

## 15.455x Sample Exam Questions

These sample exam problems are each worth 24 points. All sub-parts are weighted equally.

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1. Suppose that an asset price  $S_t$  follows a lognormal, continuous-time stochastic process,

$$dS = \mu S dt + \sigma S dB,$$

where  $\mu, \sigma$  are constants and  $B$  is a standard Brownian motion. Use Itô's lemma to find stochastic differential equations expressing  $dV$  in terms of  $dt$  and  $dB$  for the following functions  $V(S, t)$ . Are they Itô processes?

- (a)  $V = \alpha S + \beta$ ,
- (b)  $V = S^\gamma$ ,
- (c)  $V = e^{r(T-t)}S$ ,

where  $\alpha, \beta, \gamma, r$ , and  $T$  are constants.

**Solution:** These are all Itô processes.

(a)

$$dV = \alpha dS = (\alpha\mu S) dt + (\alpha\sigma S) dB_t,$$

(b)

$$\begin{aligned} dV &= dS^\gamma = \frac{(\sigma S)^2}{2} \gamma(\gamma - 1) S^{\gamma-2} dt + \gamma S^{\gamma-1} dS \\ &= \gamma S^\gamma \left( (\gamma - 1) \frac{\sigma^2}{2} dt + \frac{dS}{S} \right) \\ &= \gamma V \left( \left[ \mu + (\gamma - 1) \frac{\sigma^2}{2} \right] dt + \sigma dB_t \right), \\ \frac{dV}{V} &= \gamma \left( \mu + (\gamma - 1) \frac{\sigma^2}{2} \right) dt + (\gamma\sigma) dB_t. \end{aligned}$$

(c)

$$\begin{aligned} dV &= -re^{r(T-t)} S dt + e^{r(T-t)} dS \\ &= e^{r(T-t)} S \left( -r dt + \frac{dS}{S} \right) \\ &= (\mu - r)V dt + \sigma V dB_t, \\ \frac{dV}{V} &= (\mu - r) dt + \sigma dB_t. \end{aligned}$$

2. Let a stationary discrete-time stochastic process  $x_t$  be given by

$$x_t = A + Bx_{t-2} + C_t,$$

where  $z_t \sim \mathcal{N}(0, 1)$  is an IID Gaussian white-noise process, and  $A, B, C$  are constants.

- (a) What is the unconditional mean of the process  $x_t$ ?
- (b) An analyst decides to construct a forecast  $f_\tau$  for future values of the process by taking its expected value, conditional on information available up through the time  $t$  when the forecast is made. That is,

$$f_\tau \equiv E_t[x_\tau | x_t, x_{t-1}, \dots], \quad \tau > t.$$

Let  $A = 0.1$ ,  $B = 0.2$ ,  $C = 0.3$ , and suppose that two recent values  $x_1 = 0.4$ ,  $x_2 = 0.5$  have just been observed. What are the one-step-ahead and two-step-ahead forecasts? That is, at time  $t = 2$ , what are the forecasts  $f_3$  and  $f_4$ ? What is the variance of the forecasts?

**Solution:**

(a) Taking expectations and using stationarity,

$$\begin{aligned}\mu &\equiv E[x_t] = A + BE[x_{t-2}] + CE[\epsilon_t] = A + B\mu, \\ \mu &= \frac{A}{1-B}.\end{aligned}$$

(b) At  $t = 2$ , all returns  $x_t$  and noise terms  $\epsilon_t$  are unknown for  $t > 2$  but known for  $t \leq 2$ . From the return equation for  $r_t$ , we have

$$\begin{aligned}x_3 &= A + Bx_1 + C\epsilon_3, \\ E[x_3|x_1, x_2] &= A + Bx_1 = 0.1 + (0.2)(0.4) = 0.18, \\ x_4 &= A + Bx_2 + C\epsilon_4, \\ E[x_4|x_1, x_2] &= A + Bx_2 = 0.1 + (0.2)(0.5) = 0.2, \\ \text{Var}(x_3) &= E[(x_3 - A - Bx_1)^2] = E[(C\epsilon_3)^2] = C^2 = 0.09, \\ \text{Var}(x_4) &= E[(x_4 - A - Bx_2)^2] = E[(C\epsilon_4)^2] = C^2 = 0.09.\end{aligned}$$

where the expectations are taken at  $t = 2$ .

