

## Appendix: Lemmas to Theorem 1

Throughout the appendix,  $\approx$  means higher order terms are neglected,  $C$  refers to a positive constant which may not be the same at each occurrence, and  $\rightsquigarrow$  signifies the weak convergence of a stochastic process on the relevant probability space.

**Lemma 1**  $\hat{v} = O_p(1)$  as  $\|\delta_0\| \rightarrow 0$ .

**Proof.** Take  $v \geq 0$  for illustration and  $v < 0$  can be similarly proved. Partition  $\mathbb{R}_+ := \{v|v \geq 0\}$  into the "shells"  $S_j = \{v : 2^{j-1} < v \leq 2^j\}$  with  $j$  ranging over the integers. Given an integer  $J$ ,

$$P(\hat{v} > 2^J) \leq \sum_{j \geq J} P\left(\inf_{v \in S_j} \left(D\left(\frac{v}{\|\delta_0\|^2}\right)\right) \leq D(0) = 0\right).$$

Now,

$$\begin{aligned} & P\left(\inf_{v \in S_j} \left(D\left(\frac{v}{\|\delta_0\|^2}\right)\right) \leq 0\right) \\ & \leq P\left(\inf_{v \in S_j} \left(\mathbb{E}\left[D\left(\frac{v}{\|\delta_0\|^2}\right)\right] - \left|D\left(\frac{v}{\|\delta_0\|^2}\right) - \mathbb{E}\left[D\left(\frac{v}{\|\delta_0\|^2}\right)\right]\right|\right) \leq 0\right) \\ & \leq P\left(\sup_{v \in S_j} \left|D\left(\frac{v}{\|\delta_0\|^2}\right) - \mathbb{E}\left[D\left(\frac{v}{\|\delta_0\|^2}\right)\right]\right| \geq \inf_{v \in S_j} \mathbb{E}\left[D\left(\frac{v}{\|\delta_0\|^2}\right)\right]\right) \\ & \leq \mathbb{E}\left[\sup_{v \in S_j} \left|D\left(\frac{v}{\|\delta_0\|^2}\right) - \mathbb{E}\left[D\left(\frac{v}{\|\delta_0\|^2}\right)\right]\right|^2\right] / \inf_{v \in S_j} \mathbb{E}\left[D\left(\frac{v}{\|\delta_0\|^2}\right)\right]^2, \end{aligned}$$

where the first equality is because

$$D\left(\frac{v}{\|\delta_0\|^2}\right) - \mathbb{E}\left[D\left(\frac{v}{\|\delta_0\|^2}\right)\right] \geq -\left|D\left(\frac{v}{\|\delta_0\|^2}\right) - \mathbb{E}\left[D\left(\frac{v}{\|\delta_0\|^2}\right)\right]\right|$$

which implies  $\mathbb{E}\left[D\left(\frac{v}{\|\delta_0\|^2}\right)\right] - \left|D\left(\frac{v}{\|\delta_0\|^2}\right) - \mathbb{E}\left[D\left(\frac{v}{\|\delta_0\|^2}\right)\right]\right| \leq D\left(\frac{v}{\|\delta_0\|^2}\right)$ , the second equality is because

$$\inf_{v \in S_j} \left(\mathbb{E}\left[D\left(\frac{v}{\|\delta_0\|^2}\right)\right] - \left|D\left(\frac{v}{\|\delta_0\|^2}\right) - \mathbb{E}\left[D\left(\frac{v}{\|\delta_0\|^2}\right)\right]\right|\right) \geq \inf_{v \in S_j} \mathbb{E}\left[D\left(\frac{v}{\|\delta_0\|^2}\right)\right] - \sup_{v \in S_j} \left|D\left(\frac{v}{\|\delta_0\|^2}\right) - \mathbb{E}\left[D\left(\frac{v}{\|\delta_0\|^2}\right)\right]\right|,$$

and the last equality is from Markov's inequality. Note that the demeaned process  $\tilde{D}(v) := D\left(\frac{v}{\|\delta_0\|^2}\right) - \mathbb{E}\left[D\left(\frac{v}{\|\delta_0\|^2}\right)\right]$  satisfies for any  $s \leq t$ ,

