

CQF Module 2 Examination Solutions

Acknowledgement:

The diagrams in question 10 are courtesy of Emerson Bedford (June 2008 Cohort)

- The investment universe is composed of a set of 4 assets:

Asset	μ	σ
A	0.04	0.07
B	0.08	0.12
C	0.12	0.18
D	0.15	0.26

with the following correlation structure

$$R = \begin{pmatrix} 1 & 0.2 & 0.5 & 0.3 \\ 0.2 & 1 & 0.7 & 0.4 \\ 0.5 & 0.7 & 1 & 0.9 \\ 0.3 & 0.4 & 0.9 & 1 \end{pmatrix}$$

Denote the column vector of asset weights by \mathbf{w} , the column vector of asset returns by $\boldsymbol{\mu}$ and the covariance matrix by Σ

a. *Compute the covariance matrix Σ .*

Answer: We will use the covariance matrix decomposition described in class. Define \mathbb{S} as the diagonal matrix with standard deviation on its diagonal:

$$S = \begin{pmatrix} 0.07 & 0 & 0 & 0 \\ 0 & 0.12 & 0 & 0 \\ 0 & 0 & 0.18 & 0 \\ 0 & 0 & 0 & 0.26 \end{pmatrix}$$

Then, by the covariance matrix decomposition, the covariance matrix Σ is given by $\Sigma = SRS$, i.e.

$$\Sigma = \begin{pmatrix} 0.0049 & 0.00168 & 0.0063 & 0.00546 \\ 0.00168 & 0.0144 & 0.01512 & 0.01248 \\ 0.0063 & 0.01512 & 0.0324 & 0.04212 \\ 0.00546 & 0.01248 & 0.04212 & 0.0676 \end{pmatrix}$$

b. Consider the following optimization:

$$\min_{\mathbf{w}} \frac{1}{2} \mathbf{w}^T \Sigma \mathbf{w}$$

Subject to

$$\begin{aligned} \mathbf{w}^T \mathbf{1} &= 1 \\ \mathbf{w}^T \boldsymbol{\mu} &= 0.1 \end{aligned}$$

- Explain in plain English what this optimization does.
- Solve this optimization using the Lagrangian method.
- Compute the standard deviation of the optimal portfolio.
- On a graph of expected returns plotted against standard deviation, identify the optimal portfolio.

Answer:

- This optimization is used to select an optimal asset weight vector \mathbf{w}^* with the objective to minimize half of the portfolio variance, subject to two constraints: the budget equation, and a return constraint specifying the portfolio must provide a 10% return over the period. In short, this optimization determines the minimum-variance portfolio for a given level of return of 10%. As we vary the return constraint from $-\infty$ to $+\infty$, this optimization will enable us to parametrize the **boundary** of the opportunity set.
- In order to solve the optimization problem, we start by forming the Lagrange function: with two Lagrange multipliers λ and γ :

$$L(\mathbf{w}, \lambda, \gamma) = \frac{1}{2} \mathbf{w}^T \Sigma \mathbf{w} + \lambda(0.1 - \boldsymbol{\mu}^T \mathbf{w}) + \gamma(1 - \mathbf{1}^T \mathbf{w})$$

and solve for the first order condition:

$$\begin{aligned} \frac{\partial L}{\partial \mathbf{w}}(\mathbf{w}, \lambda, \gamma) &= \mathbf{w}^T \Sigma - \lambda \boldsymbol{\mu}^T - \gamma \mathbf{1}^T = 0 \\ \frac{\partial L}{\partial \lambda}(\mathbf{w}, \lambda, \gamma) &= 0.1 - \boldsymbol{\mu}^T \mathbf{w} = 0 \\ \frac{\partial L}{\partial \gamma}(\mathbf{w}, \lambda, \gamma) &= (1 - \mathbf{1}^T \mathbf{w}) = 0 \end{aligned}$$

