

MATHEMATICAL FINANCE CHEAT SHEET

Normal Random Variables

A random variable X is Normal $\mathbf{N}(\mu, \sigma^2)$ (aka. *Gaussian*) under a measure \mathbf{P} if and only if

$$\mathbf{E}_{\mathbf{P}}[e^{\theta X}] = e^{\theta\mu + \frac{1}{2}\theta^2\sigma^2}, \quad \text{for all real } \theta.$$

A standard normal $Z \sim \mathbf{N}(0, 1)$ under a measure \mathbf{P} has density

$$\phi(x) = \frac{1}{\sqrt{2\pi}} e^{-x^2/2}. \quad \mathbf{P}[Z \leq x] = \Phi(x) := \int_{-\infty}^x \phi(z) dz.$$

Let $X = (X_1, X_2, \dots, X_n)'$ with $X_i \sim \mathbf{N}(\mu_i, q_{ii})$ and $\mathbf{Cov}[X_i, X_j] = q_{ij}$ for $i, j = 1, \dots, n$. We call $\mu := (\mu_1, \dots, \mu_n)'$ the *mean* and $Q := (q_{ij})_{i,j=1}^n$ the *covariance matrix* of X . Assume $\det Q > 0$, then X has a *multivariate normal distribution* if it has the density

$$\phi(x) = \frac{1}{\sqrt{(2\pi)^n \det Q}} \exp\left(-\frac{1}{2}(x - \mu)'Q^{-1}(x - \mu)\right), \quad x \in \mathbf{R}^n.$$

We write $X \sim \mathbf{N}(\mu, Q)$ if this is the case. Alternatively, $X \sim \mathbf{N}(\mu, Q)$ under \mathbf{P} if and only if

$$\mathbf{E}_{\mathbf{P}}[e^{\theta'X}] = \exp\left(\theta'\mu + \frac{1}{2}\theta'Q\theta\right), \quad \text{for all } \theta \in \mathbf{R}^n.$$

If $Z \sim \mathbf{N}(0, Q)$ and $c \in \mathbf{R}^n$ then $X = c'Z \sim \mathbf{N}(0, c'Qc)$. If $C \in \mathbf{R}^{m \times n}$ (i.e., $m \times n$ matrix) then $X = CZ \sim \mathbf{N}(0, CQC')$ and CQC' is a $m \times m$ covariance matrix.

Gaussian Shifts

If $Z \sim \mathbf{N}(0, 1)$ under a measure \mathbf{P} , h is an integrable function, and c is a constant then

$$\mathbf{E}_{\mathbf{P}}[e^{cZ}h(Z)] = e^{c^2/2}\mathbf{E}_{\mathbf{P}}[h(Z+c)].$$

Let $X \sim \mathbf{N}(0, Q)$, h be a integrable function of $x \in \mathbf{R}^n$, and $c \in \mathbf{R}^n$. Then

$$\mathbf{E}_{\mathbf{P}}[e^{c'X}h(X)] = e^{\frac{1}{2}c'Qc}\mathbf{E}_{\mathbf{P}}[h(X+c)].$$

Correlating Brownian Motions

Let $(W(t))_{t \geq 0}$ and $(\widetilde{W}(t))_{t \geq 0}$ be independent Brownian motions. Given a correlation coefficient $\rho \in [-1, 1]$, define

$$\widehat{W}(t) := \rho W(t) + \sqrt{1 - \rho^2} \widetilde{W}(t),$$

then $(\widehat{W}(t))_{t \geq 0}$ is a Brownian motion and $\mathbf{E}[W(t)\widehat{W}(t)] = \rho t$.

Identifying Martingales

If $X_t = X(t)$ is a diffusion process satisfying

$$dX(t) = \mu(t, X_t)dt + \sigma(t, X_t)dW(t)$$

and $\mathbf{E}_{\mathbf{P}}[(\int_0^T \sigma(s, X_s)^2 ds)^{1/2}] < \infty$ (or, $\sigma(t, x) \leq c|x|$ as $|x| \rightarrow \infty$), then

X is a martingale $\iff X$ is driftless (i.e., $\mu(t) \equiv 0$ with \mathbf{P} -prob. 1).