$$\begin{aligned} \mathbb{E}\left[\tilde{D}(t) \mid \left\{\tilde{D}(\tau)\right\}_{\tau \leq s}\right] &= \mathbb{E}\left[\tilde{D}(t) - \tilde{D}(s) \mid \left\{\tilde{D}(\tau)\right\}_{\tau \leq s}\right] + \mathbb{E}\left[\tilde{D}(s) \mid \left\{\tilde{D}(\tau)\right\}_{\tau \leq s}\right] \\ &= \mathbb{E}\left[\tilde{D}(s) \mid \left\{\tilde{D}(\tau)\right\}_{\tau \leq s}\right] = \tilde{D}(s), \end{aligned}$$

where the second to last equality is because  $\tilde{D}(v)$  is an independent increments process with mean zero such that  $\tilde{D}(t) - \tilde{D}(s)$  is independent of  $\left\{\tilde{D}(\tau)\right\}_{\tau \leq s}$  and has mean zero. So  $\tilde{D}(v)$  is a continuous martingale

indexed by  $v$  and  $\left|\tilde{D}(v)\right|$  is a submartingale. To calculate  $\mathbb{E}\left[\sup_{v \in S_j} \left|\tilde{D}(v)\right|^2\right]$ , we apply Doob's martingale inequality (see, e.g., Theorem 20 of Protter (2004)). First, by Assumption (iii),

$$\mathbb{E}[z_{2i}] = \mathbb{E}[-2x'_i \delta_0 u_{2i} + \delta_0 x_i x'_i \delta_0 | q_i = \gamma_0 +] = \delta_0 \mathbb{E}[xx' | q = \gamma_0] \delta_0 =: \|\delta_0\|^2 Q_\delta,$$

and

$$\begin{aligned}
\mathbb{V}(z_{2i}) &= \mathbb{V}(-2x'_i \delta_0 u_{2i} + \delta_0 x_i x'_i \delta_0 | q_i = \gamma_0 +) \\
&= \mathbb{E}[\mathbb{V}(-2x'_i \delta_0 u_{2i} + \delta_0 x_i x'_i \delta_0 | x_i, q_i = \gamma_0 +) | q_i = \gamma_0 +] + \mathbb{V}(\mathbb{E}[-2x'_i \delta_0 u_{2i} + \delta_0 x_i x'_i \delta_0 | x_i, q_i = \gamma_0 +] | q_i = \gamma_0 +) \\
&= \mathbb{E}\left[(-2x'_i \delta_0 u_{2i})^2 | q_i = \gamma_0\right] + \mathbb{V}(\delta_0 x_i x'_i \delta_0 | q_i = \gamma_0) \\
&= 4\delta_0 \mathbb{E}[x x u_2^2 | q = \gamma_0] \delta_0 + O(\|\delta_0\|^4) =: 4\|\delta_0\|^2 \Omega_{2\delta},
\end{aligned}$$

where  $Q_\delta = \frac{\delta_0 \mathbb{E}[x x | q = \gamma_0] \delta_0}{\delta_0' \delta_0} \rightarrow Q$  and  $\Omega_{2\delta} = \frac{\delta_0 \mathbb{E}[x x u_2^2 | q = \gamma_0] \delta_0}{\delta_0' \delta_0} + O(\|\delta_0\|^2) \rightarrow \Omega_2$  as  $\|\delta_0\| \rightarrow 0$  from Assumption (i). Now, by Doob's martingale inequality,

$$\begin{aligned}
\mathbb{E}\left[\sup_{v \in S_j} \left| D\left(\frac{v}{\|\delta_0\|^2}\right) - \mathbb{E}\left[D\left(\frac{v}{\|\delta_0\|^2}\right)\right] \right|^2\right] &\leq 4\mathbb{E}\left[\left| D\left(\frac{2^j}{\|\delta_0\|^2}\right) - \mathbb{E}\left[D\left(\frac{2^j}{\|\delta_0\|^2}\right)\right] \right|^2\right] \\
&= 16f_q(\gamma_0) 2^j \Omega_2 + o(1),
\end{aligned}$$

