Final Honour School of Mathematics Part A

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20/02/2019

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- 1. Let $f \in C^{n+1}([x_0, x_n], \mathbb{R})$ be a function with (n+1) continuous derivatives and with given function values $f(x_i) = f_i$ at n+1 points $x_0 < x_1 < \cdots < x_n$.
 - (a) [6 marks] Write down the Lagrange interpolating polynomial $p_n(x)$ of degree n for f at the given nodes x_i , and prove that for each $x \in [x_0, x_n]$ there exists $\xi(x) \in (x_0, x_n)$ such that the following error formula holds,

$$e_n(x) := f(x) - p_n(x) = (x - x_0)(x - x_1) \dots (x - x_n) \frac{f^{(n+1)}(\xi(x))}{(n+1)!}.$$

Answer: [Book work] $p_n(x) = \sum_{k=0}^n f_k L_{n,k}(x)$, where

$$L_{n,k}(x) = \frac{(x-x_0)\cdots(x-x_{k-1})(x-x_{k+1})\cdots(x-x_n)}{(x_k-x_0)\cdots(x_k-x_{k-1})(x_k-x_{k+1})\cdots(x_k-x_n)} \quad [1 \text{ mark}].$$

The error formula is trivial for $x = x_i$, as $e_n(x) = 0$ by construction. Let $x \notin \{x_0, \ldots, x_n\}$, and define

$$\phi(t) := e_n(t) - \frac{e_n(x)}{\pi(x)}\pi(t),$$

where

$$\pi(t) := (t - x_0)(t - x_1) \cdots (t - x_n)$$

= $t^{n+1} - \left(\sum_{i=0}^n x_i\right) t^n + \dots (-1)^{n+1} x_0 x_1 \dots x_n \in \Pi_{n+1}$ [2 marks].

By construction, ϕ vanishes at the n + 2 points x and x_i (i = 0, ..., n). Therefore, ϕ' vanishes at n + 1 points $\xi_0, ..., \xi_n$ between these points, ϕ'' vanishes at n points between these new points, and so on until $\phi^{(n+1)}$ vanishes at an (unknown) point $\xi = \xi(x)$ in (x_0, x_n) [1 mark]. But

$$\phi^{(n+1)}(t) = e_n^{(n+1)}(t) - \frac{e_n(x)}{\pi(x)}\pi^{(n+1)}(t) = f^{(n+1)}(t) - \frac{e_n(x)}{\pi(x)}(n+1)!$$

since $p_n^{(n+1)}(t) \equiv 0$ and because $\pi(t)$ is a monic polynomial of degree n+1. The result follows immediately from this identity since $\phi^{(n+1)}(\xi) = 0$ [2 marks].

(b) [6 marks] Derive Simpsons Rule as a special case of Newton-Cotes quadrature.

Answer: [Adaptation of a known technique to a new example] Let x_0, x_1, x_2 be equally spaced points, i.e., $x_0 = x_1 - h$ and $x_2 = x_1 + h$, and let

$$p_2(x) = \sum_{k=0}^2 f(x_k) L_{2,k}(x)$$

= $f(x_0) \frac{(x-x_1)(x-x_1-h)}{2h^2} - f(x_1) \frac{(x-x_1+h)(x-x_1-h)}{h^2} + f(x_2) \frac{(x-x_1+h)(x-x_1)}{2h^2}$

be the Lagrange interpolating polynomial of degree 2 [2 marks]. Newton-Cotes approximates the integral $\int_{x_0}^{x_2} f(x) dx$ by

$$\int_{x_0}^{x_2} p_2(x) dx = \sum_{k=0}^2 f(x_k) \int_{x_0}^{x_2} L_{2,k} dx$$

= $\frac{1}{2h^2} \left(f(x_0) \int_{-h}^{h} x(x-h) dx - 2f(x_1) \int_{-h}^{h} (x+h)(x-h) dx + f(x_2) \int_{-h}^{h} (x+h) x dx \right)$
= $\frac{h}{3} \left(f(x_0) + 4f(x_1) + f(x_2) \right)$ [4 marks].

(c) [6 marks] Derive an error estimate of Simpsons Rule from the approximation error of the Lagrange interpolating polynomial of degree 2. What is the degree of accuracy of Simpsons Rule?

