

# Online Supplement

## Appendix A - Proofs for Sections 3 and 4.1

**Proof of Proposition 1:** Let  $\mu_{R1}$  and  $\mu_{R2}$  denote the mean reservation prices for customers who are in the target and non-target segments of the regular product, respectively. Similarly, define  $\mu_{P1}$  and  $\mu_{P2}$  to be the mean reservation prices for customers who are in the target and non-target segments of the promotional product, respectively. Then, we have

$$E(X) = q_{R1}\mu_{R1} + q_{R2}\mu_{R2}$$

and

$$E(Y) = q_{P1}\mu_{P1} + q_{P2}\mu_{P2}.$$

We also have

$$\begin{aligned} E(XY) &= q_{R1} \int_0^\infty \int_0^\infty xy f_{R1}(x)(\delta_{11}f_{P1}(y) + \delta_{12}f_{P2}(y))dxdy \\ &\quad + q_{R2} \int_0^\infty \int_0^\infty xy f_{R2}(x)(\delta_{21}f_{P1}(y) + \delta_{22}f_{P2}(y))dxdy \\ &= q_{R1} \int_0^\infty y(\delta_{11}f_{P1}(y) + \delta_{12}f_{P2}(y)) \int_0^\infty x f_{R1}(x)dxdy \\ &\quad + q_{R2} \int_0^\infty y(\delta_{21}f_{P1}(y) + \delta_{22}f_{P2}(y)) \int_0^\infty x f_{R2}(x)dxdy \\ &= q_{R1}\mu_{R1} \int_0^\infty y(\delta_{11}f_{P1}(y) + \delta_{12}f_{P2}(y))dy + q_{R2}\mu_{R2} \int_0^\infty y(\delta_{21}f_{P1}(y) + \delta_{22}f_{P2}(y))dy \end{aligned}$$

Hence,  $E(XY)$  can simply be written as

$$E(XY) = q_{R1}\mu_{R1}(\delta_{11}\mu_{P1} + \delta_{12}\mu_{P2}) + q_{R2}\mu_{R2}(\delta_{21}\mu_{P1} + \delta_{22}\mu_{P2}).$$

Then, after a few algebraic manipulations, it can be shown that

$$Cov(X, Y) = q_{R1}\mu_{R1}[\mu_{P1}(\delta_{11} - q_{P1}) + \mu_{P2}(\delta_{12} - q_{P2})] + q_{R2}\mu_{R2}[\mu_{P1}(\delta_{21} - q_{P1}) + \mu_{P2}(\delta_{22} - q_{P2})].$$

Since  $\delta_{11} + \delta_{12} = q_{P1} + q_{P2} = \delta_{21} + \delta_{22} = 1$ , we have  $\delta_{11} - q_{P1} = -(\delta_{12} - q_{P2})$  and  $\delta_{21} - q_{P1} = -(\delta_{22} - q_{P2})$ , which yields

$$\begin{aligned} Cov(X, Y) &= q_{R1}\mu_{R1}(\mu_{P1} - \mu_{P2})(\delta_{11} - q_{P1}) + q_{R2}\mu_{R2}(\mu_{P1} - \mu_{P2})(\delta_{21} - q_{P1}) \\ &= (\mu_{P1} - \mu_{P2})[q_{R1}\mu_{R1}(\delta_{11} - q_{P1}) + q_{R2}\mu_{R2}(\delta_{21} - q_{P1})] \end{aligned}$$

Finally, using the fact that  $q_{P1} = q_{R1}\delta_{11} + q_{R2}\delta_{21}$  (see (1)), we establish that

$$Cov(X, Y) = q_{R1}q_{R2}(\mu_{P1} - \mu_{P2})(\mu_{R1} - \mu_{R2})(\delta_{11} + \delta_{22} - 1).$$

Since  $\mu_{P1} - \mu_{P2} > 0$  and  $\mu_{R1} - \mu_{R2} > 0$  (due to assumption (A3)), the result follows.  $\square$