where the equality is because

$$\begin{aligned}
\mathbb{V}\left(D\left(\frac{v}{\|\delta_0\|^2}\right)\right) &= \mathbb{E}\left[\mathbb{V}\left(D\left(\frac{v}{\|\delta_0\|^2}\right) \middle| N_2(\cdot)\right)\right] + \mathbb{V}\left(\mathbb{E}\left[D\left(\frac{v}{\|\delta_0\|^2}\right) \middle| N_2(\cdot)\right]\right) \\
&= \mathbb{E}\left[N_2\left(\frac{v}{\|\delta_0\|^2}\right) \mathbb{V}(z_{2i})\right] + \mathbb{V}\left(N_2\left(\frac{v}{\|\delta_0\|^2}\right) \mathbb{E}[z_{2i}]\right) \\
&= \frac{f_q(\gamma_0)v}{\|\delta_0\|^2} \mathbb{V}(z_{2i}) + \frac{f_q(\gamma_0)v}{\|\delta_0\|^2} \mathbb{E}[z_{2i}]^2 = \frac{f_q(\gamma_0)v}{\|\delta_0\|^2} \mathbb{E}[z_{2i}^2] \\
&= f_q(\gamma_0) v (4\Omega_{2\delta} + \|\delta_0\|^2 Q_\delta^2) = 4f_q(\gamma_0) v \Omega_2 + o(1),
\end{aligned}$$

with the last equality from Assumption (i) and  $\|\delta_0\| \rightarrow 0$ . Since

$$\begin{aligned}
\mathbb{E}\left[D\left(\frac{v}{\|\delta_0\|^2}\right)\right] &= \mathbb{E}\left[\mathbb{E}\left[D\left(\frac{v}{\|\delta_0\|^2}\right) \middle| N_2(\cdot)\right]\right] = \mathbb{E}\left[N_2\left(\frac{v}{\|\delta_0\|^2}\right) \mathbb{E}[z_{2i}]\right] \\
&= f_q(\gamma_0) v \frac{\mathbb{E}[z_{2i}]}{\|\delta_0\|^2} = f_q(\gamma_0) v Q_\delta = f_q(\gamma_0) v Q + o(1),
\end{aligned}$$

we have

$$P\left(\inf_{v \in S_j} \left(D\left(\frac{v}{\|\delta_0\|^2}\right) \leq 0\right)\right) \leq \frac{16f_q(\gamma_0) 2^j \Omega_2}{(2^{j-1} f_q(\gamma_0) Q)^2} = \frac{C}{2^{j-1}},$$

where  $C = 32\Omega_2/f_q(\gamma_0) Q^2$  is a positive constant. As a result,

$$P(\hat{v} > 2^J) \leq \sum_{j \geq J} \frac{C}{2^{j-1}} \rightarrow 0$$

as  $J \rightarrow \infty$  and the proof is complete.  $\blacksquare$

**Lemma 2**  $D\left(\frac{v}{\|\delta_0\|^2}\right) \rightsquigarrow C(v) \in \mathbf{C}_{\min}(\mathbb{R})$  as  $\|\delta_0\| \rightarrow 0$ .

**Proof.** As in Lemma 1, take  $v > 0$ . Define

$$Z_\delta(v) = \frac{D\left(\frac{v}{\|\delta_0\|^2}\right) - \mathbb{E}\left[D\left(\frac{v}{\|\delta_0\|^2}\right)\right]}{\sqrt{\mathbb{V}\left(D\left(\frac{v}{\|\delta_0\|^2}\right)\right)}/v} = \frac{D\left(\frac{v}{\|\delta_0\|^2}\right) - \mu_\delta v}{\sigma_\delta},$$

where  $\mu_\delta = f_q(\gamma_0) Q_\delta$  and  $\sigma_\delta^2 = f_q(\gamma_0) (4\Omega_{2\delta} + \|\delta_0\|^2 Q_\delta^2)$ . Since  $\mu_\delta \rightarrow f_q(\gamma_0) Q$  and  $\sigma_\delta^2 \rightarrow 4f_q(\gamma_0) \Omega_2$ , by Slutsky's theorem, we need only show that  $Z_\delta(v) \rightsquigarrow W_2(v)$ . We check the two conditions in Theorem 2.3 of Kim and Pollard. Specifically, (i) fidi-convergence: for any  $v_2 > v_1 > 0$  and  $t_1, t_2 \in \mathbb{R}$ ,