Answer: [Adaptation of book work]

$$\begin{aligned} \left| \int_{x_0}^{x_2} f(x) dx - \frac{h}{3} \left(f(x_0) + 4f(x_1) + f(x_2) \right) \right| &= \left| \int_{x_0}^{x_2} f(x) - p_2(x) dx \right| \\ &\leqslant \int_{x_0}^{x_2} |e_2(x)| \, dx = \int_{-h}^{h} \left| (x+h)x(x-h) \frac{f^{(3)}(\xi(x))}{3!} \right| \, dx \\ &= \left| \frac{f^{(3)}(\xi)}{6} \right| \frac{h^4}{2}, \quad [4 \text{ marks}] \end{aligned}$$

for some $\xi \in (x_0, x_2)$, by the Integral Mean Value Theorem, which shows that Simpson's Rule is exact for polynomials of degree 2, as their third derivative vanishes [1 mark]. However, the refined theory developed in lectures shows that the degree of accuracy of Newton-Cotes integration is in fact n + 1, that is, Simpson's rule is exact for polynomials of degree 3 [1 mark].

(d) [7 marks] Recall that Gauss-Chebychev Quadrature is defined with respect to Chebychev polynomials, which are orthogonal with respect to inner product defined on $L_w^2(-1,1)$ with weight function $w(x) = 1/\sqrt{1-x^2}$. Derive the Gauss-Chebychev Quadrature rule for the Chebychev polynomial of degree 3,

$$\phi_3(x) = 4x^3 - 3x.$$

What is the degree of accuracy of this formula? [*Hint: You may use the fact that* $d/dx \arcsin x = 1/\sqrt{1-x^2}$].

Answer: [New, though application of a known technique] The roots of ϕ_3 are $x_0 = -\sqrt{3}/2$, $x_1 = 0, x_2 = \sqrt{3}/2$ [1 mark]. The Gauss-Chebychev quadrature formula is

$$\int_{-1}^{1} \frac{f(x)}{\sqrt{1-x^2}} \mathrm{d}x \approx \sum_{k=0}^{2} w_k f(x_k),$$

where

$$w_k = \int_{-1}^1 \frac{L_{2,k}(x)}{\sqrt{1-x^2}} \mathrm{d}x.$$
 [1 mark]

Thus, the quadrature formula is given by

$$\begin{split} \int_{-1}^{1} \frac{f(x)}{\sqrt{1-x^2}} \mathrm{d}x &\approx = \frac{2}{3} \Big(f(x_0) \left[\int_{-1}^{1} \frac{x^2}{\sqrt{1-x^2}} \mathrm{d}x - \frac{\sqrt{3}}{2} \int_{-1}^{1} \frac{x}{\sqrt{1-x^2}} \mathrm{d}x \right] \\ &\quad - 2f(x_1) \left[\int_{-1}^{1} \frac{x^2}{\sqrt{1-x^2}} \mathrm{d}x - \frac{3}{4} \int_{-1}^{1} \frac{1}{\sqrt{1-x^2}} \mathrm{d}x \right] + f(x_2) \left[\int_{-1}^{1} \frac{x^2}{\sqrt{1-x^2}} \mathrm{d}x + \frac{\sqrt{3}}{2} \int_{-1}^{1} \frac{x}{\sqrt{1-x^2}} \mathrm{d}x \right] \Big) \\ &\quad = \frac{2}{3} \left(f(x_0) \frac{\pi}{2} - 2f(x_1) \left[\frac{\pi}{2} - \frac{3\pi}{2} \right] + f(x_2) \frac{\pi}{2} \right) \\ &\quad = \frac{\pi}{3} \left(f(x_0) + 4f(x_1) + f(x_2) \right), \end{split}$$

where we used

$$\int_{-1}^{1} \frac{x^2}{\sqrt{1-x^2}} dx \stackrel{\text{(integration by parts)}}{=} 0 + \int_{-1}^{1} \sqrt{1-x^2} dx \stackrel{\text{(substitute } x = \sin t)}{=} \frac{\pi}{2},$$
$$\int_{-1}^{1} \frac{x}{\sqrt{1-x^2}} dx \stackrel{\text{(by symmetry)}}{=} 0,$$
$$\int_{-1}^{1} \frac{1}{\sqrt{1-x^2}} dx \stackrel{\text{(use hint)}}{=} \arcsin x \Big|_{-1}^{1} = \pi. \quad \text{[4 marks]}$$

In lectures we proved that the degree of accuracy of Gauss Quadature is 2n + 1 when a (n + 1)-th degree orthogonal polynomial is used, that is the degree of accuracy is 5 in our case. In other words, if f is a polynomial of degree 5, then the Gauss-Chebychev formula is exact [1 mark].