3. (a) Consider the quadratic form defined by

$$Q(x, y) = 2x^2 + 12xy - 7y^2.$$

Using Lagrange multipliers, find the location and value of the extrema of  $Q$  subject to the constraint  $x + 3y = 5$ . Determine whether each solution is a maximum, minimum, or neither.

- (b) Two assets have correlation  $\rho$ , and their volatilities are  $2\sigma$  and  $\sigma$  respectively. What are the weights of a minimum-variance, fully-invested portfolio of the two assets, and what is its risk? That is, minimize the portfolio variance  $\sigma_p^2 = \mathbf{w}^\top C \mathbf{w}$ , where  $C$  is the covariance matrix and  $\mathbf{w}$  is an asset weight vector whose components satisfy the budget constraint  $w_1 + w_2 = 1$ .
- (c) In the problem above, for what values of  $\rho$  and  $\sigma$  will the solution also satisfy an inequality constraint  $0 \leq w_i \leq 1$ ? (That is, the optimal portfolio is also unlevered and long-only.)

**Solution:**

- (a) To solve along the line  $x + 3y = 5$  using Lagrange multipliers, extremize

$$\begin{aligned}\mathcal{L} &= Q(x, y) - \gamma(x + 3y - 5) \\ &= 2x^2 + 12xy - 7y^2 - \gamma(x + 3y - 5).\end{aligned}$$

Then differentiating,

$$\begin{aligned}\frac{\partial \mathcal{L}}{\partial x} &= 4x + 12y - \gamma = 0, \\ \frac{\partial \mathcal{L}}{\partial y} &= 12x - 14y - 3\gamma = 0, \\ \frac{\partial \mathcal{L}}{\partial \gamma} &= x + 3y - 5 = 0.\end{aligned}$$

Eliminating  $\gamma$  by subtracting the second equation from 3 times the first gives

$$y = 0, \quad x = 5$$

with solution

$$Q(5, 0) = 50$$

This single critical point is a maximum of  $Q$  along the line. Since there is only one critical point, this can be quickly checked by evaluation  $Q$  at any other point along the line, such as  $Q(-1, 2) = -50 < Q_{\max}$

Alternatively, if one substitutes  $y = t$ ,  $x = 5 - 3t$  along the line, then  $Q = -25t^2 + 50$ , an unconstrained function of a single variable whose single maximum is clearly located at  $t = 0$ .

- (b) In components, we have

$$\mathcal{L} = 4\sigma^2 w_1^2 + \sigma^2 w_2^2 + 4\rho\sigma^2 w_1 w_2 - \gamma(w_1 + w_2 - 1)$$

which is extremized for

$$\begin{aligned}\frac{\partial \mathcal{L}}{\partial w_1} &= 8\sigma^2 w_1 + 4\rho\sigma^2 w_2 - \gamma = 0, \\ \frac{\partial \mathcal{L}}{\partial w_2} &= 4\rho\sigma^2 w_1 + 2\sigma^2 w_2 - \gamma = 0, \\ \frac{\partial \mathcal{L}}{\partial \gamma} &= w_1 + w_2 - 1 = 0.\end{aligned}$$

Subtracting the first two equations eliminates  $\gamma$  and gives

$$(8 - 4\rho)w_1 + (4\rho - 2)w_2 = 0,$$

so

$$w_2 = \frac{4 - 2\rho}{1 - 2\rho} w_1.$$

Substituting into the constraint finally gives

$$\begin{pmatrix} w_1 \\ w_2 \end{pmatrix} = \frac{1}{5-4\rho} \begin{pmatrix} 1-2\rho \\ 4-2\rho \end{pmatrix}$$

along with

$$\begin{aligned} \sigma_p^2 &= \frac{\sigma^2}{(5-4\rho)^2} [4(1-2\rho)^2 + (4-2\rho)^2 + 4\rho(1-2\rho)(4-2\rho)] \\ &= \frac{20-16\rho-20\rho^2+16\rho^3}{(5-4\rho)^2} \sigma^2 \\ &= \frac{4-4\rho^2}{5-4\rho} \sigma^2. \end{aligned}$$