**Proof of Proposition 2:** From (3), we have

$$q_{P1} - \hat{q}_{P1} = q_{P1} - \frac{q_{R1}\delta_{11}\bar{F}_{R1}(r) + q_{R2}\delta_{21}\bar{F}_{R2}(r)}{q_{R1}\bar{F}_{R1}(r) + q_{R2}\bar{F}_{R2}(r)}.$$

Then, it can be shown that  $q_{P1} - \hat{q}_{P1} > 0$  if and only if

$$q_{R1}(q_{P1} - \delta_{11})\bar{F}_{R1}(r) - q_{R2}(\delta_{21} - q_{P1})\bar{F}_{R2}(r) > 0.$$

From assumption (A3), we have  $\bar{F}_{R1}(r) > \bar{F}_{R2}(r)$  (since failure rate ordering implies usual stochastic ordering), and it can be checked from (1) that  $q_{R1}(q_{P1} - \delta_{11}) = q_{R2}(\delta_{21} - q_{P1})$ . Then, for the above inequality to hold true, we must have  $q_{P1} > \delta_{11}$ , which can be shown to be equivalent to  $\delta_{11} + \delta_{22} < 1$  using (1). Hence, the result follows.  $\square$

**Proof of Theorem 1:** Throughout the proof, recall that an optimal  $p_t$  and  $d_t$  will solve the optimization problem in (8).

**Proof of (a):** Lemma 4 proves that  $z_1^*(y, t) > z_2^*(y, t)$ . Let  $\bar{p} = z_1^*(y, t)$  and  $\bar{d} = z_1^*(y, t) - z_2^*(y, t)$ . Then  $(\bar{p}, \bar{d})$  is a feasible price-discount pair for the optimization problem in (8), since  $z_1^*(y, t) > z_2^*(y, t)$ . Furthermore,  $\Pi_1(\bar{p}, \Delta(y, t)) \geq \Pi_1(p, \Delta(y, t))$ ,  $\forall p$ , and  $\Pi_2(\bar{p} - \bar{d}, \Delta(y, t)) \geq \Pi_2(p, \Delta(y, t))$ ,  $\forall p$ , since  $z_1^*(y, t)$  and  $z_2^*(y, t)$  maximize  $\Pi_1(\cdot, \Delta(y, t))$  and  $\Pi_2(\cdot, \Delta(y, t))$ , respectively. Hence the result follows.

**Proof of (b):** If  $\delta_{11} + \delta_{22} = 1$ , it is easy to show that  $q_{P1} = \hat{q}_{P1}$ . This implies that  $z_1^*(y, t) = z_2^*(y, t)$ , and thus it is optimal to set  $p_t = z_1^*(y, t)$  and  $d_t = 0$ .

Now, in the remaining of the proof, suppose that  $\delta_{11} + \delta_{22} > 1$ . Then, Lemma 4 proves that  $z_1^*(y, t) < z_2^*(y, t)$ . Now, first, suppose for a contradiction that there exists an optimal price  $\bar{p}_t$  and optimal discount  $\bar{d}_t$  where  $\bar{d}_t > 0$ . Then, we must have  $\Pi_1(\bar{p}_t - \bar{d}_t, \Delta(y, t)) \leq \Pi_1(\bar{p}_t, \Delta(y, t))$ . (Otherwise, we obtain a contradiction to the optimality of  $\bar{p}_t$ .) Hence, using Lemma 3(a), we have  $\Pi_2(\bar{p}_t - \bar{d}_t, \Delta(y, t)) < \Pi_2(\bar{p}_t, \Delta(y, t))$ , which yields a contradiction to the optimality of  $\bar{d}_t > 0$ . Hence it must be that  $\bar{d}_t = 0$ .

Now that we know  $\bar{d}_t = 0$ , observe from (8) and the definition of  $z^*(y, t)$  (given by (11)) that  $z^*(y, t)$  will maximize the firm's revenue. It remains to show that  $z_1^*(y, t) \leq z^*(y, t) \leq z_2^*(y, t)$ .

