

1 Verification that the Ito-Wentzell Formula works

We write a weaker version of theorem 3.1 in Krylov using notation and assumptions that are closer to our model. For $p \in [0, S]$, and $t \in [0, T]$, let:

$$\begin{aligned} dQ(p, t) &= \mu_Q(p, t)dt + \sigma_Q(p, t) \int_{s=0}^S b_Q(p, s, t)W(ds, dt) \\ dP(t) &= \mu_P(t)dt + \sigma_P(t) \int_{s=0}^S b_P(s, t)W(ds, dt) \end{aligned} \quad (1)$$

where all stochastic processes are \mathcal{F}_t -adapted.

1.1 Assumption 1¹

- i) We have $P(t) \in [0, S]$.
- ii) For any $\omega \in \Omega$ and $t \in [0, T]$, the function $Q(p)$ is continuous in p
- iii) For almost any $(\omega, t) \in \Omega \times [\tau, T]$
 - a) $\mu_Q(p)$ and $\sigma_Q(p)b_Q(p, s)$ and $\sigma_Q^2(p)$ are continuous in p
 - b) the following functions are continuous functions of p :

1	$\frac{1}{2}\sigma_P^2 \frac{\partial^2 Q(p)}{\partial p^2} + \mu_P \frac{\partial Q(p)}{\partial p}$	$LF(x)$
2	$\sigma_P b_P(s) \frac{\partial Q(p)}{\partial p}$	$\Lambda^k F(x)$
3	$\sigma_P b_P(s) \frac{\partial \sigma_Q(p)b_Q(p, s)}{\partial p}$	$\Lambda^k H^k(x)$
4	$\sigma_P(t)^2 b_P(s) \left(\frac{\partial Q(p)}{\partial p} \right)^2$	$ \Lambda F(x) _{l^2}$
5	$\sigma_P^2(t) \int_0^S (b_P(s) \frac{\partial \sigma_Q(p)b_Q(p, s)}{\partial p})^2 ds$	$ \Lambda H(x) _{l^2}$
- iv) For $p \in [0, S]$, we have almost surely:

$$\int_0^T Q(p, t) |\mu_P(t) + \frac{1}{2}\sigma_P^2(t)| + \frac{1}{2}Q^2(p, t)\sigma_P^2(t) + |\mu_Q(p, t)| + \sigma_Q^2(p, t) dt < \infty$$

and equation (1) admits a solution.

- v) We have almost surely:

$$\int_{p=0}^S \int_0^T Q(p, t) (|\mu_P(p, t)| + \frac{1}{2}\sigma_P^2(t)) dt + \frac{1}{2} \left(\int_0^T Q^2(p, t) \sigma_P^2(t) dt \right)^{1/2} dp < \infty$$

and, for all $p \in [0, S]$,

$$\begin{aligned} \int_0^T |\mu_Q(p, t)| + \frac{1}{2}\sigma_P^2(t) \frac{\partial^2 Q(p, t)}{\partial p^2} + \mu_P(t) \frac{\partial Q(p, t)}{\partial p} + \left(\sigma_P(t) \frac{\partial Q(p, t)}{\partial p} \right)^2 + \sigma_Q^2(p) \\ + \int_0^S (\sigma_P(t)b_P(s, t))^2 \left[\frac{\partial}{\partial p} \sigma_Q(p, t)b_Q(p, s, t) \right]^2 ds dt < \infty \end{aligned}$$

¹We do not follow exactly the numbering of the assumptions of theorem 3.1 in Krylov. Note that our assumptions are simpler because p is assumed bounded and we always take the L_2 norm.

1.2 Verification of Assumption 1

We study the following model:

$$dh(p, t) = a(p)(\bar{h}(p) - h(p, t))dt + \sigma_h(p) \int_{s=0}^S b_h(p, s)W(ds, dt) \quad \forall p \in [0, \mathfrak{S}]$$

$$q(p, t) = \exp \int_{x=0}^p h(x, t)dx \quad \forall p \in [0, \mathfrak{S}]$$

$$d\eta(t) = a_\eta(\bar{\eta} - \eta(t))dt + g(\eta(t)) \left(\int_{s=0}^S \sigma_\eta(s)W(ds, dt, \omega) \right) \quad (4)$$

$$Q(p, t) = \eta(t, \omega) \int_{x=0}^S q(x, t)dx - \int_{x=0}^p q(x, t)dx \quad \forall p \in [0, S] \quad (5)$$

where:

- $a(p)$, $\bar{h}(p)$, $\sigma_h(p)$, and $b_h(p, s)$ are continuous functions of p , bounded on $p \in [0, S]$
- $a_\eta \geq 0$
- $\sigma_h(p)^2$ is a continuously differentiable function of p
- g is a nonnegative Lipschitz continuous function such that $g(\eta) = 0$ if $\eta \leq 0$ or $\eta \geq 1$.