We then get the optimal weight vector \mathbf{w}^*

$$\mathbf{w}^* = (\Sigma)^{-1}(\lambda \boldsymbol{\mu} + \gamma \mathbf{1})$$

where

$$\begin{cases} \lambda & \approx & 0.2542605 \\ \gamma & \approx & -0.0078808 \end{cases}$$

and therefore

$$\mathbf{w}^* = \begin{pmatrix} 5.87\% \\ 75.90\% \\ -31.95\% \\ 50.18\% \end{pmatrix}$$

- The standard deviation of the portfolio is equal to

$$\sigma_{\Pi} = \sqrt{\mathbf{w}^{*T} \Sigma \mathbf{w}^*} \approx 13.25\%$$

- The optimal portfolio is located on the efficient frontier, at the coordinates (0.1, 0.1325).
- c. Consider the following optimization:

$$\min_{\mathbf{w}} \frac{1}{2} \mathbf{w}^T \Sigma \mathbf{w}$$

Subject to

$$\mathbf{w}^T \mathbf{1} = 1$$

- Explain in plain English what this optimization does.
- Solve this optimization using the Lagrangian method.
- Compute the standard deviation of the optimal portfolio.
- On a graph of expected returns plotted against standard deviation, identify and name the optimal portfolio.

Answer:

- This optimization is used to select an optimal asset weight vector \mathbf{w}^* with the objective to minimize half of the portfolio variance, subject to the sole budget constraint. In short, this optimization problem gives us the **global minimum variance portfolio**.
- In order to solve the optimization problem, we start by forming the Lagrange function: with two lagrange multipliers λ and γ :

$$L(\mathbf{w}, \lambda) = \frac{1}{2} \mathbf{w}^T \Sigma \mathbf{w} + \lambda(1 - \mathbf{1}^T \mathbf{w})$$

and solve for the first order condition:

$$\begin{aligned} \frac{\partial L}{\partial \mathbf{w}}(\mathbf{w}, \lambda) &= \mathbf{w}^T \Sigma - \lambda \mathbf{1}^T = 0 \\ \frac{\partial L}{\partial \lambda}(\mathbf{w}, \lambda) &= (1 - \mathbf{1}^T \mathbf{w}) = 0 \end{aligned}$$

From the first equation, we get the optimal weight vector \mathbf{w}^*

$$\mathbf{w}^* = (\Sigma)^{-1}(\lambda \mathbf{1})$$

and substituting into the second, we obtain

$$\lambda = \frac{1}{\mathbf{1}^T \Sigma^{-1} \mathbf{1}} \approx 0.000664337$$

and therefore

$$\mathbf{w}^* = \begin{pmatrix} 90.54\% \\ 82.91\% \\ -137.46\% \\ 64.01\% \end{pmatrix}$$

- The return of the portfolio is equal to

$$\sigma_{\Pi} = \mathbf{w}^{*T} \mu \approx 3.3608\%$$

- The standard deviation of the portfolio is equal to

$$\sigma_{\Pi} = \sqrt{\mathbf{w}^{*T} \Sigma \mathbf{w}^*} \approx 2.5775\%$$

- The optimal portfolio is the **global minimum variance portfolio**. It is located at the tip of the efficient frontier, at the coordinates (0.033608, 0.025775).

2. A butterfly spread can be created by buying call options with strike prices of £15 and £20, and selling two call options with strike prices of £17.5. The initial investment is therefore

$$4 + 0.5 - 2 \times 2 = 0.5$$

The table shows the variation of profit with the final stock price:

Stock Price $S(T)$	Profit
$S(T) < 15$	-0.5
$15 < S(T) < 17.5$	$S(T) - 15.5$
$17.5 < S(T) < 20$	$19.5 - S(T)$
$S(T) > 20$	-0.5

3. Fill in the table

Position	Strategy	Max loss	Max gain	Breakeven
Long Call	Bullish	Premium	Unlimited	Strike + Premium
Short Call	Bearish/neutral	Unlimited	Premium	Strike + Premium
Long Put	Bearish	Premium	Strike - Premium	Strike - Premium
Short Put	Bullish/neutral	Strike - Premium	Premium	Strike - Premium