First, suppose for a contradiction that  $z^*(y, t) < z_1^*(y, t)$ . By definition of  $z_1^*(y, t)$ , we have  $\Pi_1(z^*(y, t), \Delta(y, t)) < \Pi_1(z_1^*(y, t), \Delta(y, t))$ . Now, using Lemma 3(a), we observe  $\Pi_2(z^*(y, t), \Delta(y, t)) < \Pi_2(z_1^*(y, t), \Delta(y, t))$ . The last two inequalities together yield a contradiction to the optimality of  $z^*(y, t)$  for  $\lambda_P \Pi_1(\cdot, \Delta(y, t)) + \lambda_R \beta_R \Pi_2(\cdot, \Delta(y, t))$ . Hence,  $z^*(y, t) \geq z_1^*(y, t)$ .

Similarly, suppose for a contradiction that  $z^*(y, t) > z_2^*(y, t)$ . By definition of  $z_2^*(y, t)$ , we have  $\Pi_2(z_2^*(y, t), \Delta(y, t)) \geq \Pi_2(z^*(y, t), \Delta(y, t))$ , and by Lemma 3(b), we obtain  $\Pi_1(z_2^*(y, t), \Delta(y, t)) > \Pi_1(z^*(y, t), \Delta(y, t))$ . The last two inequalities together yield a contradiction to the optimality of  $z^*(y, t)$ , so we must have  $z^*(y, t) \leq z_2^*(y, t)$ . The result follows.

□

**Proof of Proposition 3:** The proposition follows from the more general result proven in Lemma 5. □

**Lemma 1** Define  $\theta_i(p, \Delta) = (p - \Delta) \bar{F}_{P_i}(p)$ ,  $i = 1, 2$ . Let  $\eta_i(y, t) = \arg \max_p \{\theta_i(p, \Delta(y, t))\}$ ,  $i = 1, 2$ . Let  $z_1^*(y, t)$ ,  $z_2^*(y, t)$  and  $z^*(y, t)$  be as defined by (9), (10) and (11). Then:

- (a)  $\eta_2(y, t) \leq \eta_1(y, t)$ ,
- (b)  $\frac{d\theta_i(p, \Delta)}{dp} > (=)(<)0$  for  $p < (=)(>)\eta_i(y, t)$ ,
- (c)  $z_1^*(y, t) \in [\eta_2(y, t), \eta_1(y, t)]$ ,  $z_2^*(y, t) \in [\eta_2(y, t), \eta_1(y, t)]$ ,  $z^*(y, t) \in [\eta_2(y, t), \eta_1(y, t)]$ .

**Proofs of (a) and (b):** It is not difficult to show that  $\theta_i(p, \Delta)$  is strictly unimodal in  $p$  due to assumption (A2). (See, for example, Lariviere and Porteus, 2001.) Hence,  $\eta_i(y, t)$  must satisfy the first-order condition (FOC) for  $\theta_i(p, \Delta(y, t))$  with respect to  $p$ . The FOC for  $\theta_i(p, \Delta(y, t))$ ,  $i = 1, 2$ , with respect to  $p$  is given by:

$$\frac{d\theta_i(p, \Delta)}{dp} = \bar{F}_{P_i}(p) - (p - \Delta) f_{P_i}(p) = \bar{F}_{P_i}(p) \left( 1 - (p - \Delta) \frac{f_{P_i}(p)}{\bar{F}_{P_i}(p)} \right) = 0, i = 1, 2, \quad (\text{A-2})$$

Now the FOCs in (A-2) along with assumption (A3) yields part (a). Part (b) follows directly from the unimodality of  $\theta_i(p, \Delta)$  in  $p$ .

**Proof of (c):** Recall that  $z_1^*(y, t) = \inf\{p^* : \Pi_1(p^*, \Delta(y, t)) \geq \Pi_1(p, \Delta(y, t)), \forall p\}$ . From (2) and (6), note that  $\Pi_1(p, \Delta(y, t)) = q_{P_1} \theta_1(p, \Delta(y, t)) + q_{P_2} \theta_2(p, \Delta(y, t))$ . Since  $\theta_i(p, \Delta(y, t))$  is unimodal in  $p$  and  $\eta_2(y, t) \leq \eta_1(y, t)$ , it follows that  $\Pi_1(p, \Delta(y, t))$  is increasing in  $p$  in the region where  $p < \eta_2(y, t)$  and decreasing in  $p$  in the region where  $p > \eta_1(y, t)$ . Hence, it must be that  $z_1^*(y, t) \in [\eta_2(y, t), \eta_1(y, t)]$ . The proofs of  $z_2^*(y, t) \in [\eta_2(y, t), \eta_1(y, t)]$  and  $z^*(y, t) \in [\eta_2(y, t), \eta_1(y, t)]$  are similar. □