$$\mathbb{E} \left[ \exp \left\{ \sqrt{-1} [t_1 Z_\delta(v_1) + t_2 Z_\delta(v_2)] \right\} \right] \longrightarrow \exp \left\{ -\frac{1}{2} (t_1, t_2) \begin{pmatrix} v_1 & v_1 \\ v_1 & v_2 \end{pmatrix} \begin{pmatrix} t_1 \\ t_2 \end{pmatrix} \right\};$$

and (ii) stochastic equicontinuity: for any  $\epsilon > 0, \eta > 0$ , there exist a  $\Delta > 0$  such that

$$\overline{\lim}_{n \rightarrow \infty} P \left( \sup_{|v_2 - v_1| < \Delta} |Z_\delta(v_2) - Z_\delta(v_1)| > \epsilon \right) < \eta.$$

To prove (i), note that

$$\begin{aligned} & \mathbb{E} \left[ \exp \left\{ \sqrt{-1} [t_1 Z_\delta(v_1) + t_2 Z_\delta(v_2)] \right\} \right] \\ &= \mathbb{E} \left[ \mathbb{E} \left[ \exp \left\{ \sqrt{-1} [t_1 Z_\delta(v_1) + t_2 Z_\delta(v_2)] \right\} \middle| N_2(\cdot) \right] \right] \\ &= \exp \left\{ \sqrt{-1} \left[ -t_1 \frac{\mu_\delta v_1}{\sigma_\delta} - t_2 \frac{\mu_\delta v_2}{\sigma_\delta} \right] \right\} \cdot \mathbb{E} \left[ \exp \left\{ \sqrt{-1} \left[ t_1 \frac{\sum_{i=1}^{N_2(\frac{v_1}{\|\delta_0\|^2})} z_{2i}}{\sigma_\delta} + t_2 \frac{\sum_{i=1}^{N_2(\frac{v_2}{\|\delta_0\|^2})} z_{2i}}{\sigma_\delta} \right] \right\} \right] \right] \\ &= \exp \left\{ \sqrt{-1} \left[ -t_1 \frac{\mu_\delta v_1}{\sigma_\delta} - t_2 \frac{\mu_\delta v_2}{\sigma_\delta} \right] \right\} \cdot \mathbb{E} \left[ \mathbb{E} \left[ \prod_{i=1}^{N_2(\frac{v_1}{\|\delta_0\|^2})} \mathbb{E} \left[ \exp \left\{ \sqrt{-1} (t_1 + t_2) \frac{z_{2i}}{\sigma_\delta} \right\} \right] \middle| N_2(\cdot) \right] \right] \right] \\ &\quad \cdot \mathbb{E} \left[ \mathbb{E} \left[ \prod_{i=N_2(\frac{v_1}{\|\delta_0\|^2})+1}^{N_2(\frac{v_2}{\|\delta_0\|^2})} \mathbb{E} \left[ \exp \left\{ \sqrt{-1} t_2 \frac{z_{2i}}{\sigma_\delta} \right\} \right] \middle| N_2(\cdot) \right] \right] \right] \\ &= \exp \left\{ \sqrt{-1} \left[ -t_1 \frac{\mu_\delta v_1}{\sigma_\delta} - t_2 \frac{\mu_\delta v_2}{\sigma_\delta} \right] \right\} \cdot \sum_{k=0}^{\infty} \left\{ \frac{e^{-f_q(\gamma_0)v_1/\|\delta_0\|^2} (f_q(\gamma_0) \frac{v_1}{\|\delta_0\|^2})^k}{k!} \prod_{i=1}^k \mathbb{E} \left[ \exp \left\{ \sqrt{-1} (t_1 + t_2) \frac{z_{2i}}{\sigma_\delta} \right\} \right] \right\} \\ &\quad \cdot \sum_{k=0}^{\infty} \left\{ \frac{e^{-f_q(\gamma_0)(v_2-v_1)/\|\delta_0\|^2} (f_q(\gamma_0) \frac{v_2-v_1}{\|\delta_0\|^2})^k}{k!} \prod_{i=1}^k \mathbb{E} \left[ \exp \left\{ \sqrt{-1} t_2 \frac{z_{2i}}{\sigma_\delta} \right\} \right] \right\} \\ &=: T_1 \cdot T_2 \cdot T_3, \end{aligned}$$