4. (i):

$$\begin{array}{rcc} & & 15 \\ & V_1 & \\ V & & 0 \\ & V_{-1} & \\ & & 0 \end{array}$$

To find V_1 from portfolio: $\Pi = V - \Delta S$. Then from T_1 to T we have

$$\Pi \longrightarrow \begin{cases} 15 - \Delta(\alpha + 20) \\ 0 - \Delta\alpha \end{cases}$$

so for risk free portfolio \Rightarrow choose $15 - \Delta(\alpha + 20) = -\Delta\alpha \longrightarrow \Delta = 3/4$. For no arbitrage we want

$$V_1 - \Delta(\alpha + 10) = -\Delta\alpha$$

since $r = 0$. Solving gives $V_1 = 7.5$

For V_{-1} :

$$\Pi \longrightarrow \begin{cases} 0 - \Delta\alpha \\ 0 - \Delta(\alpha - 20) \end{cases} \Rightarrow \Delta = 0$$

Therefore $V_{-1} = 0$.

For V :

$$V - \Delta\alpha = \begin{cases} V_1 - \Delta(\alpha + 10) & \equiv \frac{15}{2} - \Delta(\alpha + 10) \\ V_{-1} - \Delta(\alpha - 10) & \equiv 0 - \Delta(\alpha - 10) \end{cases}$$

so $\frac{15}{2} - 20\Delta = 0 \longrightarrow \Delta = 3/8$.

Finally $V - \Delta\alpha = -(\alpha - 10) \longrightarrow V = 10\Delta = 15/4$. Hence

$$\begin{array}{rcc} & & 15 \\ & & 15/2 \\ V = 15/4 & & 0 \\ & 0 & \\ & & 0 \end{array}$$

(ii)

$$\begin{array}{rcc} & & 0 \\ & V_1 & \\ V & & 0 \\ & V_{-1} & \\ & & 15 \end{array}$$

same method gives:-

$$\begin{array}{rcc} & & 0 \\ & 0 & \\ V = 15/4 & & 0 \\ & 15/2 & \\ & & 15 \end{array}$$

(iii)

$$V(S, T) = \begin{cases} S - \alpha - 5 & S \geq \alpha + 5 \\ 0 & \alpha - 5 \leq S \leq \alpha + 5 \\ \alpha - 5 - S & S \leq \alpha - 5 \end{cases}$$

hence

$$\begin{array}{ccc} & & 15 \\ & & V_1 \\ V & & 0 \\ & & V_{-1} \\ & & 15 \end{array}$$

add (i) and (ii)

$$\text{Payoff} = \max(S - \alpha - 5, 0) + \max(\alpha - 5 - S, 0) \equiv \text{(iii)}$$

$$\begin{array}{ccc} & & 15 \\ & & 15/2 \\ V = 15/2 & & 0 \\ & & 15/2 \\ & & 15 \end{array}$$

5. See spreadsheet

6. a) We define the return R_i on an asset S_i as

$$R_i = \frac{S_i - S_{i-1}}{S_{i-1}}$$

In the EWMA model, the variance rate (i.e. square of volatility) calculated for day n is a weighted average of the R_{n-i}^2 's ($i = 1, 2, 3, \dots$).

For some constant λ ($0 < \lambda < 1$) the weight given to R_{n-i-1}^2 is λ times the weight given to R_{n-i}^2 . The volatility estimated for day n , σ_n is related to the volatility estimated for day $(n-1)$, σ_{n-1} , by

$$\sigma_n^2 = \lambda \sigma_{n-1}^2 + \frac{(1-\lambda)}{\delta t} R_{n-1}^2.$$

We note a very attractive property of the EWMA model. To calculate the volatility estimate for day n , it is sufficient to know the volatility estimate for $(n-1)$ and σ_{n-1} .

b) In reducing λ from 0.95 to 0.85, we are putting more weight on recent observations of R_i^2 and less weight is given to older observations. Volatilities calculated with $\lambda = 0.85$ will react more quickly to new information and will move around much more than volatilities calculated with $\lambda = 0.95$.