**Proof of Proposition 4 :** We will consider two separate cases:

**Case 1:**  $\delta_{11} + \delta_{22} < 1$  — From Theorem 1(a), we have that  $p_t^*(y) = z_1^*(y, t)$ . We know that  $z_1^*(y, t)$  is increasing in  $\Delta(y, t)$  (by Lemma 6). Now noting that  $\Delta(y, t)$  is decreasing in  $y$  and increasing in  $t$  (by Proposition 3), we conclude that  $p_t^*(y)$  is decreasing in  $y$  and increasing in  $t$ . As for  $p_t^*(y) - d_t^*(y)$ , we note from Theorem 1(a) that  $p_t^*(y) - d_t^*(y) = z_2^*(y, t)$ , and we know that  $z_2^*(y, t)$  is increasing in  $\Delta(y, t)$  (by Lemma 6). Since  $\Delta(y, t)$  is decreasing in  $y$  and increasing in  $t$  (by Proposition 3), it follows that  $p_t^*(y) - d_t^*(y)$  is decreasing in  $y$  and increasing in  $t$ .

**Case 2:**  $\delta_{11} + \delta_{22} \geq 1$  — From Theorem 1(b), we have that

$$p_t^*(y) = p_t^*(y) - d_t^*(y) = z^*(y, t) := \inf\{p^* : \Pi(p^*, \Delta(y, t)) \geq \Pi(p, \Delta(y, t)), \forall p\}$$

where  $\Pi(p, \Delta(y, t)) = \lambda_P \Pi_1(p, \Delta(y, t)) + \lambda_R \beta_R \Pi_2(p, \Delta(y, t))$ . After noting that  $z^*(y, t)$  is increasing in  $\Delta(y, t)$  (by Lemma 6), the desired result follows since  $\Delta(y, t)$  is decreasing in  $y$  and increasing in  $t$  (by Proposition 3).  $\square$

**Lemma 2** *Suppose that  $\delta_{11} + \delta_{22} < (>)(=)1$ . Then,  $\frac{d(\Pi_1(p, \Delta(y, t)) - \Pi_2(p, \Delta(y, t)))}{dp} > (<)(=)0$  for  $p \in (\eta_2(y, t), \eta_1(y, t))$ .*

**Proof of Lemma 2:** We have

$$\Pi_1(p, \Delta(y, t)) - \Pi_2(p, \Delta(y, t)) = (\beta_P(p) - \alpha(p))(p - \Delta(y, t)).$$

Then,

$$\begin{aligned} \frac{d(\Pi_1(p, \Delta(y, t)) - \Pi_2(p, \Delta(y, t)))}{dp} &= (\beta_P(p) - \alpha(p)) + (\beta'_P(p) - \alpha'(p))(p - \Delta(y, t)) \\ &= (q_{P1} - \widehat{q}_{P1})(\overline{F}_{P1}(p) - (p - \Delta(y, t))f_{P1}(p)) \\ &\quad + (q_{P2} - \widehat{q}_{P2})(\overline{F}_{P2}(p) - (p - \Delta(y, t))f_{P2}(p)) \\ &= (q_{P1} - \widehat{q}_{P1}) \\ &\quad \times [\overline{F}_{P1}(p) - (p - \Delta(y, t))f_{P1}(p) - \overline{F}_{P2}(p) + (p - \Delta(y, t))f_{P2}(p)] \end{aligned}$$

From Proposition 2, we know that  $q_{P1} - \widehat{q}_{P1} > (<)(=)0$  if  $\delta_{11} + \delta_{22} < (>)(=)1$ . Furthermore, observe that the term in brackets is  $\frac{d\theta_1(p, \Delta)}{dp} - \frac{d\theta_2(p, \Delta)}{dp}$ , which is strictly positive for  $p \in (\eta_2(y, t), \eta_1(y, t))$  by Lemma 1(b). The result now follows.  $\square$

**Lemma 3** Let  $x_1$  and  $x_2$  be such that  $\eta_2(y, t) < x_1 < x_2 < \eta_1(y, t)$ . Suppose that  $\delta_{11} + \delta_{22} < (>)(=)1$ .