Novikov's Condition

In the case $dX(t) = \sigma(t)X(t)dW(t)$ for some \mathcal{F} -previsible process $(\sigma(t))_{t \geq 0}$, then we have the simpler condition

$$\mathbf{E}_{\mathbf{P}}\left[\exp\left(\frac{1}{2}\int_0^T \sigma(s)^2 ds\right)\right] < \infty \Rightarrow X \text{ is a martingale.}$$

Itô's Formula

For $X_t = X(t)$ given by $dX(t) = \mu(t)dt + \sigma(t)dW(t)$ and a function $g(t, x)$ that is twice differentiable in x and once in t . Then for $Y(t) = g(t, X_t)$, we have

$$dY(t) = \frac{\partial g}{\partial t}(t, X_t)dt + \frac{\partial g}{\partial x}(t, X_t)dX_t + \frac{1}{2}\sigma(t)^2 \frac{\partial^2 g}{\partial x^2}(t, X_t)dt.$$

The Product Rule

Given $X(t)$ and $Y(t)$ adapted to the same Brownian motion $(W(t))_{t \geq 0}$,

$$dX(t) = \mu(t)dt + \sigma(t)dW(t), \quad dY(t) = \nu(t)dt + \rho(t)dW(t).$$

Then $d(X(t)Y(t)) = X(t)dY(t) + Y(t)dX(t) + \underbrace{d\langle X, Y \rangle(t)}_{\sigma(t)\rho(t)dt}$.

In the other case, if $X(t)$ and $Y(t)$ are adapted to two different and independent Brownian motions $(W(t))_{t \geq 0}$ and $(\widetilde{W}(t))_{t \geq 0}$,

$$dX(t) = \mu(t)dt + \sigma(t)dW(t), \quad dY(t) = \nu(t)dt + \rho(t)d\widetilde{W}(t).$$

Then $d(X(t)Y(t)) = X(t)dY(t) + Y(t)dX(t)$ as $d\langle X, Y \rangle(t) = 0$.

Radon-Nikodým Derivative

Given \mathbf{P} and \mathbf{Q} equivalent measures and a time horizon T , we can define a random variable $\frac{d\mathbf{Q}}{d\mathbf{P}}$ defined on \mathbf{P} -possible paths, taking positive real values, such that

- $\mathbf{E}_{\mathbf{Q}}[X_T] = \mathbf{E}_{\mathbf{P}}\left[\frac{d\mathbf{Q}}{d\mathbf{P}}X_T\right]$, for all claims X_T knowable by time T ,
- $\mathbf{E}_{\mathbf{Q}}[X_t|\mathcal{F}_s] = \zeta_s^{-1}\mathbf{E}_{\mathbf{P}}[\zeta_t X_t|\mathcal{F}_s]$, for $s \leq t \leq T$,

where ζ_t is the process $\mathbf{E}_{\mathbf{P}}[\frac{d\mathbf{Q}}{d\mathbf{P}}|\mathcal{F}_t]$.

Cameron-Martin-Girsanov Theorem

If $(W(t))_{t \geq 0}$ is a \mathbf{P} -Brownian motion and $(\gamma(t))_{t \geq 0}$ is an \mathcal{F} -previsible process satisfying the boundedness condition $\mathbf{E}_{\mathbf{P}}\left[\exp\left(\frac{1}{2}\int_0^T \gamma(t)^2 dt\right)\right] < \infty$, then there exists a measure \mathbf{Q} such that:

- \mathbf{Q} is equivalent to \mathbf{P} ,
- $\frac{d\mathbf{Q}}{d\mathbf{P}} = \exp\left(-\int_0^T \gamma(t)dW(t) - \frac{1}{2}\int_0^T \gamma(t)^2 dt\right)$,
- $\widetilde{W}(t) := W(t) + \int_0^t \gamma(s)ds$ is a \mathbf{Q} -Brownian motion.

In other words, $W(t)$ is a drifting \mathbf{Q} -Brownian motion with drift $-\gamma(t)$ at time t .