7. Find the Fourier transform $\widehat{f}(\omega)$ of

$$f(x) = \begin{cases} 1/2\epsilon & |x| \leq \epsilon \\ 0 & |x| > \epsilon \end{cases}$$

and show that

$$\lim_{\epsilon \rightarrow 0^+} \widehat{f}(\omega) \rightarrow \frac{1}{\sqrt{2\pi}}$$

$$\begin{aligned} \widehat{f}(\omega) &= \mathcal{F}(f) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} f(x) e^{ix\omega} dx \quad (\text{ignore } 1/\sqrt{2\pi} \text{ for time being}) \\ &= \frac{1}{2\epsilon} \int_{-\epsilon}^{\epsilon} e^{ix\omega} dx = \frac{1}{2\epsilon} \left. \frac{e^{ix\omega}}{i\omega} \right|_{-\epsilon}^{\epsilon} = \frac{1}{2\epsilon} \frac{(e^{i\epsilon\omega} - e^{-i\epsilon\omega})}{i\omega} \\ &= \frac{1}{2\epsilon} \cdot \frac{2}{\omega} \cdot \frac{(e^{i\epsilon\omega} - e^{-i\epsilon\omega})}{2i} = \frac{1}{\omega\epsilon} \cdot \sin(\epsilon\omega) \\ \widehat{f}(\omega) &= \frac{1}{\sqrt{2\pi}} \frac{1}{\omega\epsilon} \cdot \sin(\epsilon\omega) \end{aligned}$$

$$\lim_{\epsilon \rightarrow 0^+} \frac{1}{\sqrt{2\pi}} \frac{\sin(\epsilon\omega)}{\omega\epsilon}$$

using L'Hospital's rule

$$\begin{aligned} \lim_{\epsilon \rightarrow 0^+} \frac{1}{\sqrt{2\pi}} \frac{\omega \cos(\epsilon\omega)}{\omega} &= \lim_{\epsilon \rightarrow 0^+} \frac{1}{\sqrt{2\pi}} \cos(\epsilon\omega) \\ &\rightarrow \frac{1}{\sqrt{2\pi}} \end{aligned}$$

8. Evaluate $\int_1^4 (x-1)^2 (4-x)^3 dx$ using the Beta function. **Hint:** consider a change of variable.

$$I = \int_1^4 (x-1)^2 (4-x)^3 dx$$

let

$$x = 3u + 1 \rightarrow dx = 3du$$

the limits $x = 1$ and 4 become in turn $u = 0$ and 1 .

$$\begin{aligned} I &= \int_0^1 (3u)^2 (3-3u)^3 \cdot 3du \\ &= 3^6 \int_0^1 u^2 (1-u)^3 du = 3^6 B(3, 4) \\ &= 3^6 \frac{\Gamma(3) \Gamma(4)}{\Gamma(7)} = 3^6 \frac{2!3!}{6!} = \frac{3^5}{20} \\ &= \frac{243}{20} \end{aligned}$$

9. Identify any singular points and classify them (you are not expected to solve any equation). Start by writing each in standard form $y'' + p(x)y' + q(x)y = 0$.

(i) $x^2y'' + 2xy' + y = 0$

$$y'' + \frac{2}{x}y' + \frac{1}{x^2}y = 0$$

$x = 0$ is a singular point. Checking $x \cdot \frac{2}{x}$ and $x^2 \cdot \frac{1}{x^2}$ tells us that the singularity is a regular singular point

(ii) $x^2y'' + xy' + (x^2 - 4)y = 0$

$$y'' + \frac{1}{x}y' + \frac{(x^2 - 4)}{x^2}y = 0$$

$x = 0$ is a singular point. $x \cdot \frac{1}{x}$ and $x^2 \cdot \frac{(x^2 - 4)}{x^2}$ removes the singularity hence we have a regular singular point.