(a) If  $\Pi_1(x_1, \Delta) \geq (\leq)(=)\Pi_1(x_2, \Delta)$ , then  $\Pi_2(x_1, \Delta) > (<)(=)\Pi_2(x_2, \Delta)$ .

(b) If  $\Pi_2(x_1, \Delta) \leq (\geq)(=)\Pi_2(x_2, \Delta)$ , then  $\Pi_1(x_1, \Delta) < (>)(=)\Pi_1(x_2, \Delta)$

**Proof of Lemma 3:** Suppose  $\delta_{11} + \delta_{22} < 1$ . Then, by Lemma 2, we have  $\Pi_1(x_1, \Delta) - \Pi_2(x_1, \Delta) < \Pi_1(x_2, \Delta) - \Pi_2(x_2, \Delta)$ . In addition, suppose  $\Pi_1(x_1, \Delta) \geq \Pi_1(x_2, \Delta)$ . The last two inequalities together imply that  $\Pi_2(x_1, \Delta) > \Pi_2(x_2, \Delta)$ , which completes the proof of part (a) for  $\delta_{11} + \delta_{22} < 1$ . Symmetric arguments yield part (b). The proofs for  $\delta_{11} + \delta_{22} > 1$  and  $\delta_{11} + \delta_{22} = 1$  are similar.  $\square$

**Lemma 4** Suppose that  $\delta_{11} + \delta_{22} < (>)(=)1$ . Then,  $z_1^*(y, t) > (<)(=)z_2^*(y, t)$ .

**Proof of Lemma 4:** We will prove the result for  $\delta_{11} + \delta_{22} < 1$ ; the proofs of the other cases are similar. The proof is by contradiction. Suppose  $z_1^*(y, t) < z_2^*(y, t)$ . By definition of  $z_2^*(y, t)$ , we have

$$\Pi_2(z_1^*(y, t), \Delta(y, t)) < \Pi_2(z_2^*(y, t), \Delta(y, t)).$$

Then, applying Lemma 3(b) with  $x_1 = z_1^*(y, t)$  and  $x_2 = z_2^*(y, t)$ , we have

$$\Pi_1(z_1^*(y, t), \Delta(y, t)) < \Pi_1(z_2^*(y, t), \Delta(y, t)),$$

which is a contradiction to the optimality of  $z_1^*(y, t)$  for  $\Pi_1(\cdot, \Delta(y, t))$ . Hence, when  $\delta_{11} + \delta_{22} < 1$ , we must have  $z_1^*(y, t) \geq z_2^*(y, t)$ . It still remains to show that  $z_1^*(y, t) \neq z_2^*(y, t)$ . To that end, suppose for a contradiction that  $z_1^*(y, t) = z_2^*(y, t)$ . Note that we must have  $\left. \frac{d\Pi_2(p, \Delta(y, t))}{dp} \right|_{p=z_2^*(y, t)} = 0$  (since  $z_2^*(y, t)$  is an interior optimizer of  $\Pi_2(\cdot, \Delta(y, t))$ ). Hence, by Lemma 2, we must have  $\left. \frac{d\Pi_1(p, \Delta(y, t))}{dp} \right|_{p=z_1^*(y, t)} > 0$ , which is a contradiction to the optimality of  $z_1^*(y, t)$  for  $\Pi_1(\cdot, \Delta(y, t))$ . Therefore, we cannot have  $z_1^*(y, t) = z_2^*(y, t)$ , which concludes the proof for  $\delta_{11} + \delta_{22} < 1$ .  $\square$