Cameron-Martin-Girsanov Converse

If $(W(t))_{t \geq 0}$ is a \mathbf{P} -Brownian motion, and \mathbf{Q} is a measure equivalent to \mathbf{P} , then there exists a \mathcal{F} -previsible process $(\gamma(t))_{t \geq 0}$ such that

$$\widetilde{W}(t) := W(t) + \int_0^t \gamma(s)ds$$

is a \mathbf{Q} -Brownian motion. That is, $W(t)$ plus drift $\gamma(t)$ is a \mathbf{Q} -Brownian motion. Additionally,

$$\frac{d\mathbf{Q}}{d\mathbf{P}} = \exp\left(-\int_0^t \gamma(t)dW(t) - \frac{1}{2}\int_0^T \gamma(t)^2 dt\right).$$

Martingale Representation Theorem

Suppose $(M(t))_{t \geq 0}$ is a \mathbf{Q} -martingale process whose volatility $\sqrt{\mathbf{E}_{\mathbf{Q}}[M(t)^2]} = \sigma(t)$ satisfies $\sigma(t) \neq 0$ for all t (with \mathbf{Q} -probability one). Then if $(N(t))_{t \geq 0}$ is any other \mathbf{Q} -martingale, there exists an \mathcal{F} -previsible process $(\phi(t))_{t \geq 0}$ such that $\int_0^T \phi(t)^2 \sigma(t)^2 dt < \infty$ (with \mathbf{Q} -prob. one), and N can be written as

$$N(t) = N(0) + \int_0^t \phi(s)dM(s),$$

or in differential form, $dN(t) = \phi(t)dM(s)$. Further, ϕ is (essentially) unique.

Multidimensional Diffusions, Quadratic Covariation, and Itô's Formula

If $X := (X_1, X_2, \dots, X_n)'$ is a n -dimensional diffusion process with form

$$X(t) = X(0) + \int_0^t \mu(s)ds + \int_0^t \Sigma(s)dW(s),$$

where $\Sigma(t) \in \mathbf{R}^{n \times m}$ and W is a m -dimensional Brownian motion. The *quadrature covariation* of the components X_i and X_j is

$$\langle X_i, X_j \rangle(t) = \int_0^t \Sigma_i(s)' \Sigma_j(s)ds,$$

or in differential form $d\langle X_i, X_j \rangle(t) = \Sigma_i(t)' \Sigma_j(t)dt$, where $\Sigma_i(t)$ is the i^{th} column of $\Sigma(t)$. The *quadratic variation* of $X_i(t)$ is $\langle X_i \rangle(t) = \int_0^t \Sigma_i(s)' \Sigma_i(s)ds$. The *multi-dimensional Itô formula* for $Y(t) = f(t, X_1(t), \dots, X_n(t))$ is

$$dY(t) = \frac{\partial f}{\partial t}(t, X_1(t), \dots, X_n(t))dt + \sum_{i=1}^n \frac{\partial f}{\partial x_i}(t, X_1(t), \dots, X_n(t))dX_i(t) + \frac{1}{2} \sum_{i,j=1}^n \frac{\partial^2 f}{\partial x_i \partial x_j}(t, X_1(t), \dots, X_n(t))d\langle X_i, X_j \rangle(t).$$

The (*vector-valued*) *multi-dimensional Itô formula* for

$$Y(t) = f(t, X(t)) = (f_1(t, X(t)), \dots, f_n(t, X(t)))'$$

where $f_k(t, X) = f_k(t, X_1, \dots, X_n)$ and $Y(t) = (Y_1(t), Y_2(t), \dots, Y_n(t))'$ is given component-wise (for $k = 1, \dots, n$) as

$$dY_k(t) = \frac{\partial f_k(t, X(t))}{\partial t}dt + \sum_{i=1}^n \frac{\partial f_k(t, X(t))}{\partial x_i}dX_i(t) + \frac{1}{2} \sum_{i,j=1}^n \frac{\partial^2 f_k(t, X(t))}{\partial x_i \partial x_j}d\langle X_i, X_j \rangle(t).$$