where  $\prod_{i=1}^0 := 1$ . Because  $\|\delta_0\| \rightarrow 0$ , we take Taylor expansion of

$$\begin{aligned} \mathbb{E} \left[ \exp \left\{ \sqrt{-1} (t_1 + t_2) \frac{z_{2i}}{\sigma_\delta} \right\} \right] &= \mathbb{E} \left[ \exp \left\{ \sqrt{-1} (t_1 + t_2) \|\delta_0\| \frac{z_{2i}/\|\delta_0\|}{\sigma_\delta} \right\} \right] \\ &=: \mathbb{E} \left[ \exp \left\{ \sqrt{-1} (t_1 + t_2) \|\delta_0\| Z_{2i} \right\} \right] := g((t_1 + t_2) \|\delta_0\|) \end{aligned}$$

about  $(t_1 + t_2) \|\delta_0\|$  at 0, giving

$$\begin{aligned} & \mathbb{E} \left[ \exp \left\{ \sqrt{-1} (t_1 + t_2) \|\delta_0\| Z_{2i} \right\} \right] \\ &= 1 + \sqrt{-1} (t_1 + t_2) \|\delta_0\| \mathbb{E} [Z_{2i}] - \frac{1}{2} (t_1 + t_2)^2 \|\delta_0\|^2 \mathbb{E} [Z_{2i}^2] + o\left((t_1 + t_2)^2 \|\delta_0\|^2\right) \\ &= 1 + \sqrt{-1} (t_1 + t_2) \frac{\mathbb{E} [z_{2i}]}{\sigma_\delta} - \frac{1}{2} (t_1 + t_2)^2 \frac{\mathbb{E} [z_{2i}^2]}{\sigma_\delta^2} + o\left(\|\delta_0\|^2\right), \end{aligned}$$

where  $E [Z_{2i}^2] = O(1)$ . In consequence,

$$\begin{aligned}
T_2 &= \sum_{k=0}^{\infty} \left\{ \frac{e^{-f_q(\gamma_0)v_1/\|\delta_0\|^2} \left( f_q(\gamma_0) \frac{v_1}{\|\delta_0\|^2} \right)^k}{k!} \prod_{i=1}^k \left[ 1 + \sqrt{-1} (t_1 + t_2) \frac{\mathbb{E}[z_{2i}]}{\sigma_\delta} - \frac{1}{2} (t_1 + t_2)^2 \frac{\mathbb{E}[z_{2i}^2]}{\sigma_\delta^2} + o(\|\delta_0\|^2) \right] \right\} \\
&\approx e^{-f_q(\gamma_0)v_1/\|\delta_0\|^2} \sum_{k=0}^{\infty} \frac{\left( f_q(\gamma_0) \frac{v_1}{\|\delta_0\|^2} \right)^k}{k!} \left( 1 + \sqrt{-1} (t_1 + t_2) \frac{\|\delta_0\|^2 Q_\delta}{\sigma_\delta} - \frac{1}{2} (t_1 + t_2)^2 \frac{\mathbb{E}[z_{2i}^2]}{\frac{f_q(\gamma_0)}{\|\delta_0\|^2} \mathbb{E}[z_{2i}^2]} \right)^k \\
&= e^{-f_q(\gamma_0)v_1/\|\delta_0\|^2} \sum_{k=0}^{\infty} \frac{1}{k!} \left( f_q(\gamma_0) \frac{v_1}{\|\delta_0\|^2} + \sqrt{-1} (t_1 + t_2) \frac{f_q(\gamma_0) Q_\delta v_1}{\sigma_\delta} - \frac{1}{2} (t_1 + t_2)^2 v_1 \right)^k \\
&= e^{-f_q(\gamma_0)v_1/\|\delta_0\|^2} \exp \left\{ f_q(\gamma_0) \frac{v_1}{\|\delta_0\|^2} + \sqrt{-1} (t_1 + t_2) \frac{f_q(\gamma_0) Q_\delta v_1}{\sigma_\delta} - \frac{1}{2} (t_1 + t_2)^2 v_1 \right\} \\
&= \exp \left\{ \sqrt{-1} (t_1 + t_2) \frac{f_q(\gamma_0) Q_\delta v_1}{\sigma_\delta} - \frac{1}{2} (t_1 + t_2)^2 v_1 \right\}.
\end{aligned}$$