- (c) The weights  $w_i$  are independent of  $\sigma$ . Since the correlation is bounded,  $-1 \leq \rho \leq 1$ ,  $w_2$  is always non-negative. However  $w_1$  changes sign when  $\rho \rightarrow 1/2$ , so the portfolio is unlevered and long-only provided that the correlation  $-1 \leq \rho < 1/2$ .

4. The returns on a set of  $N$  assets are believed to follow the mean-reverting process

$$R_{it} - \mu_i = -\lambda(R_{i(t-1)} - \mu_i) + \sigma_i z_{it},$$

where  $\mu_i, \sigma_i, \lambda$  are constants,  $i = 1, \dots, N$ ;  $|\lambda| < 1$ ; and

$$\mathbb{E}[z_{it}] = 0; \quad \mathbb{E}[z_{it}z_{js}] = \begin{cases} 1 & \text{if } t = s \text{ and } i = j, \\ 0 & \text{if } t \neq s \text{ or } i \neq j; \end{cases}$$

A market-neutral long/short trading strategy attempts to profit by investing capital in weights assigned according to

$$w_{it} = -\frac{1}{N}(R_{it} - \bar{R}_t),$$

where the market average return is defined by

$$\bar{R}_t = \frac{1}{N} \sum_{i=1}^N R_{it}.$$

Assume there are no transaction costs and the risk-free rate  $R_f = 0$ . Find the expected portfolio return

$$\mathbb{E}[R_p] = \mathbb{E} \left[ \sum_i w_{i(t-1)} R_{it} \right]$$

in terms of the parameters given. Under what conditions is this expected return positive?



**Solution:** We consider the more general case of using lag- $k$  returns as the weights. The time series of daily *portfolio returns*,  $\pi_t$  is given by

$$\pi_t(k) = \sum_{i \in U} w_{i(t-k)} R_{it} . \quad (1)$$

This depends on the lag parameter  $k$ , so the answer should demonstrate the  $k$ -dependence of the strategy.

Notice that the weights  $w$  depend on the the market average  $\bar{R}_t$ , which in turn depends on all of the stocks. In the most general case, stocks could be cross-correlated and the result requires a full cross-covariance matrix  $\Gamma_k$ , whose  $(ij)$  matrix elements are  $\text{Cov}(R_{it}, R_{j,(t-k)})$ . Each  $R_i$  could be multiplied by every  $R_j$ .

In the present case, each stock is correlated only with its own lagged returns, so things are simpler.  $\Gamma_k$  is a diagonal matrix and everything can be written as a sum of independent AR(1) autocovariances  $\gamma_k(i)$ .

The zeroth-order autocovariance is just the variance itself,

$$\begin{aligned} \gamma_0 &= \text{Var}[R_t] = \text{E} [(R_t - \mu)^2] \\ &= \lambda^2 \text{E} [(R_{t-1} - \mu)^2] + \text{E} [\epsilon_t^2] \\ &= \lambda^2 \gamma_0 + \sigma^2, \end{aligned}$$

so that

$$\gamma_0 = \frac{\sigma^2}{1 - \lambda^2}.$$

The higher order autocovariances can be obtained by recursion.

$$\begin{aligned} \gamma_k &= \text{E} [(R_t - \mu)(R_{t-k} - \mu)] \\ &= -\lambda \text{E} [(R_{t-1} - \mu)(R_{t-k} - \mu)] \\ &= -\lambda \gamma_{k-1}, \end{aligned}$$

so that

$$\gamma_k = (-\lambda)^k \gamma_0 = \frac{(-\lambda)^k}{1 - \lambda^2} \sigma^2.$$

Now we can use this result to compute the closed-form expectation of the strategy P/L.