(a) [5 marks] Define the notion of a norm || · || on a real vector space V. Define the notion of an inner product ⟨·, ·⟩ on V, and prove that ||u|| := ⟨u, u⟩^{1/2} defines a norm on V, the inner product norm. If you use the Cauchy-Schwartz inequality, you need to prove its validity.

Answer: [Book work] A mapping $\|\cdot\|: V \to \mathbb{R}$ is a norm on V if it satisfies

- i) $||u|| \ge 0$ for all $u \in V$, with ||u|| = 0 if and only if $u = 0 \in V$,
- ii) $\|\lambda u\| = |\lambda| \|u\|$ for all $\lambda \in \mathbb{R}$ and all $u \in V$,
- iii) $||u + v|| \le ||u|| + ||v||$ for all $u, v \in V$. [1 mark]

A mapping $\langle \cdot, \cdot \rangle : V \times V \to \mathbb{R}$ is an inner product if it satisfies

- I) $\langle u, u \rangle \ge 0$ for all $u \in V$ and $\langle u, u \rangle = 0$ if, and only if u = 0,
- II) $\langle u, v \rangle = \langle v, u \rangle$ for all $u, v \in V$,
- III) $\langle \alpha u + \beta v, z \rangle = \alpha \langle u, z \rangle + \beta \langle v, z \rangle$ for all $u, v, z \in V$ and all $\alpha, \beta \in \mathbb{R}$. [1 mark]

Now let $||u|| := \langle u, u \rangle^{1/2}$. Then I) implies property i), and

$$\|\lambda u\|^2 \stackrel{\text{III}}{=} \lambda \langle u, \lambda u \rangle \stackrel{\text{III}}{=} \lambda \langle \lambda u, u \rangle \stackrel{\text{III}}{=} \lambda^2 \|u\|^2$$

implies property ii) [1 mark]. Finally,

$$||u+v||^{2} = \langle u+v, u+v \rangle \stackrel{\text{II}),\text{III}}{=} ||u||^{2} + 2\langle u, v \rangle + ||v||^{2} \stackrel{\text{C.S.}}{\leqslant} ||u||^{2} + 2||u|| ||v|| + ||v||^{2} = (||u|| + ||v||)^{2}$$

implies property iii) [1 mark], where the Cauchy-Schwartz Inequality (C.S.) is proven as follows: For every $\lambda > 0$,

$$0 \leqslant \langle u - \lambda v, u - \lambda v \rangle = \|u\|^2 - 2\lambda \langle u, v \rangle + \lambda^2 \|v\|^2 =: \phi(\lambda),$$

which is a quadratic function in λ and takes its minimiser at $\lambda^* = \langle u, v \rangle / ||v||^2$. The C.S.-Inequality now follows from $\phi(\lambda^*) \ge 0$ [1 mark].

(b) [5 marks] Prove that any inner product norm $||u|| := \langle u, u \rangle^{1/2}$ satisfies the *parallelogram* law

$$2||u||^{2} + 2||v||^{2} = ||u+v||^{2} + ||u-v||^{2}, \quad \forall u, v \in V.$$
(1)

Hence, or otherwise, prove that the L_1 norm on V = C([0, 1]) is not an inner product norm.