Similarly,

$$T_3 \approx \exp \left\{ \sqrt{-1} t_2 \frac{f_q(\gamma_0) Q_\delta (v_2 - v_1)}{\sigma_\delta} - \frac{1}{2} t_2^2 (v_2 - v_1) \right\}.$$

In summary,

$$\begin{aligned}
&\mathbb{E} \left[ \exp \left\{ \sqrt{-1} [t_1 Z_\delta (v_1) + t_2 Z_\delta (v_2)] \right\} \right] \\
&\approx \exp \left\{ \sqrt{-1} \left[ -t_1 \frac{f_q(\gamma_0) Q_\delta v_1}{\sigma_\delta} - t_2 \frac{f_q(\gamma_0) Q_\delta v_2}{\sigma_\delta} \right] \right\} \exp \left\{ \sqrt{-1} (t_1 + t_2) \frac{f_q(\gamma_0) Q_\delta v_1}{\sigma_\delta} - \frac{1}{2} (t_1 + t_2)^2 v_1 \right\} \\
&\quad \cdot \exp \left\{ \sqrt{-1} t_2 \frac{f_q(\gamma_0) Q_\delta (v_2 - v_1)}{\sigma_\delta} - \frac{1}{2} t_2^2 (v_2 - v_1) \right\} \\
&= \exp \left\{ -\frac{1}{2} (t_1 + t_2)^2 v_1 - \frac{1}{2} t_2^2 (v_2 - v_1) \right\} \\
&= \exp \left\{ -\frac{1}{2} (t_1, t_2) \begin{pmatrix} v_1 & v_1 \\ v_1 & v_2 \end{pmatrix} \begin{pmatrix} t_1 \\ t_2 \end{pmatrix} \right\},
\end{aligned}$$

giving the required result.

To prove (ii), note that

$$\begin{aligned}
&P \left( \sup_{|v_2 - v_1| < \Delta} |Z_\delta (v_2) - Z_\delta (v_1)| > \epsilon \right) = P \left( \sup_{|v_2 - v_1| < \Delta} \left| \frac{\tilde{D}(v_2) - \tilde{D}(v_1)}{\sigma_\delta} \right| > \epsilon \right) \\
&\leq \mathbb{E} \left[ \sup_{|v_2 - v_1| < \Delta} \left| \tilde{D}(v_2) - \tilde{D}(v_1) \right|^2 \right] / \sigma_\delta^2 \epsilon^2 \leq 4 \mathbb{E} \left[ \left| \tilde{D}(v_1 + \Delta) - \tilde{D}(v_1) \right|^2 \right] / \sigma_\delta^2 \epsilon^2 \\
&= \frac{4\sigma_\delta^2 \Delta}{\sigma_\delta^2 \epsilon^2} = C \frac{\Delta}{\epsilon^2},
\end{aligned}$$

where the last inequality is from Doob's martingale inequality with the submartingale defined as  $|\tilde{D}(s + v_1) - \tilde{D}(v_1)|$  indexed by  $0 \leq s \leq \Delta$ . So we can choose  $\Delta$  arbitrarily small to make the above probability smaller than the given  $\eta$ .

Finally, we confirm that  $C(v) \in \mathbf{C}_{\min}(\mathbb{R})$ . It is not hard to check that  $C(v)$  is continuous, has a unique minimum (see Lemma 2.6 of Kim and Pollard (1990)), and  $\lim_{|v| \rightarrow \infty} C(v) = \infty$  almost surely, which follows since  $\lim_{|v| \rightarrow \infty} W_\ell(v) / |v| = 0$  almost surely by virtue of the law of the iterated logarithm for Brownian motion.

■

**Additional Reference:**

Protter, P.E., 2004, Stochastic Integration and Differential Equations, 2nd ed., New York: Springer.