Stochastic Exponential

The *stochastic exponential* of X is $\mathcal{E}_t(X) = \exp(X(t) - \frac{1}{2}\langle X \rangle(t))$. It satisfies

$$\mathcal{E}(0) = 1, \quad \mathcal{E}(X)\mathcal{E}(Y) = \mathcal{E}(X+Y)e^{(X,Y)}, \quad \mathcal{E}(X)^{-1} = \mathcal{E}(-X)e^{(X,X)}.$$

The process $Z = \mathcal{E}(X)$ is a positive process and solves the SDE

$$dZ = Z dX, \quad Z(0) = e^{X(0)}.$$

Solving Linear ODEs

The linear *ordinary differential equation*

$$\frac{dz(t)}{dt} = m(t) + \mu(t)z(t), \quad z(a) = \zeta,$$

for $a \leq t \leq b$ has solution given by

$$z(t) = \zeta \epsilon_t + \int_a^t \epsilon_t \epsilon_u^{-1} m(u)du, \quad \epsilon_t := \exp\left(\int_a^t \mu(u)du\right), \\ = \zeta \exp\left(\int_a^t \mu(u)du\right) + \int_a^t m(u) \exp\left(\int_u^t \mu(r)dr\right)du.$$

Solving Linear SDEs

The linear *stochastic differential equation*

$$dZ(t) = [m(t) + \mu(t)Z(t)]dt + [q(t) + \sigma(t)Z(t)]dW(t), \quad Z(a) = \zeta,$$

for $a \leq t \leq b$ has solution given by

$$Z(t) = \zeta \mathcal{E}_t + \int_a^t \mathcal{E}_t \mathcal{E}_u^{-1} [m(u) - q(u)\sigma(u)]du + \int_a^t \mathcal{E}_t \mathcal{E}_u^{-1} q(u)dW(u),$$

where $\mathcal{E}_t := \mathcal{E}_t(X)$ and $X(t) = \int_a^t \mu(u)du + \int_a^t \sigma(u)dW(u)$. In other words,

$$\mathcal{E}_t = \exp\left(\int_a^t \mu(u)du + \int_a^t \sigma(u)dW(u) - \frac{1}{2}\int_a^t \sigma(u)^2 du\right).$$

Fundamental Theorem of Asset Pricing

Let X be some \mathcal{F}_T -measurable claim, payable at time T . The arbitrage-free price \mathcal{V} of X at time t is

$$\mathcal{V}(t) = \mathbf{E}_{\mathbf{Q}} \left[\exp \left(- \int_t^T r(s) ds \right) X \middle| \mathcal{F}_t \right],$$

where \mathbf{Q} is the risk-neutral measure.

Market Price Of Risk

Let $X_t = X(t)$ be the price of a non-tradable asset with dynamics $dX(t) = \mu(t)dt + \sigma(t)dW(t)$ where $(\sigma(t))_{t \geq 0}$ and $(\mu(t))_{t \geq 0}$ are previsible processes and $(W(t))_{t \geq 0}$ is a \mathbf{P} -Brownian motion. Let $Y(t) := f(X_t)$ be the price of a tradable asset where $f: \mathbf{R} \rightarrow \mathbf{R}$ is a deterministic function. Then the *market price of risk* is

$$\gamma(t) := \frac{\mu_t f'(X_t) + \frac{1}{2} \sigma_t^2 f''(X_t) - r f(X_t)}{\sigma_t f'(X_t)},$$

and the behaviour of X_t under the risk-neutral measure \mathbf{Q} is given by

$$dX(t) = \sigma(t) d\widetilde{W}(t) + \frac{r f(X_t) - \frac{1}{2} \sigma_t^2 f''(X_t)}{f'(X_t)} dt.$$

Black's Model

Consider a European option with strike price K on a asset with value V_T at maturity time T . Let F_T be the forward price of V_T , F_0 the current forward price. If $\log V_T \sim \mathbf{N}(F_0, \sigma^2 T)$ then the Call and Put prices are given by

$$\mathcal{C} = P(0, T)(F_0 \Phi(d_1) - K \Phi(d_2)), \quad \mathcal{P} = P(0, T)(K \Phi(-d_2) - F_0 \Phi(-d_1)),$$

where $d_1 = \frac{\log(F_T/K) + \sigma^2 T/2}{\sigma \sqrt{T}}$ and $d_2 = d_1 - \sigma \sqrt{T}$.