Answer: [New but elementary]

$$||u+v||^{2} + ||u-v||^{2} = \langle u+v, u+v \rangle + \langle u-v, u-v \rangle$$

= $\langle u, u \rangle + 2\langle u, v \rangle + \langle v, v \rangle + \langle u, u \rangle - 2\langle u, v \rangle + \langle v, v \rangle$
= $2||u||^{2} + 2||v||^{2}$. [3 marks]

Let

$$f(x) = \begin{cases} 1, & (x \in [0, 1/2]), \\ 0, & (x \in (1/2, 1]) \end{cases} \quad g(x) = \begin{cases} 0, & (x \in [0, 1/2]), \\ 1, & (x \in (1/2, 1]). \end{cases}$$

Then $h(x) := f(x) + g(x) \equiv 1$ and $\ell(x) := f(x) - g(x) = 1$ for $x \in [0, 1/2]$ and $\ell(x) = -1$ for $x \in (1/2, 1]$. We have

$$\begin{split} \|f+g\|_{L_{1}}^{2}+\|f-g\|_{L_{1}}^{2} &= \|h\|_{L_{1}}^{2}+\|\ell\|_{L_{1}}^{2} = 1+1 \\ &\neq \frac{1}{2}+\frac{1}{2} = 2\|f\|_{L_{1}}^{2}+2\|g\|_{L_{1}}^{2} \end{split}$$

The L_1 -norm does not satisfy the parallelogram law, and hence it cannot be induced by an inner product [2 marks].

(c) [7 marks] Let $\{\phi_0, \phi_1, \ldots\}$ be a set of orthogonal polynomials with respect to some inner product $\langle \cdot, \cdot \rangle$, and such that $\phi_k(x)$ is of exact degree k ($k = 0, 1, \ldots$). Prove that there exist sequences of real numbers $(\alpha_k)_{k=1}^{\infty}$, $(\beta_k)_{k=1}^{\infty}$ and $(\gamma_k)_{k=1}^{\infty}$ such that the following three-term recursion holds,

$$\phi_{k+1}(x) = \alpha_k(x - \beta_k)\phi_k(x) - \gamma_k\phi_{k-1}(x), \quad (k = 1, 2, \dots).$$
(2)

Answer: [Book work] Since $x\phi_k \in \Pi_{k+1}$ and $\{\phi_0, \phi_1, \ldots, \phi_{k+1}\}$ is a basis for Π_{k+1} , there exist $\sigma_{k,0}, \sigma_{k,1}, \ldots, \sigma_{k,k+1}$ in \mathbb{R} such that

$$x\phi_k(x) = \sum_{i=0}^{k+1} \sigma_{k,i}\phi_i(x).$$
 [1 mark] (3)

Using $x\phi_j \in \Pi_{k-1}$ for $j \leq k-2$, which implies $\langle x\phi_k, \phi_j \rangle = \langle \phi_k(x), x\phi_j(x) \rangle = 0$ [1 mark], we take inner products on both sides of (3) with ϕ_j and find that for $j \leq k-2$,

$$0 = \langle x\phi_k, \phi_j \rangle = \left\langle \sum_{i=0}^{k+1} \sigma_{k,i}\phi_i, \phi_j \right\rangle = \sum_{i=0}^{k+1} \sigma_{k,i}\langle \phi_i, \phi_j \rangle = \sigma_{k,j}\langle \phi_j, \phi_j \rangle.$$
 [2 mark]

Hence, $\sigma_{k,j} = 0$ for $j \leq k-2$, so that (3) becomes

$$x\phi_k(x) = \sigma_{k,k+1}\phi_{k+1}(x) + \sigma_{k,k}\phi_k(x) + \sigma_{k,k-1}\phi_{k-1}(x).$$
 [2 marks]

Since $x\phi_k$ is of exact degree k + 1, we must have $\sigma_{k,k+1} \neq 0$, which makes it possible to solve for ϕ_{k+1} as an expression of the form (2) [1 mark].

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(d) [8 marks] Recall that the Laguerre polynomials are orthogonal with respect to the inner product $\langle f, g \rangle = \int_0^\infty e^{-x} f(x)g(x)dx$. For k = 1 compute the parameters β_k and γ_k/α_k in the three-term recursion (2) that relates the degree 2 Laguerre polynomial $\phi_2(x)$ to the Laguerre polynomials of degrees 1 and 0, $\phi_1(x) = 1 - x$ and $\phi_0(x) = 1$.