Lipschitz continuity ensures that (3) has a solution. A possible form (which is not Lipschitz continuous, but can be mollified into one) is:

$$g(\eta) = \sigma_\eta \sqrt{\eta(1 - \eta)}$$

1.2.1 Solution and Properties

The solution of (2) is:

$$h(x, t) = h_{0,x} \exp(-a(x)t) + \bar{h}(x)(1 - \exp(-a(x)t)) + \quad (6)$$

$$\int_0^t \exp(-a(x)(t-u)) \int_0^S b_h(p, u)W(ds, u)du \quad (7)$$

As stated above, (4) has a solution, thus (5) is fully defined. Differentiating (5), we have:

$$\frac{\partial Q}{\partial p} = \exp \left(\int_{x=0}^p h(x, t)dx \right) \quad (8)$$

$$\frac{\partial^2 Q(p)}{\partial p^2} = -h(p) \exp \left(\int_{x=0}^p h(x)dx \right) \quad (9)$$

As for the infinitesimal parameters, let:

$$\begin{aligned}
f(p, t) = & \int_{x=0}^p h(x, t) \exp\left(\int_{x=0}^x h(x, t) dx\right) a(p)(\bar{h}(p) - x) dx \\
& \frac{1}{2} \int_{x=0}^p h^2(x, t) \exp\left(\int_{x=0}^x h(x, t) dx\right) \sigma_h^2(x) dx + \\
& \frac{1}{2} (h(p, t) \exp\left(\int_{x=0}^p h(x, t) dx\right) \sigma_h^2(p) - h(0, t) \sigma_h^2(0)) \\
& \frac{1}{2} \int_{x=0}^p h(x, t) \exp\left(\int_{y=0}^x h(y, t) dy\right) [h(x, t) \sigma_h^2(x) + \frac{d}{dx} \sigma_h^2(x)] dx
\end{aligned} \tag{10}$$

By Ito's lemma, the drift of $Q(p, t)$ is:

$$\begin{aligned}
\mu_Q(p, t) = & \eta(t) f(S, t) - f(p, t) + \\
& a_\eta(\bar{\eta} - \eta(t)) \int_{x=0}^S \exp\left(\int_{y=0}^x h(y, t) dy\right) dx + \\
& g(\eta(t)) \int_{x=0}^S h(x, t) \exp\left(\int_{y=0}^x h(y, t) dy\right) \sigma_h(x) dx
\end{aligned} \tag{11}$$

while the volatility of $Q(p, t)$ is:

$$\begin{aligned}
\sigma_Q(p, t) b_Q(p, s, t) = & g(\eta(t)) (Q(S, t) - Q(0, t)) + \\
& \eta \int_{x=0}^S h(x, t) \exp\left(\int_{y=0}^x h(y, t) dy\right) \sigma_h(x) b_h(x, s) dx - \\
& \int_{x=0}^p h(x, t) \exp\left(\int_{y=0}^x h(y, t) dy\right) \sigma_h(x) b_h(x, s) dx
\end{aligned} \tag{12}$$

1.2.2 Verification of all assumptions.

Assumption (i) (boundedness of $P(t)$) is verified in the main text. It is clear that, in our model Q is twice-differentiable in p , thus assumption (ii) (continuity of Q in p) is satisfied.

Verification of assumption iii.a) Drift: The advantage of taking a string is that, since $b_h(p, s)$ is continuous in p , then $\int_{s=0}^S b_h(p, s) W(ds, dt)$ is also continuous in p . Thus $h(\cdot, t)$ is continuous. By (8) and (9), it is clear that both $\frac{\partial Q}{\partial p}$ and $\frac{\partial^2 Q}{\partial p^2}$ are continuous functions of p . By continuity $a(p)$ and of $\frac{d}{dp} \sigma_h^2(p)$ the process $f(p, t)$ is differentiable in p . Inspection of (11) shows that μ_Q is continuous in p .

Volatility: We need only investigate the third row of (12), which is clearly a continuous function of p . So is $\sigma_Q(p, t)$.

Verification of assumption iii.b) From (12), we calculate

$$\frac{\partial[\sigma_Q(p,t)b_Q(p,s,t)]}{\partial p} = -h(p,t,\omega) \exp\left(\int_{x=0}^p h(x,t)dx\right) \sigma_h(p,t)b_h(p,s,t) \quad (13)$$

which is clearly a continuous function of p . by (8) and (9), $\frac{\partial Q}{\partial p}$ and $\frac{\partial^2 Q}{\partial p^2}$ are continuous functions of p . This shows that $\frac{1}{2}\sigma_P^2(t)\frac{\partial^2 Q(p,t)}{\partial p^2} + \mu_P(t)\frac{\partial Q(p,t)}{\partial p}$ as well as $\sigma_P(t)b_P(t,s)\frac{\partial Q(p)}{\partial p}$ and $\sigma_P(t)b_P(t,s)\frac{\partial\sigma_Q(p,t)b_Q(p,s,t)}{\partial p}$ are continuous functions of p . Continuity of $\sigma_P(t)^2b_P(s)(\frac{\partial Q(p)}{\partial p})^2$ and $\sigma_P^2(t)\int_0^S(b_P(s)\frac{\partial\sigma_Q(p)b_Q(p,s)}{\partial p})^2ds$ follows trivially.