(iii) $xy'' + x^2y' + y = 0$

$$y'' + xy' + \frac{1}{x}y = 0$$

$x = 0$ is a singular point. $x^2 \cdot \frac{1}{x}$ has a Taylor series expansion (TSE), so $x = 0$ is a regular singular point.

(iv) $(1 - x^2)^2 y'' + xy' + y = 0$

$$y'' + \frac{x}{(1 - x^2)^2}y' + \frac{1}{(1 - x^2)^2}y = 0$$

$x = \pm 1$ is a singular point. Recall that if x_0 is a singularity then we classify according to the existence of TSE of

$$(x - x_0)p(x) \text{ and } (x - x_0)^2 q(x)$$

Rewriting the the equation as

$$y'' + \frac{x}{(1 - x)^2(1 + x)^2}y' + \frac{1}{(1 - x)^2(1 + x)^2}y = 0$$

So at $x_0 = 1$

$$(x - 1)p(x) = (x - 1) \frac{x}{(1 + x)^2(1 - x)^2}$$

which does not have a TSE at $x = 1$ hence it is irregular. At $x_0 = -1$

$$(x + 1)p(x) = (x + 1) \frac{x}{(1 + x)^2(1 - x)^2}$$

which is singular at -1 . Hence it is an irregular singular point. We could have checked $(x - x_0)^2 q(x)$, but the fact that the initial test failed confirms the nature of the singular point.

$$(v) \quad (x+5)y'' + x^4y = 0$$

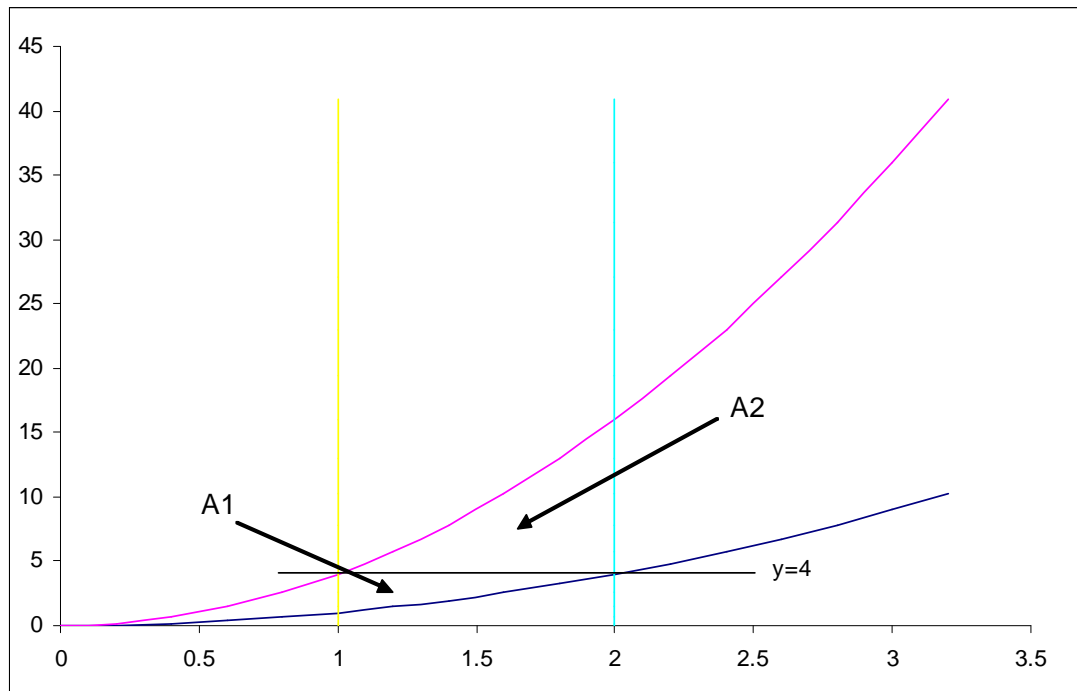
$$y'' + \frac{x^4}{(x+5)}y = 0$$

singular point at $x = -5$. Here $p(x)$ is zero. Since $(x+5)^2 \cdot \frac{x^4}{(x+5)}$ is $(x+5)x^4$ which is analytic and has a TSE about the point $x = 5$, it is a regular singular point.