Forward Rates, Short Rates, Yields, and Bond Prices

The *forward rate* at time t that applies between times T and S is defined as

$$F(t, T, S) = \frac{1}{S - T} \log \frac{P(t, T)}{P(t, S)}.$$

The *instantaneous forward rate* at time t is $f(t, T) = \lim_{S \rightarrow T} F(t, T, S)$. The *instantaneous risk-free rate* or *short rate* is $r(t) = \lim_{T \rightarrow t} f(t, T)$. The *cash account* is given by

$$B(t) = \exp \left(\int_0^t r(s) ds \right),$$

and satisfies $dB(t) = r(t)B(t)dt$ with $B(0) = 1$. The instantaneous forward rates and the yield can be written in terms of the bond prices as

$$f(t, T) = -\frac{\partial}{\partial T} \log P(t, T), \quad R(t, T) = -\frac{\log P(t, T)}{T - t}.$$

Conversely,

$$P(t, T) = \exp \left(- \int_t^T f(t, u) du \right) \quad \text{and} \quad P(t, T) = \exp(-(T - t)R(t, T)).$$

Short Rate and No-Arbitrage Models

The short-rate $r(t)$ follows a process of the form

$$dr(t) = a(t, r(t))dt + b(t, r(t))dW(t),$$

where $a(t, r)$ and $b(t, r)$ are chosen following:

Model	$a(t, r)$	$b(t, r)$	P?	MR?	CF?
Merton	μ	σ	N	N	Y
Dothan	$\mu r(t)$	$\sigma r(t)$	Y	N	Y
Vasicek	$\alpha(\mu - r(t))$	σ	N	Y	Y
CIR	$\alpha(\mu - r(t))$	$\sigma \sqrt{r(t)}$	Y	Y	Y
Pearson-Sun	$\alpha(\mu - r(t))$	$\sigma \sqrt{r(t) - \beta}$	Y	Y	Y
Ho & Lee	$\theta(t)$	σ	N	N	Y
Hull & White	$\alpha(\mu(t) - r(t))$	σ	N	Y	Y
Extended Vasicek	$\alpha(t)(\mu(t) - r(t))$	$\sigma(t)$	N	Y	Y
Black-Karasinski	$\alpha r(t)(\bar{\mu}(t) - \ln r(t))$	$\sigma(t)r(t)$	Y	Y	N

P means the process stays positive, MR means r_t is mean-reverting, and CF means that a closed-form solution exists for bond prices and for European put and call options.

Bond Pricing for Affine Models

Given an affine short-rate model

$$dr(t) = (b(t) + \beta(t)r(t))dt + \sqrt{a(t) + \alpha(t)r(t)}d\widetilde{W}(t),$$

the zero-coupon T -bond prices at time t have the form

$$P(t, T) = \exp(A(t, T) - B(t, T)r(t)),$$

where the functions A and B satisfy the system of ODEs:

$$\begin{aligned} \frac{d}{dt} A(t, T) &= -\frac{1}{2} a(t) B^2(t, T) + b(t) B(t, T), & A(T, T) &= 0, \\ \frac{d}{dt} B(t, T) &= \frac{1}{2} \alpha(t) B^2(t, T) - \beta(t) B(t, T) - 1, & B(T, T) &= 0. \end{aligned}$$

The Heath-Jarrow-Morton Framework

Given a initial forward curve $T \mapsto f(0, T)$ then, for every maturity T and under the real-world probability measure \mathbf{P} , the forward rate process $t \mapsto f(t, T)$ follows

$$f(t, T) = f(0, T) + \int_0^t \alpha(s, T) ds + \int_0^t \sigma(s, T) dW(s), \quad t \leq T,$$

where $\alpha(t, T) \in \mathbf{R}$ and $\sigma(t, T) := (\sigma_1(t, T), \dots, \sigma_n(t, T))$ satisfy the technical conditions: (1) α and σ are previsible and adapted to \mathcal{F}_t ; (2) $\int_0^T \int_0^T |\alpha(s, t)| ds dt < \infty$ for

all T ; (3) $\sup_{s, t \leq T} \|\sigma(s, t)\| < \infty$ for all T . The short-rate process is given by

$$r(t) = f(t, t) = f(0, t) + \int_0^t \alpha(s, t) ds + \int_0^t \sigma(s, t) dW(s),$$

so the cash account and zero coupon T -bond prices are well-defined and obtained through

$$B(t) = \exp \left(\int_0^t r(s) ds \right), \quad P(t, T) = \exp \left(- \int_t^T f(t, u) du \right).$$

