

Solutions 2

A.1. First, $p(a, b, c) = \frac{1}{2(\pi^{3/2})} \exp(-(a^2 + 2b^2 + c^2)/2)$. Next, writing term by term the change of variables gives

$$\begin{aligned} a(\lambda, \mu, \theta) &= \lambda \cos^2 \theta + \mu \sin^2 \theta = \frac{\lambda + \mu}{2} + \frac{\lambda - \mu}{2} \cos(2\theta), \\ b(\lambda, \mu, \theta) &= -(\lambda - \mu) \sin \theta \cos \theta = -\frac{\lambda - \mu}{2} \sin(2\theta), \\ c(\lambda, \mu, \theta) &= \lambda \sin^2 \theta + \mu \cos^2 \theta = \frac{\lambda + \mu}{2} - \frac{\lambda - \mu}{2} \cos(2\theta), \end{aligned}$$

so the Jacobian is given by

$$\begin{aligned} J(\lambda, \mu, \theta) &= \det \begin{bmatrix} \cos^2 \theta & \sin^2 \theta & -(\lambda - \mu) \sin(2\theta) \\ -\sin \theta \cos \theta & \sin \theta \cos \theta & -(\lambda - \mu) \cos(2\theta) \\ \sin^2 \theta & \cos^2 \theta & (\lambda - \mu) \sin(2\theta) \end{bmatrix} \\ &= (\lambda - \mu) \det \begin{bmatrix} \cos^2 \theta & \sin^2 \theta & -\sin(2\theta) \\ -\sin \theta \cos \theta & \sin \theta \cos \theta & -\cos(2\theta) \\ \sin^2 \theta & \cos^2 \theta & \sin(2\theta) \end{bmatrix} = \lambda - \mu. \end{aligned}$$

Since $a^2 + 2b^2 + c^2 = \text{Tr}(H_1 H_1^T) = \lambda^2 + \mu^2$, we deduce that

$$p(\lambda, \mu, \theta) = \frac{1}{2(\pi^{3/2})} e^{-(\lambda^2 + \mu^2)/2} |\lambda - \mu|.$$

Now, since $p(\lambda, \mu, \theta)$ is independent of θ , $p(\theta) = \frac{2}{\pi}$ on $[0, \frac{\pi}{2}]$ and

$$p(\lambda, \mu) = \frac{1}{4\sqrt{\pi}} e^{-(\lambda^2 + \mu^2)/2} |\lambda - \mu|.$$

One can check that this is indeed a joint density function on \mathbb{R}^2 (i.e. $\iint_{\mathbb{R}^2} d\lambda d\mu p(\lambda, \mu) = 1$).

Since $\lambda + \mu = \text{Tr}(H_1) = a + c$ and $\lambda\mu = \det(H_1) = ac - b^2$, we have

$$\mathbb{E}(\lambda + \mu) = 0 \quad \text{and} \quad \mathbb{E}(\lambda\mu) = -\frac{1}{2}.$$

The next computation requires some more work:

$$\begin{aligned} \mathbb{E}(\max\{\lambda, \mu\}) &= \iint_{\mathbb{R}^2} d\lambda d\mu \max\{\lambda, \mu\} p(\lambda, \mu) \\ &= \frac{1}{4\sqrt{\pi}} \iint_{\mathbb{R}^2} d\lambda d\mu \max\{\lambda, \mu\} |\mu - \lambda| e^{-(\lambda^2 + \mu^2)/2} \\ &= \frac{1}{2\sqrt{\pi}} \int_{\mathbb{R}} d\lambda \int_{-\infty}^{\lambda} d\mu \lambda (\lambda - \mu) e^{-(\lambda^2 + \mu^2)/2}, \end{aligned}$$

by symmetry of the function. Computing integrals (using integration by parts) gives

$$\mathbb{E}(\max\{\lambda, \mu\}) = \frac{\sqrt{\pi}}{2}.$$

Finally, a computation similar to the preceding gives

$$p(\lambda) = \frac{e^{-\lambda^2}}{2\sqrt{\pi}} + \frac{e^{-\lambda^2/2}}{2\sqrt{\pi}} \lambda \int_0^\lambda d\mu e^{-\mu^2/2}.$$

A.2. The matrix of eigenvectors of H_2 is fixed and given by

$$V = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}.$$

So we have the change of variables

$$\begin{bmatrix} a & c \\ c & a \end{bmatrix} = \frac{1}{2} \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix} \begin{bmatrix} \lambda & 0 \\ 0 & \mu \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}$$

i.e. $a = \frac{1}{2}(\lambda + \mu)$ and $c = \frac{1}{2}(\lambda - \mu)$. Computing the Jacobian of this linear transformation gives

$$J(\lambda, \mu) = -\frac{1}{2}.$$

Since $p(a, c) = \frac{1}{2\pi} \exp\left(-\frac{a^2+c^2}{2}\right)$ and $2(a^2 + c^2) = \text{Tr}(H_2 H_2^T) = \lambda^2 + \mu^2$, we obtain

$$p(\lambda, \mu) = p(a(\lambda, \mu), b(\lambda, \mu)) |J(\lambda, \mu)| = \frac{1}{4\pi} \exp\left(-\frac{\lambda^2 + \mu^2}{4}\right),$$

i.e. λ and μ are i.i.d. $\sim \mathcal{N}_{\mathbb{R}}(0, 2)$ random variables.