**Lemma 5** Consider a slightly more general version of the DP formulation in (5):

$$\begin{aligned} V_t(y) &= V_{t-1}(y) + \max_{p_t \in A, d_t \in B(p_t)} \{ \lambda_P \beta_P(p_t) (p_t + V_{t-1}(y-1) - V_{t-1}(y)) \\ &\quad + \lambda_R \beta_R [\alpha(p_t - d_t) (p_t - d_t + V_{t-1}(y-1) - V_{t-1}(y))] \} \end{aligned} \quad (\text{A-3})$$

where both  $A$  and  $B$  are non-negative sets of real numbers,  $B$  possibly depends on  $p_t$  and for any  $d_t \in B(p_t)$ ,  $p_t - d_t \geq 0$ . Then, we have:

$$(a) V_{t+1}(y+1) - V_{t+1}(y) \geq V_t(y+1) - V_t(y), \text{ or } \Delta(y+1, t) \geq \Delta(y+1, t-1).$$

$$(b) V_{t+1}(y) - V_t(y) \geq V_{t+2}(y) - V_{t+1}(y)$$

$$(c) V_t(y+1) - V_t(y) \geq V_t(y+2) - V_t(y+1), \text{ or } \Delta(y+1, t-1) \geq \Delta(y+2, t-1).$$

**Proof of Lemma 5:** Following Bitran and Mondschein (1993), we will use an inductive argument on  $k = y + t$ . For  $k = 0$ , all inequalities hold. Now, assume that all inequalities hold for  $y + t < k$ . We will prove that they also hold for  $y + t = k$ .

**Proof of (a):** Since there exists an optimal solution, for some  $p^o \in A$  and  $d^o \in B(p^o)$ , we have

$$\begin{aligned} V_{t+1}(y) &= V_t(y) + \lambda_P \beta_P(p^o)(p^o + V_t(y-1) - V_t(y)) \\ &\quad + \lambda_R \beta_R [\alpha(p^o - d^o)(p^o - d^o + V_t(y-1) - V_t(y))] \end{aligned}$$

which can also be written as

$$\begin{aligned} V_{t+1}(y) - V_t(y) &= \lambda_P \beta_P(p^o)(p^o + V_t(y-1) - V_t(y)) \\ &\quad + \lambda_R \beta_R [\alpha(p^o - d^o)(p^o - d^o + V_t(y-1) - V_t(y))]. \end{aligned} \quad (\text{A-4})$$

Then, we also have

$$\begin{aligned} V_{t+1}(y+1) &\geq V_t(y+1) + \lambda_P \beta_P(p^o)(p^o + V_t(y) - V_t(y+1)) \\ &\quad + \lambda_R \beta_R [\alpha(p^o - d^o)(p^o - d^o + V_t(y) - V_t(y+1))], \end{aligned}$$

which can also be written as

$$\begin{aligned} V_{t+1}(y+1) - V_t(y+1) &\geq \lambda_P \beta_P(p^o)(p^o + V_t(y) - V_t(y+1)) \\ &\quad + \lambda_R \beta_R [\alpha(p^o - d^o)(p^o - d^o + V_t(y) - V_t(y+1))]. \end{aligned} \quad (\text{A-5})$$

By the induction assumption for (c),

$$V_t(y) - V_t(y+1) \geq V_t(y-1) - V_t(y).$$

Using this in (A-5), we get

$$\begin{aligned} V_{t+1}(y+1) - V_t(y+1) &\geq \lambda_P \beta_P(p^o)(p^o + V_t(y-1) - V_t(y)) \\ &\quad + \lambda_R \beta_R [\alpha(p^o - d^o)(p^o - d^o + V_t(y-1) - V_t(y))]. \end{aligned} \quad (\text{A-6})$$

Finally, from (A-4) and (A-6), we conclude that

$$V_{t+1}(y+1) - V_{t+1}(y) \geq V_t(y+1) - V_t(y).$$

**Proof of (b):** Since there exists an optimal solution, for some  $p^o \in A$  and  $d^o \in B(p^o)$ , we have

$$\begin{aligned} V_{t+2}(y) &= V_{t+1}(y) + \lambda_P \beta_P(p^o)(p^o + V_{t+1}(y-1) - V_{t+1}(y)) \\ &\quad + \lambda_R \beta_R [\alpha(p^o - d^o)(p^o - d^o + V_{t+1}(y-1) - V_{t+1}(y))], \end{aligned}$$

which can also be written as

$$\begin{aligned} V_{t+2}(y) - V_{t+1}(y) &= \lambda_P \beta_P(p^o)(p^o + V_{t+1}(y-1) - V_{t+1}(y)) \\ &\quad + \lambda_R \beta_R [\alpha(p^o - d^o)(p^o - d^o + V_{t+1}(y-1) - V_{t+1}(y))]. \end{aligned} \quad (\text{A-7})$$