10. (i) Consider the double integral

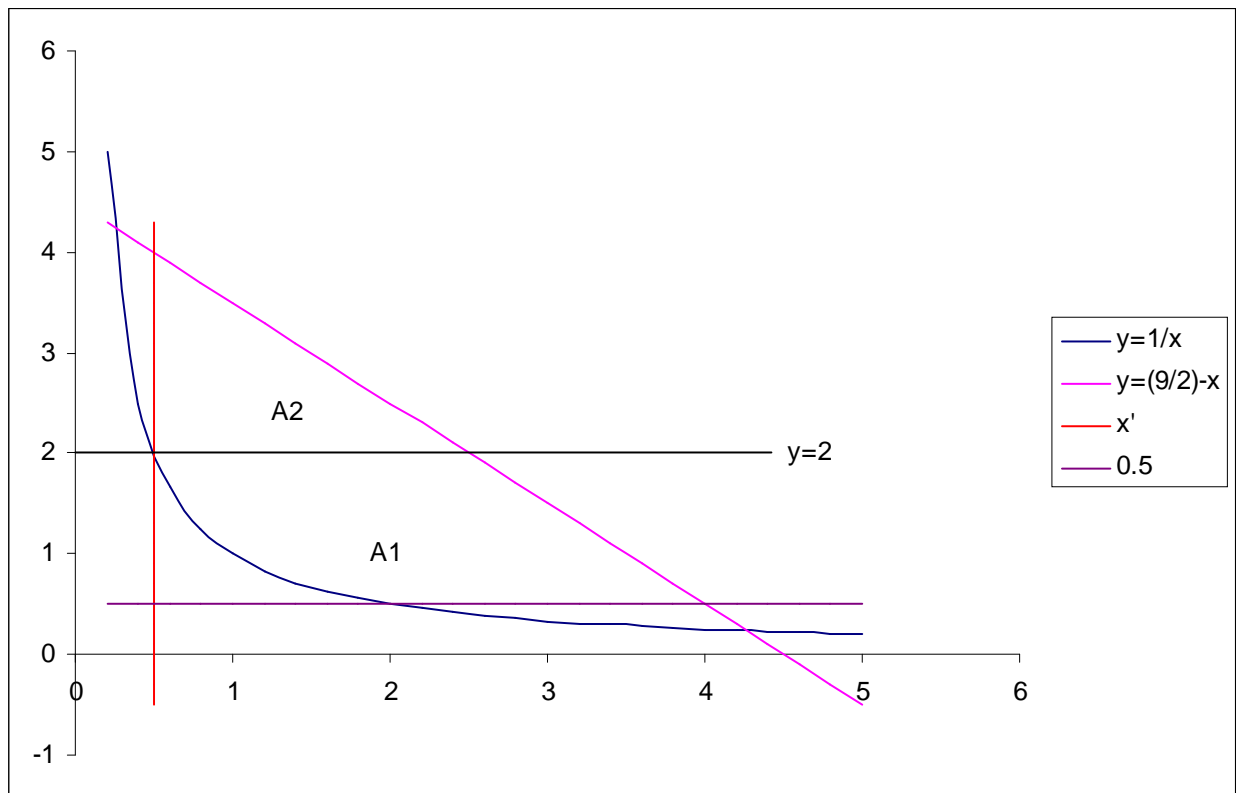
$$\int_1^2 \int_{x^2}^{4x^2} f(x, y) dy dx$$

where $f(x, y) = x + y$. By changing the order of integration, evaluate the integral. The region is drawn below. This is more of a mathematical exercise - normally you would never reverse the integral in this case as it introduces a lot more unnecessary work. Call the complete area bounded by the four lines $A = A_1 + A_2$. The line $y = 4$ is only drawn in to split the region in to two:



$$\iint_A = \iint_{A_1} + \iint_{A_2}$$

$$\begin{aligned}
\int_1^2 \int_{x^2}^{4x^2} f(x, y) dy dx &= \iint_{A_1} f(x, y) dx dy + \iint_{A_2} f(x, y) dx dy \\
&= \int_1^4 \int_1^{\sqrt{y}} (x + y) dx dy + \int_4^{16} \int_{\sqrt{y}/2}^2 (x + y) dx dy \\
&= \int_1^4 \left. \frac{1}{2}x^2 + xy \right|_1^{\sqrt{y}} dy + \int_4^{16} \left. \frac{1}{2}x^2 + xy \right|_{\sqrt{y}/2}^2 dy \\
&= \int_1^4 \left(y^{3/2} - \frac{1}{2}y - \frac{1}{2} \right) dy + \int_4^{16} \left(2 - \frac{1}{2}y^{3/2} + \frac{15}{8}y \right) dy \\
&= 231/4
\end{aligned}$$



(ii) Using double integration, calculate the area bounded by

$$y = \frac{1}{x}, y = \frac{9}{2} - x, x = \frac{1}{2}, y = \frac{1}{2}.$$

There are a number of ways to tackle this. Here we draw the line $y = 2$ in, to split the bounded region into A_1 and A_2

$$\begin{aligned}
\iint_A &= \iint_{A_1} dx dy + \iint_{A_2} dx dy \\
&= \int_{1/2}^2 \int_{1/y}^{9/2-y} dx dy + \int_2^4 \int_{1/2}^{9/2-y} dx dy \\
&= \int_{1/2}^2 x|_{1/y}^{9/2-y} dy + \int_2^4 x|_{1/2}^{9/2-y} dy \\
&= \int_{1/2}^2 \left(\frac{9}{2} - y - 1/y \right) dy + \\
&\quad \int_2^4 (4 - y) dy \\
&= \frac{9}{2}y - \frac{1}{2}y^2 - \ln y \Big|_{1/2}^2 + 4y - \frac{1}{2}y^2 \Big|_2^4 \\
&= \frac{55}{8} - 2 \ln 2
\end{aligned}$$

- (iii) Calculate the area of the circle $x^2 + y^2 = 4y$ between $\theta = \pi/3$ and $\theta = \pi/4$ **Hint:** use plane polars in (iii)

Completing the square on $x^2 + y^2 = 4y$ gives $x^2 + (y - 2)^2 = 2^2$ which is a circle centre $(0, 2)$ radius 2. This is the region defined by A

$$\iint_A dx dy = \iint_A r dr d\theta$$

where $x = r \cos \theta$, $y = r \sin \theta$ which gives $r^2 \cos^2 \theta + r^2 \sin^2 \theta = 4r \sin \theta$, i.e. $r = 0$ to $r = 4 \sin \theta$. The θ limits go from $\pi/4$ to $\pi/3$.

$$\begin{aligned} \int_{\pi/4}^{\pi/3} \int_0^{4 \sin \theta} r dr d\theta &= \int_{\pi/4}^{\pi/3} \left. \frac{r^2}{2} \right|_0^{4 \sin \theta} d\theta = \int_{\pi/4}^{\pi/3} 8 \sin^2 \theta d\theta \\ &= 4 \int_{\pi/4}^{\pi/3} (1 - \cos 2\theta) d\theta \\ &= 4 \left[\theta - \frac{1}{2} \sin 2\theta \right]_{\pi/4}^{\pi/3} \\ &= 2 - \sqrt{3} + \frac{\pi}{3} \end{aligned}$$

11. Consider the stochastic process $Y(t)$ satisfying the SDE

$$dY(t) = f(t)dt + g(t)dX(t), \quad Y(0) = Y_0 \quad (1)$$

where $f(t)$ and $g(t)$ are two time-dependent functions and $X(t)$ is a standard Brownian motion.

How should we choose $f(t)$ if we want the process $Z(t) = e^{Y(t)}$ to be an exponential martingale?

