

# Numerical Programming 1 (CSE) 2014

(C. Lasser, A. Schreiber and G. Trigila)

## Solutions for Worksheet 7

### Exercise 1

- Let  $v_0$  be an eigenvector for  $\lambda_{max}$ :  $Av_0 = \lambda_{max}v_0$ . If  $\|\cdot\|$  is a matrix norm which is induced by a vector norm, i. e.  $\|A\| = \max_{v \neq 0} \frac{\|Av\|}{\|v\|}$ , then we show the inequality as follows:

$$\|A\| = \max_{v \neq 0} \frac{\|Av\|}{\|v\|} \geq \frac{\|Av_0\|}{\|v_0\|} = \frac{\|\lambda_{max}v_0\|}{\|v_0\|} = |\lambda_{max}|.$$

For general matrix norms (which may not be induced by a vector norm), it is still possible to show the inequality: Consider the matrix  $V_0 = (v_0|v_0|\dots|v_0)$  which repeats the column vector  $v_0$ . Then one has  $AV_0 = \lambda_{max}V_0$  and thus

$$\|A\|\|V_0\| \geq \|AV_0\| = \|\lambda_{max}V_0\| = |\lambda_{max}|\|V_0\|,$$

i. e.  $\|A\| \geq |\lambda_{max}|$ .

- By the definition of the condition number of a square matrix,

$$k(AB) = \|AB\| \underbrace{\|(AB)^{-1}\|}_{\|B^{-1}A^{-1}\|} \leq \|A\|\|B\|\|B^{-1}\|\|A^{-1}\| = k(A)k(B).$$

### Exercise 2

We begin with the values of some integrals we will frequently need:

$$\frac{1}{2} \int_{-1}^1 \begin{matrix} x^2 \\ x^3 \\ x^4 \\ x^5 \\ x^6 \end{matrix} dx = \begin{matrix} \frac{1}{3} \\ 0 \\ \frac{1}{5} \\ 0 \\ \frac{1}{7} \end{matrix}$$

Now, let us start with the Gram-Schmidt process on  $\{1, x, x^2\}$ :

- Normalize  $\tilde{v}_1 = 1$ :

$$v_1 := \frac{\tilde{v}_1}{\|\tilde{v}_1\|} = \frac{\tilde{v}_1}{\sqrt{\langle \tilde{v}_1, \tilde{v}_1 \rangle}} = \frac{\tilde{v}_1}{1} = 1$$

- Subtract the orthogonal projection of  $x$  onto  $span(v_1)$  from  $x$  to make it orthogonal on  $v_1$ :

$$\tilde{v}_2 = x - \underbrace{\langle v_1, x \rangle}_{=0} v_1 = x$$

- Normalize  $\tilde{v}_2 = x$ :

$$v_2 := \frac{\tilde{v}_2}{\|\tilde{v}_2\|} = \frac{x}{\underbrace{\|x\|}_{\sqrt{1/3}}} = \sqrt{3}x.$$

- Subtract the orthogonal projection of  $x^2$  onto the plane  $\text{span}(v_1, v_2)$  from  $x^2$  to make it orthogonal on  $\text{span}(v_1, v_2)$ :

$$\tilde{v}_3 = x^2 - \underbrace{\langle v_1, x^2 \rangle}_{=1/3} v_1 - \underbrace{\langle v_2, x^2 \rangle}_{=0} v_2 = x^2 - \frac{1}{3}$$

- Normalize  $\tilde{v}_3 = x^2 - \frac{1}{3}$ :

$$v_3 := \frac{\tilde{v}_3}{\|\tilde{v}_3\|} = \frac{x^2 - \frac{1}{3}}{\|x^2 - \frac{1}{3}\|} = \frac{x^2 - \frac{1}{3}}{\underbrace{\sqrt{\langle x^2 - 1/3, x^2 - 1/3 \rangle}}_{\sqrt{4/45}}} = (x^2 - \frac{1}{3}) \frac{3}{2} \sqrt{5}.$$