Answer: [New] By the previous part, we know that $\alpha_k, \beta_k, \gamma_k$ exist. We may also assume that the ϕ_j are scaled so that $\langle \phi_j, \phi_j \rangle = 1$, which is easily checked for ϕ_0 and ϕ_1 [1 mark]. Since we must ensure that ϕ_{k+1} is orthogonal to ϕ_j ($j \leq k$), we must force this condition in particular for j = k - 1, k, which yields two constraints [1 mark] on the parameters $\alpha_k, \beta_k, \gamma_k$: Orthogonality with ϕ_k yields

$$0 = \langle \phi_{k+1}, \phi_k \rangle = \alpha_k \langle (x - \beta_k) \phi_k, \phi_k \rangle - \gamma_k \langle \phi_{k-1}, \phi_k \rangle = \alpha_k \left(\langle x \phi_k, \phi_k \rangle - \beta_k \right),$$

where we have used $\langle \phi_k, \phi_k \rangle = 1$ and $\langle \phi_k, \phi_{k-1} \rangle = 0$, hence,

$$\beta_k = \int_0^\infty e^{-x} x \phi_k^2(x) dx. \quad [2 \text{ marks}]$$

Orthogonality with ϕ_{k-1} implies

$$0 = \langle \phi_{k+1}, \phi_{k-1} \rangle = \alpha_k \langle (x - \beta_k) \phi_k, \phi_{k-1} \rangle - \gamma_k \langle \phi_{k-1}, \phi_{k-1} \rangle = \alpha_k \langle x \phi_k, \phi_{k-1} \rangle - \gamma_k,$$

where we have used $\langle \phi_{k-1}, \phi_{k-1} \rangle = 1$ and $\langle \phi_k, \phi_{k-1} \rangle = 0$, hence,

$$\frac{\gamma_k}{\alpha_k} = \int_0^\infty e^{-x} x \phi_k(x) \phi_{k-1}(x) dx. \quad [1 \text{ mark}]$$

In the case k = 1 this yields

$$\beta_1 = \int_0^\infty e^{-x} x(1-x)^2 dx = 1 - 2 \cdot 2! + 3! = 3,$$

$$\frac{\gamma_1}{\alpha_1} = \int_0^\infty e^{-x} x(1-x) dx = 1! - 2! = -1,$$

where we used

$$\int_0^\infty x^k e^{-x} dx = -x^k e^{-x} \Big|_0^\infty + k \int_0^\infty x^{k-1} e^{-x} dx = \dots = k! \quad [3 \text{ marks}]$$

- 3. Let $n \ge 2$ be an integer.
 - (a) [6 marks] Define the notion of an orthogonal matrix $Q \in \mathbb{R}^{n \times n}$. Prove that for any given vector $u \in \mathbb{R}^n$ there exists a vector $w \in \mathbb{R}^n$ such that H(w)u = v, where $v^{\mathrm{T}} = (\sqrt{u^{\mathrm{T}}u} \ 0 \ \dots \ 0)$ and H(w) is the Householder reflection

$$H(w) = \mathbf{I} - \frac{2}{w^{\mathrm{T}}w} w w^{\mathrm{T}}.$$

Express w explicitly in terms of u, and show that H(w) is an orthogonal matrix.

Answer: [Book work] Q is orthogonal if $Q^{T}Q = I$, that is, $Q^{T} = Q^{-1}$ [1 mark]. If u = 0, we may choose $w \neq 0$ arbitrarily, as H(w)0 = 0 [1 mark]. If $u \neq 0$, choose w = (u - v) (which is explicit in u, since v is defined in terms of u), so that H(w) is well defined and

$$H(w)u = Iu - \frac{2(u-v)^{\mathrm{T}}u}{(u-v)^{\mathrm{T}}(u-v)}(u-v) = u - (u-v), \quad [2 \text{ marks}]$$

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where the last equality follows from

$$(u-v)^{\mathrm{T}}(u-v) = u^{\mathrm{T}}u - 2u^{\mathrm{T}}v + v^{\mathrm{T}}v = u^{\mathrm{T}}u - 2u^{\mathrm{T}}v + u^{\mathrm{T}}u = 2(u-v)^{\mathrm{T}}u.$$
 [1 mark]

H(w) is orthogonal, because

$$H(w)^{\mathrm{T}}H(w) = \mathbf{I} - \frac{4}{w^{\mathrm{T}}w}ww^{\mathrm{T}} + \frac{4w^{\mathrm{T}}w}{(w^{\mathrm{T}}w)^{2}}ww^{\mathrm{T}} = \mathbf{I}.$$
 [1 mark]

(b) [7 marks] Define the notion of QR-factorisation of a matrix $A \in \mathbb{R}^{n \times m}$ and show how this factorisation can be computed via Householder reflections. Show how the QR-factorisation can, alternatively, be computed via Gram-Schmidt orthogonalisation and compare the flop count of both algorithms.