Since $\lambda + \mu = \text{Tr}(H_2) = 2a$ and $\lambda - \mu = \det(H_2) = a^2 - c^2$, we have

$$\mathbb{E}(\lambda + \mu) = 0 \quad \text{and} \quad \mathbb{E}(\lambda\mu) = 0.$$

Now, we also obtain by symmetry that

$$\begin{aligned} \mathbb{E}(\max\{\lambda, \mu\}) &= 2 \int_{\mathbb{R}} d\lambda \int_{\lambda}^{\infty} d\mu \mu p(\lambda, \mu) \\ &= 2 \int_{\mathbb{R}} d\lambda \frac{1}{4\pi} e^{-\lambda^2/4} \int_{\lambda}^{\infty} d\mu \mu e^{-\mu^2/4} \\ &= \frac{1}{\pi} \int_{\mathbb{R}} d\lambda e^{-\lambda^2/2} = \sqrt{\frac{2}{\pi}}. \end{aligned}$$

Finally, we trivially have

$$p(\lambda) = \frac{1}{\sqrt{4\pi}} e^{-\lambda^2/4}.$$

B.1. The direct computation of the eigenvalues of H_1 gives

$$\lambda_1 = \frac{a+c}{2} + \sqrt{\left(\frac{a-c}{2}\right)^2 + b^2} \quad \text{and} \quad \lambda_2 = \frac{a+c}{2} - \sqrt{\left(\frac{a-c}{2}\right)^2 + b^2}.$$

NB: we have written λ_1, λ_2 instead of λ, μ in order to point out that λ_1, λ_2 are ordered here ($\lambda_1 \geq \lambda_2$).

Noticing that $\frac{a+c}{2}$, $\frac{a-c}{2}$ and b are i.i.d. $\sim \mathcal{N}_{\mathbb{R}}(0, \frac{1}{2})$ random variables, we obtain that the random vector (λ_1, λ_2) is distributed as $(x+y, x-y)$ where x and y are independent, $x \sim \mathcal{N}_{\mathbb{R}}(0, \frac{1}{2})$ and y is distributed as a Rayleigh random variable (i.e. as the modulus of a $\mathcal{N}_{\mathbb{C}}(0, 1)$ random variable):

$$p(y) = 2y \exp(-y^2), \quad y \geq 0.$$

The joint distribution of (x, y) is given by

$$p(x, y) = \frac{1}{\sqrt{\pi}} e^{-(x^2+y^2)} (2y), \quad x \in \mathbb{R}, y \geq 0.$$

The Jacobian of the linear transformation $x = \frac{\lambda_1+\lambda_2}{2}$, $y = \frac{\lambda_1-\lambda_2}{2}$ is

$$J(\lambda_1, \lambda_2) = -\frac{1}{2}$$

and we notice that $x^2 + y^2 = \frac{\lambda_1^2 + \lambda_2^2}{2}$. Therefore,

$$p(\lambda_1, \lambda_2) = \frac{1}{2\sqrt{\pi}} e^{-\frac{\lambda_1^2 + \lambda_2^2}{2}} (\lambda_1 - \lambda_2), \quad \text{for } \lambda_1 \geq \lambda_2.$$

The joint distribution of the unordered eigenvalues (λ, μ) is therefore given by

$$p(\lambda, \mu) = \frac{1}{4\sqrt{\pi}} e^{-\frac{\lambda^2 + \mu^2}{2}} |\lambda - \mu|, \quad \lambda, \mu \in \mathbb{R}.$$

B.2. Computing directly the eigenvalues of H_2 gives $\lambda = a + c$ and $\mu = a - c$. Since for any $\alpha, \beta \in \mathbb{R}$,

$$\alpha\lambda + \beta\mu = (\alpha + \beta)a + (\alpha - \beta)c$$

is a Gaussian random variable (being the sum of two independent Gaussian random variables), the vector (λ, μ) is a Gaussian random vector, with zero mean and covariance matrix given by

$$\Sigma = \begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix}.$$

We therefore conclude that λ and μ are i.i.d. $\sim \mathcal{N}_{\mathbb{R}}(0, 2)$ random variables.