The vector representation of  $\{v_1, v_2, v_3\}$  in the basis  $\{1, x, x^2\}$  is given by  $\begin{pmatrix} v_0 \\ v_1 \\ v_2 \end{pmatrix} = A \begin{pmatrix} 1 \\ x \\ x^2 \end{pmatrix}$  with  $A = \begin{pmatrix} 1 & 0 & 0 \\ 0 & \sqrt{3} & 0 \\ -\frac{1}{2}\sqrt{5} & 0 & \frac{3}{2}\sqrt{5} \end{pmatrix}$ . The other way round, we have  $\begin{pmatrix} 1 \\ x \\ x^2 \end{pmatrix} = A^{-1} \begin{pmatrix} v_0 \\ v_1 \\ v_2 \end{pmatrix}$  with  $A^{-1} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1/\sqrt{3} & 0 \\ 1/3 & 0 & \frac{2}{3\sqrt{5}} \end{pmatrix}$ . The rows of  $A^{-1}$  represent  $\{1, x, x^2\}$  in the basis  $\{v_1, v_2, v_3\}$ .

Continue with the QR decomposition of  $A^{-1}$  e. g. with Housholder reflection:

So we have to project the first row  $\begin{pmatrix} 1 \\ 0 \\ 1/3 \end{pmatrix}$  to a multiple of the first unit vector

$$-\alpha e_1 = \begin{pmatrix} -\sqrt{\frac{10}{9}} \\ 0 \\ 0 \end{pmatrix}. \text{ We have } v = b_1 + x = \begin{pmatrix} 1 - \sqrt{\frac{10}{9}} \\ 0 \\ 1/3 \end{pmatrix} \text{ and so}$$

$$H_v = I - 2 \frac{vv^T}{v^T v} = \begin{pmatrix} 1 & & \\ & 1 & \\ & & 1 \end{pmatrix} - \frac{2}{20/9 - 2\sqrt{\frac{10}{9}}} \begin{pmatrix} (1 - \sqrt{\frac{10}{9}})^2 & \frac{1}{3}(1 - \sqrt{\frac{10}{9}}) \\ \frac{1}{3}(1 - \sqrt{\frac{10}{9}}) & 1/9 \end{pmatrix}$$

One can see that  $R := H_v A^{-1}$  is an upper triangular matrix, so the QR decomposition is  $A^{-1} = H_v^{-1} R = H_v R$ .

For the minimization problem, geometrically we have to project  $x^3$  orthogonally (w.r.t. the inner product considered in this exercise) onto the (three-dimensional) space of polynomials of degree up to 2.

$$\text{Proj}(x^3) = \langle v_1, x^3 \rangle v_1 + \langle v_2, x^3 \rangle v_2 + \langle v_3, x^3 \rangle v_3 = \frac{3}{5}x$$

### Exercise 3

$$H_v v = \left(I - 2 \frac{v v^T}{v^T v}\right) v = v - 2 \frac{v v^T v}{v^T v} = v - 2v = -v$$
$$H_v u = \left(I - 2 \frac{v v^T}{v^T v}\right) u = u - 2 \frac{\overbrace{v v^T u}^{=0}}{v^T v} = u \text{ for } u \text{ perpendicular to } v.$$

Geometric interpretation: The Householder reflector of  $v$  reflects at the hypersurface (dimension:  $n-1$ )  $v^\perp$  orthogonal on  $v$ . So e. g.  $v$  is mapped to  $-v$  and the set  $v^\perp$  is fixed.

Eigenvalues: If one chooses a basis  $\{v_1, v_2, \dots, v_{n-1}\}$  of  $v^\perp$ , then a basis of  $\mathbb{R}^n$  is given by  $\{v, v_1, v_2, \dots, v_{n-1}\}$ . These are eigenvectors for the eigenvalues  $-1, 1, 1, \dots, 1$ . The choice of eigenvectors is not unique in this case, as we could have chosen any basis on  $v^\perp$ , but the eigenvalues of a map/matrix are unique. the determinant is given by the product of all the eigenvalues, so  $-1 \cdot 1 \cdot 1 \cdots 1 = -1$ . This means that volumes are retained but orientations are reversed.