $k = 1, 2, 3$ .

6. (10 points) Recall that the Lanczos iteration tridiagonalizes a hermitian  $A$  by building towards

$$T_n = \begin{pmatrix} \alpha_1 & \beta_1 & & & \\ \beta_1 & \alpha_2 & \beta_2 & & \\ & \beta_2 & \alpha_3 & \ddots & \\ & & \ddots & \ddots & \beta_{n-1} \\ & & & \beta_{n-1} & \alpha_n \end{pmatrix}.$$

The Lanczos algorithm is given by

$$\beta_0 = 0, \quad q_0 = 0, \quad b = \text{arbitrary}, \quad q_1 = b/\|b\|$$

for  $n = 1, 2, 3, \dots$

$$v = Aq_n$$

$$a_n = q_n^T v$$

$$v = v - \beta_{n-1}q_{n-1} - \alpha_n q_n$$

$$\beta_n = \|v\|$$

$$q_{n+1} = v/\beta_n$$

Assuming exact arithmetic, prove that during Lanczos iterations  $q_{j+1}$  is orthogonal to  $q_1, q_2, \dots, q_j$ .

7. (10 points) Consider the problem

$$\text{extremize } x_1x_2 + x_1^2 \text{ subject to } x_1^2 - 2 \leq x_2 \leq -x_1^2 + 2.$$

- (a) Write down the KKT conditions for this problem and find all points that satisfy them.
- (b) Determine whether or not the points in part (a) satisfy the second order necessary conditions for being local maximizers or minimizers.
- (c) Determine whether or not the points that satisfy the necessary conditions in part (b) also satisfy the second order sufficient conditions for being local maximizers or minimizers.

Hint: Draw a rough sketch of the objective function and the constraints. Maximization of a function  $f$  can be treated as minimization of  $-f$ .

8. (10 points) Let  $X \subseteq \mathbb{R}^n$  be a convex set and let  $f, g_1, \dots, g_m$  be convex functions over  $X$ . Assume there exists  $\hat{x} \in X$  such that

$$g_1(\hat{x}) < 0, \dots, g_m(\hat{x}) < 0.$$

Let  $c \in \mathbb{R}$ . Show that the following are equivalent (nonlinear Farkas lemma).

- (a) The following implication holds:

$$x \in X, g_1(x) \leq 0, \dots, g_m(x) \leq 0 \quad \Rightarrow \quad f(x) \geq c.$$

- (b) There exist  $\lambda_1, \dots, \lambda_m \geq 0$  such that

$$\min_{x \in X} \left\{ f(x) + \sum_{i=1}^m \lambda_i g_i(x) \right\} \geq c.$$

Hint: One direction is easy. For the other direction you may use the following result on the separation of two convex sets: Let  $C_1, C_2 \subseteq \mathbb{R}^n$  be two nonempty convex sets with  $C_1 \cap C_2 = \emptyset$ . Then there exists  $a \in \mathbb{R}^n, a \neq 0$  with  $a^T x \leq a^T y$  for any  $x \in C_1, y \in C_2$ .

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9. (10 points) Let  $f \in C_L^{1,1}(\mathbb{R}^n)$  and assume that  $\nabla^2 f(x) \geq 0$  (positive semi-definite) for all  $x \in \mathbb{R}^n$ . Suppose that the optimal value of the problem  $\min_{x \in \mathbb{R}^n} f(x)$  is  $f^*$ . Let  $\{x_k\}_{k \geq 0}$  be the sequence generated by the gradient descent method with constant stepsize  $\frac{1}{L}$ . Show that if  $\{x_k\}_{k \geq 0}$  is bounded, then  $f(x_k) \rightarrow f^*$  as  $k \rightarrow \infty$ .

Hint:  $C_L^{1,1}(\mathbb{R}^n)$  denotes the set of continuously differentiable functions on  $\mathbb{R}^n$  whose gradient satisfies  $\|\nabla f(x) - \nabla f(y)\|_2 \leq L\|x - y\|_2$  for all  $x, y \in \mathbb{R}^n$ .