Problem Set: Mathematical Foundations SOLUTIONS

1. (a)

$$\mathbf{v} \cdot \mathbf{v} = 1$$

$$\beta^2 (9 + 16) = 1$$

$$\beta^2 = \frac{1}{25}$$

$$\beta = \pm \frac{1}{5}$$

(b) Let:

$$\mathbf{w} = \begin{bmatrix} a \\ b \end{bmatrix}$$

for some $a, b \in \mathbb{R}$

Normality condition:

$$\mathbf{w} \cdot \mathbf{w} = 1$$
$$a^2 + b^2 = 1$$

Orthogonality condition:

$$\mathbf{v} \cdot \mathbf{w} = 0$$
$$\beta(3a + 4b) = 0$$

Rearrange 2nd condition and sub in 1st:

$$b = -3a/4$$

$$\implies a^2 + \frac{9}{16}a^2 = 1$$

$$\implies \frac{25}{16}a^2 = 1$$

$$\implies a^2 = \frac{16}{25}$$

$$\implies a = \frac{4}{5}$$

$$\implies b = \frac{-3}{5}$$

So:

$$\mathbf{w} = \frac{1}{5} \begin{bmatrix} 4 \\ -3 \end{bmatrix}$$

- (c) No, because $-\mathbf{w}$ is also orthonormal to \mathbf{v}
- (d) Yes
- (e) If **A** is invertible then $\det \mathbf{A} \neq 0$ by the Invertible Matrix Theorem.

$$\det \mathbf{A} = (4.9) - \alpha.\alpha$$

$$= 36 - \alpha^2$$

$$\implies 36 - \alpha^2 \neq 0$$

$$\implies 36 \neq \alpha^2$$

$$\implies \alpha \neq \pm 6$$

(f) If **A** is positive definite then its eigenvalues, $\{\lambda_i\}_{i=1}^2$, are positive.

Eigenvalues are satisfied by the following characteristic polynomial:

$$\det(\mathbf{A} - \lambda_i \mathbf{I}) = 0$$

$$\begin{vmatrix} 4 - \lambda_i & \alpha \\ \alpha & 9 - \lambda_i \end{vmatrix} = 0$$

$$(4 - \lambda_i)(9 - \lambda_i) - \alpha^2 = 0$$

$$(\lambda_i)^2 - 13\lambda_i + 36 - \alpha^2 = 0$$

$$\implies \lambda_i = \frac{13 \pm \sqrt{13^2 - 4(36 - \alpha^2)}}{2}$$

So for $\lambda_i > 0$:

$$13 - \sqrt{13^2 - 4(36 - \alpha^2)} > 0$$

$$13^2 > 13^2 - 4(36 - \alpha^2)$$

$$36 > \alpha^2$$

$$\implies -6 < \alpha < 6$$

(g) If **A** represents a covariance matrix then the diagonal elements give the variances, σ_1^2 , σ_2^2 :

$$\sigma_1^2 = 4 \implies \sigma_1 = 2$$
 $\sigma_2^2 = 9 \implies \sigma_1 = 3$

The correlation, ρ , is given by the off-diagonal elements, so that:

$$\rho\sigma_1\sigma_2 = \alpha$$

$$\rho = \frac{\alpha}{\sigma_1\sigma_2}$$

$$= \frac{-3}{2 \times 3}$$

$$= -0.5$$

(h) The shape will be an ellipse

The ellipse will be axis-aligned

The ratio of the semi-major to the semi-minor axis will be: 3:2

2. (a) Note that the form of f(x) is:

$$f(x) = 2x^2 - 2(a_1 + a_2)x + (a_1^2 + a_2^2)$$

So, forming the Hessian:

$$\mathcal{H}(x) = 4$$

Thus the Hessian is positive definite.

So f(x) is strictly convex

So the solution is globally optimal and unique (by properties of strictly convex functions)

(b) Note that g(x) has a characteristic 'V' shape, with change of slope at b (a sketch would also be acceptable in order to demonstrate this)

State that g(x) is convex, but not strictly convex

So the solution is globally optimal (by properties of convex functions).

But the solution is unique (by geometry).

(c) Note that h(x) has a characteristic '_/' shape, with changes of slope at b_1 and b_2 (a sketch would also be acceptable in order to demonstrate this)

State that h(x) is convex, but not strictly convex

So the solution is globally optimal (by properties of convex functions).

And the solution is not unique (by geometry).

(d) Note that k(x) = f(x) + h(x), and that the sum of a convex function and a strictly convex function is strictly convex.