Answer: [First part is book work. Alternative method is new, though students have seen Gram-Schmidt orthogonalisation.] A QR factorisation of a $n \times n$ matrix A is the splitting of A into a product A = QR of an orthogonal matrix Q and an upper triangular matrix $R = (r_{ij})$, that is, $r_{ij} = 0$ if i > j [1 mark]. Let us recursively construct Householder matrices $H(w_i)$ (i = 1, ..., n - 1) such that

$$H(w_{n-1}) \cdot H(w_{n-2}) \dots H(w_1)A = R$$

is upper triangular, so that $Q = H(w_1)H(w_2) \dots H(w_{n-1})$ satisfies A = QR. Inductively, let $B^0 = A$ and

$$B^{k-1} = H(w_{k-1}) \cdot H(w_{n-2}) \cdot \ldots \cdot H(w_1)A_{k-1}$$

which is of block form

$$B^{k-1} = \begin{pmatrix} b_{ij}^{k-1} \end{pmatrix} = \begin{bmatrix} R^{k-1} & C^{k-1} \\ 0 & D^{k-1} \end{bmatrix}$$
(4)

with \mathbb{R}^{k-1} a $(k-1) \times (k-1)$ upper triangular matrix. Define

$$u_{k} = \begin{bmatrix} b_{kk}^{k-1} \\ \vdots \\ b_{nk}^{k-1} \end{bmatrix} \in \mathbb{R}^{n-k+1}, \quad v_{k} = \begin{bmatrix} \sqrt{u_{k}^{\mathrm{T}}u_{k}} \\ 0 \\ \vdots \\ 0 \end{bmatrix} \in \mathbb{R}^{n-k+1}, \quad w_{k} = \begin{bmatrix} 0 \\ \vdots \\ 0 \\ u_{k} - v_{k} \end{bmatrix} \in \mathbb{R}^{n}.$$

Then

$$H(w_k) \begin{bmatrix} R^{k-1} \\ 0 \end{bmatrix} = \begin{bmatrix} R^{k-1} \\ 0 \end{bmatrix}$$

and the last n - k entries of the first column of

$$H(w_k) \begin{bmatrix} C^{k-1} \\ D^{k-1} \end{bmatrix}$$

are zero, so that B^k is also of the required (partially triangularised) block form (4). The claim thus follows by induction [3 marks]. The complexity is $\sum_{k=0}^{n-1} (n-k)^2 = n^3/3 + n^2/2 + n/6$ flops, plus $O(n^2)$ flops for the construction of the w_k [1 mark]. A QR factorisation can also be computed via Gram-Schmidt orthogonalisation, in which the k-th column q_k of Q is inductively defined from the k-th column a_k of A as follows,

$$h_k = a_k - \sum_{j=1}^{k-1} (q_j^{\mathrm{T}} a_k) q_k,$$
$$q_k = \frac{1}{\sqrt{h_k^{\mathrm{T}} h_k}} h_k,$$

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and setting $R = (r_{ij})$, where $r_{ij} = q_i^{\mathrm{T}} a_j$ for i < j and $r_{jj} = \sqrt{h_j^{\mathrm{T}} h_j}$, a calculation that takes $\sum_{k=1}^{n} 2nk = n^2(n+1) = n^3 + n^2$ flops. This procedure is 3 times more costly, but the columns of Q become available one by one, while the method with Householder matrices can compute these columns only at the end [2 marks].

(c) [5 marks] Prove that for any vector $x \in \mathbb{R}^n$ there exists an $n \times n$ orthogonal Givens rotation matrix J(i,j) such that y = J(i,j)x has entries $y_j = 0$, $y_k = x_k$ $(k \notin \{i,j\})$. Determine y_i and the entries of J(i,j) as a function of the entries of x.