Lemma Let $Y_i(s,t)$ and $Z_i(s,t)$ be collections of random variables, for $i = 1, \dots, n$, $0 \leq s \leq S$, and $0 \leq t \leq T$. Suppose $E[\int_0^S \int_0^T \sum_{i=1}^n Y_i(s,t)Z_i(s,t)ds] < \infty$. A sufficient condition for

$$P\left(\int_0^S \int_0^T \sum_{i=1}^n Y_i(s,t)Z_i(s,t)ds < \infty\right) = 1$$

is that for all s, t and $i = 1, \dots, n$

$$\begin{aligned} E[Y_i(s,t)^8] &< \infty \\ E[Z_i(s,t)^8] &< \infty \end{aligned}$$

Verification of assumptions iv and v Clearly $\int_{y=0}^p h(y,t)dy$ is normal, thus $q(p,t)$ is lognormal, and has thus finite moments. For simplicity, we drop the argument t . By the lemma above, we can break the argument into verifying finiteness of the 16th moments of the following random variables:

1. **Net demand $Q(p)$.** By Jensen's inequality:

$$E\left[\left(\int_0^S q(x)dx\right)^{16}\right] \leq \int_0^S E\left[\exp\left(16 \int_0^x h(y)dy\right)\right]ds$$

which is bounded. By (5)

$$-\int_{x=0}^S q(x)dx \leq Q(p) \leq \int_{x=0}^S q(x)dx$$

Thus $Q(p)$ is bounded.

2. **Volatility of net demand σ_Q** :Observe that $\sigma_h(x,s)$ is bounded. We apply Jensen's inequality again, and it is sufficient to see that the following

term is bounded:

$$E[(\int_{x=0}^S h(x) \exp(\int_{y=0}^x h(y) dy) dx)^{16}] = E[(\int_{x=0}^S \frac{d}{dx} \exp(\int_{y=0}^x h(y) dy) dx)^{16}] = E[\exp(\int_{y=0}^S h(y) dy) dx]^{16}$$

Thus $\sigma_Q(p)$ is bounded.

3. **Drift of net demand μ_Q .** Applying Jensen's inequality and the Cauchy-Schwartz inequalities to (11), it is sufficient to verify the finiteness of the appropriate moments of:

$f(S), \int_{x=0}^S \exp(\int_{y=0}^x h(y) dy) dx$, and $\int_{x=0}^S h(x) \exp(\int_{y=0}^x h(y) dy) \sigma_h(x) dx$. For $f(S)$, the same type of development applies, and $f(S)$ has finite appropriate moment because $\sigma_h^2(x)$ is differentiable.

4. **Price drift μ_P .** We refer the reader to (*) in lemma 1 in the main text. Since the numerator has finite 16th moment, it is sufficient, by the Cauchy-Schwartz inequality, to prove boundedness of:

$$E[(\frac{1}{\partial Q / \partial p})^{16}] = E[\exp\left(-16 \int_{x=0}^p h(x, t) dx\right)]$$

which is obvious.

5. **Price volatility $\sigma_P(t)$** : We use the same argument as above, since

$$\sigma_P = \frac{\sigma_Q}{\partial Q / \partial p}$$

6. **First derivative of net demand $\frac{\partial Q(p)}{\partial p}$** Finiteness of the moment is clear, since the random variable in (8) is lognormal.

7. **Second Derivative of Net Demand $\frac{\partial^2 Q(p)}{\partial p^2}$**

Using Cauchy-Schwartz on (9), we see that this has finite appropriate moment.

8. **Derivative of Net Demand Volatility $\frac{\partial \sigma_Q(p) b_Q(p, s)}{\partial p}$**

We reuse (13), and use the same logic as above.

1.2.3 Proof of lemma

Let

$$X_i(s, t) = Y_i(s, t) Z_i(s, t)$$

By Chebyshev's inequality, it is sufficient to prove that:

$$E[(\int_0^S \int_0^T \sum_{i=1}^n X_i(s, t) ds)^2] < \infty$$

Since:

$$E[(\int_0^S \int_0^T \sum_{i=1}^n X_i(s, t) ds)^2] \leq S^2 T^2 \max_{\substack{0 \leq s, s' \leq S, \\ 0 \leq t, t' \leq T}} E[(\sum_{i=1}^n X_i(s, t))(\sum_{j=1}^n X_j(s', t'))]$$

Applying Jensen's inequality twice, and then the Cauchy-Schwartz inequality:

$$\begin{aligned} E[(\sum_{i=1}^n X_i(s, t) \sum_{j=1}^n X_j(s', t'))^2] &\leq n^2 \sum_{i=1}^n \sum_{j=1}^n E[X_i(s, t)^2 X_j(s', t')^2] \\ &\leq n^2 \sum_{i=1}^n \sum_{j=1}^n (E[X_i(s, t)^4] E[X_j(s', t')^4])^{1/2} \end{aligned}$$

We conclude by the Cauchy-Schwartz inequality:

$$E[X_i(s, t)^4] \leq (E[Y_i(s, t)^8] E[Z_i(s, t)^8])^{1/2}$$