Consider the process $Z(t) = e^{Y(t)}$. Applying Itô to the function $F(y) = e^y$ and the process $Y(t)$ given in (1), we obtain:

$$\begin{aligned} dZ(t) &= de^{Y(t)} \\ &= \frac{dF}{dy} (f(t)dt + g(t)dX(t)) + \frac{1}{2} \frac{d^2F}{dy^2} g^2(t)dt \\ &= e^{Y(t)} \left(f(t) + \frac{1}{2}g^2(t) \right) dt + e^{Y(t)}g(t)dX(t) \end{aligned}$$

$Z(t)$ is a martingale iff it is a driftless process, and therefore for $Z(t)$ to be a martingale we must have

$$e^{Y(t)} \left(f(t) + \frac{1}{2}g^2(t) \right) = 0$$

This is only possible if

$$f(t) = -\frac{1}{2}g^2(t).$$

12. Consider the function $m_n(t)$ defined as

$$m_n(t) = \mathbb{E}[X^n(t)], \quad n = 1, 2, \dots \quad (2)$$

where $X(t)$ is a standard Brownian motion.

Applying Itô's lemma, show that:

$$m_n(t) = \frac{1}{2}n(n-1) \int_0^t m_{n-2}(t), \quad n = 2, 3, \dots \quad (*)$$

Deduce from (*) that

$$m_4(t) = 3t^2$$

and compute $m_6(t)$.

Because of the expectation, we cannot tackle expression (2) up-front.

Consider instead the auxiliary function $g_n(t, x) = x^n$ for $n \geq 2$. Note the relation between $g_n(t, x)$ and $m_n(t)$:

$$m_n(t) = \mathbf{E}[g_n(t, X(t))]$$

Applying Itô's lemma to the function g_n and the standard Brownian motion, we get

$$\begin{aligned} g_n(t) &= g_n(0) + \int_0^t \frac{\partial g_n}{\partial s} ds + \int_0^t \frac{\partial g_n}{\partial x} dX(s) + \frac{1}{2} \int_0^t \frac{\partial^2 g_n}{\partial x^2} ds \\ &= n \int_0^t X^{n-1}(s) dX(s) + \frac{1}{2} n(n-1) \int_0^t X^{n-2}(s) ds \end{aligned}$$

since $g_n(0) = 0$

Take expectation on both sides to get:

$$\begin{aligned} m_n(t) &= \mathbf{E}[g_n(t, X(t))] \\ &= n \mathbf{E} \left[\int_0^t X^{n-1}(s) dX(s) + \frac{1}{2} n(n-1) \int_0^t X^{n-2}(s) ds \right] \end{aligned}$$

By linearity of expectation,

$$m_n(t) = n \mathbf{E} \left[\int_0^t X^{n-1}(s) dX(s) \right] + \frac{1}{2} n(n-1) \mathbf{E} \left[\int_0^t X^{n-2}(s) ds \right]$$

Recall that $\int_0^t X^{n-1}(s) dX(s)$ is an Itô integral and it is therefore a martingale, so $\mathbf{E}[\int_0^t X^{n-1}(s) dX(s)] = 0$.

Interchanging the order of integration to take the expectation inside the integral (that's Fubini's theorem), we finally get

$$\begin{aligned} m_n(t) &= \frac{1}{2} n(n-1) \int_0^t \mathbf{E}[X^{n-2}(s)] ds \\ &= \frac{1}{2} n(n-1) \int_0^t m_{n-2}(s) ds \end{aligned}$$

Now let's apply this formula for $n = 4$:

$$\begin{aligned} m_4(t) &= 6 \int_0^t \mathbf{E} [X^2(s)] ds \\ &= 6 \int_0^t s ds \\ &= 3t^2 \end{aligned}$$

What about $n = 6$?

$$\begin{aligned} m_6(t) &= 15 \int_0^t \mathbf{E} [X^4(s)] ds \\ &= 45 \int_0^t s^2 ds \\ &= 15t^3 \end{aligned}$$