Answer: [Book work and adaptation of seen material] If $x_i = x_j = 0$, then J = I will do and there is nothing to prove [1 mark]. If $(x_i, x_j) \neq (0, 0)$, let

$$J = (g_{ij}) = \begin{bmatrix} 1 & & & & \\ & c & s & & \\ & & c & s & & \\ & & -s & c & & \\ & & & & 1 \end{bmatrix}$$

with $g_{ii} = g_{jj} = c := x_i/\sqrt{x_i^2 + x_j^2}$, $g_{ij} = -g_{ji} = s := x_j/\sqrt{x_i^2 + x_j^2}$, and $g_{k\ell} = \delta_{k,\ell}$ if $(k,\ell) \notin \{i,j\}^2$ [2 marks]. Then $J^T J = I$ and y = Jx satisifes $y_k = x_k$ for $k \notin \{i,j\}$, $y_i = cx_i + sx_j = \sqrt{x_i^2 + y_i^2}$, and $y_j = -sx_i + cx_j = 0$. Thus, J(i,j) := J satisfies the requirements [2 marks].

(d) [7 marks] Show how, for any given symmetric $n \times n$ matrix A, Householder reflections may be used to construct an orthogonal matrix Q such that $B = Q^{T}AQ$ is a tri-diagonal matrix, and compute the number of floating point operations this reduction requires. Show how the eigenvalues and eigenvectors of A can be obtained from the eigenvalues and eigenvectors of B. Show how Givens rotations may be used to compute a QR factorisation of B, and explain the advantage of applying the symmetric QR algorithm to B instead of A for the purposes of computing the eigenvalues and eigenvectors of A.

Answer: [Book work and adaptation of seen material] Let A be symmetric, and let

$$H(w_{n-1}) \cdot H(w_{n-2}) \cdot \ldots \cdot H(w_1)A = Q^{\mathrm{T}}A = R$$

be the upper triangularisation of A under the construction of Part (b). Then $Q^{T}AQ = RQ$ is symmetric, and since right multiplication by $H(w_k)$ only operates on columns k, \ldots, n ,

$$RQ = RH(w_1)H(w_2)\dots H(w_{n-1})$$

has only one lower nonzero subdiagonal. By symmetry, it also has only one upper nonzero superdiagonal, hence $B := Q^{T}AQ$ is tri-diagonal [2 marks]. The cost is twice the cost we computed in Part (b) (except that v does not have to be computed, that is, $(2/3)n^{3}+O(n^{2})$ [1 mark].

Since *B* is similar to *A*, it has the same eigenvalues, and the eigenvectors of *A* can easily be found from the eigenvectors of *B* by left multiplication with *Q* [1 mark]. In each step of the symmetric QR factorisation applied to *B*, we need to compute a QR-factorisation $B = \tilde{Q}\tilde{R}$ of *B* and form $B_+ := \tilde{R}\tilde{Q}$. If this were applied to *A* this would involve a cubic cost. However, since *B* is tri-diagonal, the QR factorisation is achieved by the following procedure, at a total cost of $(16 + c_1)(n - 1)$ flops [1 mark] for a small constant c_1 :
$$\begin{split} \tilde{R} &:= B; \\ \text{for } i = 1, \dots, n-1 \\ x &:= \tilde{R}(:, i) \text{ (i-th column$)}; \\ \text{form } J(i, i+1) \text{ as in Part (c) (4 flops, plus constant c_1 flops for square root); } \\ \tilde{R} \leftarrow J(i, i+1) \tilde{R} \text{ (12 flops)}; \\ \text{end.} \end{split}$$

Since, due to the action of the rotations J(i, j), the factor \tilde{R} has only two nonzero superdiagonals, the formation of $B_+ = \tilde{R}\tilde{Q} = \tilde{R}J(1,2)^T J(2,3)^T \dots J(n-1,n)^T$ also only takes 12(n-1) flops and results in a symmetric matrix with only one non-zero subdiagonal, i.e., a tri-diagonal matrix. Thus, each iteration of the symmetric QR algorithm takes only linear instead of cubic time [2 marks].