The discounted asset price $Z(t, T) = P(t, T)/B(t)$ satisfies

$$dZ(t, T) = Z(t, T) \left[\underbrace{\left(\frac{1}{2} S^2(t, T) - \int_t^T \alpha(t, u) du \right)}_{b(t, T)} dt + S(t, T) dW(t) \right],$$

where $S(t, T) := - \int_t^T \sigma(s, u) du$. The *HJM drift condition* states that

$$\mathbf{Q} \text{ is EMM (i.e., no arbitrage for bonds)} \iff b(t, T) = -S(t, T)\gamma(t),$$

where $\widetilde{W}(t) := W(t) - \int_0^t \gamma(s) ds$ is a \mathbf{Q} -Brownian motion. If this holds, then under \mathbf{Q} , the forward rate process follows

$$f(t, T) = f(0, T) + \underbrace{\int_0^t \left(\sigma(s, T) \int_s^T \sigma(s, u) du \right) ds}_{\text{HJM drift}} + \int_0^t \sigma(s, T) d\widetilde{W}(s),$$

and the discounted asset $Z(t, T)$ satisfies $dZ(t, T) = Z(0, T) \mathcal{E}_t(X)$ with

$$X(t) = \int_0^t S(s, T) d\widetilde{W}(s).$$

The LIBOR Market Model

For a tenor $\delta > 0$, the *LIBOR rate* $L(t, T, T + \delta)$ is the rate such that an investment of 1 at time T will grow to $1 + \delta L(t, T, T + \delta)$ at time $T + \delta$. The *forward LIBOR rate* (i.e., a contract made at time t under which we pay 1 at time T and receive back $1 + \delta L(t, T, T + \delta)$ at time $T + \delta$) is defined as

$$L(t, T) := L(t, T, T + \delta) = \frac{1}{\delta} \left(\frac{P(t, T)}{P(t, T + \delta)} - 1 \right),$$

and satisfies $L(t, T) = L(T, T, T + \delta)$.

Under the real-world probability measure \mathbf{P} , The LMM assumes that each LIBOR process $(L(t, T_m))_{0 \leq t \leq T_m}$ satisfies

$$dL(t, T_m) = L(t, T_m) [\mu(t, L(t, T_m))dt + \lambda_m(t, L(t, T_m))' dW(t)],$$

where $W = (W^1, \dots, W^d)$ is a d -dimensional Brownian motion with instantaneous correlations

$$d\langle W^i, W^j \rangle(t) = \rho_{i,j}(t) dt, \quad i, j = 1, 2, \dots, d.$$

The function $\lambda(t, L): [0, T_j] \times \mathbf{R} \rightarrow \mathbf{R}^{N \times d}$ is the volatility, and $\mu(t): [0, T_j] \rightarrow \mathbf{R}$ is the drift.

Let $0 \leq m, n \leq N - 1$. Then the dynamics of $L(t, T_m)$ under the forward measure $\mathbf{P}_{T_{n+1}}$ is for $m < n$ given by

$$dL(t, T_m) = L(t, T_m) \left[-\lambda(t, T_m) \sum_{r=m+1}^n \sigma_{T_r, T_{r+1}}(t)' dt + \lambda(t, T_m) dW^m(t) \right]$$

For $m = n$,

$$dL(t, T_m) = L(t, T_m) \lambda(t, T_m) dW_t^m$$

and for $m > n$ we have

$$dL(t, T_m) = L(t, T_m) \left[\lambda(t, T_m) \sum_{r=n+1}^m \sigma_{T_r, T_{r+1}}(t)' dt + \lambda(t, T_m) dW_t^m \right]$$