$$\begin{aligned} \text{E} [\pi_t(k)] &= \frac{1 - N}{N^2} \sum_{i=1}^N \gamma_k(i) - \frac{1}{N} \sum_{i=1}^N (\mu_i - \bar{\mu})^2 \\ &= -\frac{(-\lambda)^k}{1 - \lambda^2} \left(1 - \frac{1}{N}\right) \left(\frac{1}{N} \sum_{i=1}^N \sigma_i^2\right) - \frac{1}{N} \sum_{i=1}^N (\mu_i - \bar{\mu})^2. \end{aligned}$$

The last term, which is always negative, represents dispersion of the means. It is independent of  $k$  and vanishes only if all the mean returns are equal. The first term alternates in sign, so that only odd lags contribute positive P/L, and decreases in magnitude with  $k$ .

5. Two stocks have prices  $S_1$  and  $S_2$  that follow geometric Brownian motion with the same stochastic process  $dB$ :

$$dS_1 = \mu_1 S_1 dt + \sigma_1 S_1 dB,$$

$$dS_2 = \mu_2 S_2 dt + \sigma_2 S_2 dB.$$

- (a) A contract has value  $V = S_1 S_2$ . You can show that  $V$  also follows geometric Brownian motion. What are its drift and volatility parameters?
- (b) What is the process followed by  $1/V$ ?
- (c) A call option on  $V$  with strike  $K$  has value  $C(t, V)$  and payoff at expiration  $\max(S_1 S_2 - K, 0)$ . What PDE does the option satisfy?

**Solution:** It is convenient to write the series by dividing through such that the right-hand sides are Itô processes with constant coefficients:

$$\begin{aligned}\frac{dS_1}{S_1} &= \mu_1 dt + \sigma_1 dB, \\ \frac{dS_2}{S_2} &= \mu_2 dt + \sigma_2 dB.\end{aligned}$$

(a) Applying the product rule to  $V = S_1 S_2$ ,

$$dV = S_1 dS_2 + S_2 dS_1 + dS_1 dS_2,$$

so that

$$\begin{aligned}\frac{dV}{V} &= \frac{dS_1}{S_1} + \frac{dS_2}{S_2} + \frac{dS_1}{S_1} \cdot \frac{dS_2}{S_2} \\ &= (\mu_1 + \mu_2) dt + (\sigma_1 + \sigma_2) dB + \sigma_1 \sigma_2 (dB)^2\end{aligned}$$

where we can replace  $(dB)^2 \rightarrow dt$  and write

$$\frac{dV}{V} = (\mu_1 + \mu_2 + \sigma_1 \sigma_2) dt + (\sigma_1 + \sigma_2) dB.$$

This standard form for geometric Brownian motion lets us read off that the drift and volatility parameters are

$$\begin{aligned}\mu_V &= \mu_1 + \mu_2 + \sigma_1 \sigma_2, \\ \sigma_V &= \sigma_1 + \sigma_2.\end{aligned}$$

(b) For  $F(t, V) = 1/V$ , apply Itô's formula:

$$\begin{aligned}dF &= d\left(\frac{1}{V}\right) = \frac{-1}{V^2} dV + \frac{b^2}{2} \left(\frac{2}{V^3}\right) dt, \\ \frac{dF}{F} &= -\frac{dV}{V} + \frac{(\sigma_1 + \sigma_2)^2 V^2}{2} \left(\frac{2}{V^2}\right) dt \\ &= [(\sigma_1 + \sigma_2)^2 - (\mu_1 + \mu_2 + \sigma_1 \sigma_2)] dt - (\sigma_1 + \sigma_2) dB.\end{aligned}$$

This is also in the form of a geometric Brownian motion.

(c) Since  $V$  follows a standard geometric Brownian motion, options with  $V$  as an underlying have values  $C(t, V)$  that satisfy the usual Black-Scholes PDE, with the appropriate volatility parameter for  $V$ :

$$\frac{\partial C}{\partial t} + \frac{(\sigma_1 + \sigma_2)^2}{2} V^2 \frac{\partial^2 C}{\partial V^2} + rV \frac{\partial C}{\partial V} - rC = 0.$$