So the solution is globally optimal (by properties of convex functions)

And the solution is unique (by properties of strictly convex functions)

3. Consider $x \ge 1$:

$$(1-x) \le 0 < e^{-x}$$

Consider 0 < x < 1:

Binomial expansion, valid for |x| < 1:

$$\frac{1}{1-x} = 1 + x + x^2 + x^3 + \dots$$

Taylor expansion of e^x :

$$e^x = 1 + x + \frac{1}{2}x^2 + \frac{1}{3!}x^3 + \dots$$

For positive x this implies:

$$e^x < \frac{1}{1-x}$$

$$\implies (1-x) < e^{-x}$$

Consider x = 0:

$$\implies (1-x) = 1 = e^0 = e^{-x}$$

Consider x < 0:

Let y = -x > 0:

$$\implies (1-x) = (1+y)$$

And:

$$\implies e^{-x} = e^y = 1 + y + \frac{1}{2}y^2 + + \frac{1}{3!}y^3 + \dots$$

So, since y > 0:

$$(1+y) < e^y$$

$$\implies (1-x) < e^{-x}$$

4. (This solution assumes a linear, risk-neutral, utility function throughut).

$$\mathbb{E}[1 \text{ roll}] = \frac{1}{6} \times 1 + \frac{1}{6} \times 2 + \frac{1}{6} \times 3 + \frac{1}{6} \times 4 + \frac{1}{6} \times 5 + \frac{1}{6} \times 6$$
$$= \frac{1}{6} \times (1 + 2 + 3 + 4 + 5 + 6)$$
$$= 3.5$$

(b) Assume 1st roll takes place at t = 0, 2nd roll takes place at t = 1

$$\mathbb{E}[2\text{nd roll at } t = 1] = \frac{1}{6} \times (1 + 2 + 3 + 4 + 5 + 6)$$

= 3.5

$$\mathbb{E}[\text{1st roll at } t = 0] = \frac{1}{6} \times \left(\max(1, \mathbb{E}[\text{2nd roll at } t = 1]) + \max(2, \mathbb{E}[\text{2nd roll at } t = 1]) + \max(3, \mathbb{E}[\text{2nd roll at } t = 1]) + \max(4, \mathbb{E}[\text{2nd roll at } t = 1]) + \max(5, \mathbb{E}[\text{2nd roll at } t = 1]) + \max(6, \mathbb{E}[\text{2nd roll at } t = 1]) \right)$$

$$= \frac{1}{6} \times (3.5 + 3.5 + 3.5 + 4 + 5 + 6)$$

$$= 4.25$$

(c) Assume 1st roll takes place at t=0, 2nd roll takes place at t=1, 3rd roll takes place at t=2

$$\mathbb{E}[3\text{rd roll at } t = 2] = \frac{1}{6} \times (1 + 2 + 3 + 4 + 5 + 6)$$
$$= 3.5$$

$$\mathbb{E}[2\text{nd roll at } t = 1] = \frac{1}{6} \times \left(\max(1, \mathbb{E}[3\text{rd roll at } t = 2]) + \max(2, \mathbb{E}[3\text{rd roll at } t = 2]) + \max(3, \mathbb{E}[3\text{rd roll at } t = 2]) + \max(4, \mathbb{E}[3\text{rd roll at } t = 2]) + \max(5, \mathbb{E}[3\text{rd roll at } t = 2]) + \max(6, \mathbb{E}[3\text{rd roll at } t = 2])\right)$$

$$= \frac{1}{6} \times (3.5 + 3.5 + 3.5 + 4 + 5 + 6)$$

$$= 4.25$$

$$\mathbb{E}[\text{1st roll at } t = 0] = \frac{1}{6} \times \left(\max(1, \mathbb{E}[\text{2nd roll at } t = 1]) + \max(2, \mathbb{E}[\text{2nd roll at } t = 1]) + \max(3, \mathbb{E}[\text{2nd roll at } t = 1]) + \max(4, \mathbb{E}[\text{2nd roll at } t = 1]) + \max(5, \mathbb{E}[\text{2nd roll at } t = 1]) + \max(6, \mathbb{E}[\text{2nd roll at } t = 1]) \right)$$

$$= \frac{1}{6} \times (4.25 + 4.25 + 4.25 + 4.25 + 5 + 6)$$

$$= 4.67$$

- 5. (a) (Univariate) Normal or Gaussian distribution
 - (b) μ

$$\frac{dp}{dx} = \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{(x-\mu)^2}{2\sigma^2}\right) \times \left(-\frac{2(x-\mu)}{2\sigma^2}\right)$$

$$= 0$$

$$\implies -\frac{2(x-\mu)}{2\sigma^2} = 0$$

$$\implies x = \mu$$