Then, we also have

$$\begin{aligned} V_{t+1}(y) &\geq V_t(y) + \lambda_P \beta_P(p^o)(p^o + V_t(y-1) - V_t(y)) \\ &\quad + \lambda_R \beta_R [\alpha(p^o - d^o)(p^o - d^o + V_t(y-1) - V_t(y))], \end{aligned}$$

which can also be written as

$$\begin{aligned} V_{t+1}(y) - V_t(y) &\geq \lambda_P \beta_P(p^o)(p^o + V_t(y-1) - V_t(y)) \\ &\quad + \lambda_R \beta_R [\alpha(p^o - d^o)(p^o - d^o + V_t(y-1) - V_t(y))]. \end{aligned} \quad (\text{A-8})$$

By the induction assumption on (a),

$$V_t(y-1) - V_t(y) \geq V_{t+1}(y-1) - V_{t+1}(y).$$

Using this in (A-8), we get

$$\begin{aligned} V_{t+1}(y) - V_t(y) &\geq \lambda_P \beta_P(p^o)(p^o + V_{t+1}(y-1) - V_{t+1}(y)) \\ &\quad + \lambda_R \beta_R [\alpha(p^o - d^o)(p^o - d^o + V_{t+1}(y-1) - V_{t+1}(y))]. \end{aligned} \quad (\text{A-9})$$

Finally, from (A-7) and (A-9), we conclude that

$$V_{t+1}(y) - V_t(y) \geq V_{t+2}(y) - V_{t+1}(y).$$

**Proof of (c):** Following as in parts (a) and (b), we can show that for some  $p^\circ \in A$  and  $d^\circ \in B(p^\circ)$ , we have

$$\begin{aligned}
V_t(y+2) - V_{t-1}(y+1) &= \lambda_P \beta_P(p^\circ)(p^\circ + V_{t-1}(y+1) - V_{t-1}(y+2)) \\
&+ \lambda_R \beta_R [\alpha(p^\circ - d^\circ)(p^\circ - d^\circ + V_{t-1}(y+1) - V_{t-1}(y+2))] \\
&+ V_{t-1}(y+2) - V_{t-1}(y+1) \\
&= (V_{t-1}(y+2) - V_{t-1}(y+1))(1 - \lambda_P \beta_P(p^\circ) - \lambda_R \beta_R \alpha(p^\circ - d^\circ)) \\
&+ \lambda_P \beta_P(p^\circ)p^\circ + [\lambda_R \beta_R \alpha(p^\circ - d^\circ)(p^\circ - d^\circ)]. \tag{A-10}
\end{aligned}$$

and

$$\begin{aligned}
V_{t+1}(y+1) - V_t(y) &\geq \lambda_P \beta_P(p^\circ)(p^\circ + V_t(y) - V_t(y+1)) \\
&+ \lambda_R \beta_R [\alpha(p^\circ - d^\circ)(p^\circ - d^\circ + V_t(y) - V_t(y+1))] \\
&+ V_t(y+1) - V_t(y) \\
&= (V_t(y+1) - V_t(y))(1 - \lambda_P \beta_P(p^\circ) - \lambda_R \beta_R \alpha(p^\circ - d^\circ)) \\
&+ \lambda_P \beta_P(p^\circ)p^\circ + [\lambda_R \beta_R \alpha(p^\circ - d^\circ)(p^\circ - d^\circ)]. \tag{A-11}
\end{aligned}$$

By the induction assumptions on (a) and (c), we have

$$V_t(y+1) - V_t(y) \geq V_{t-1}(y+1) - V_{t-1}(y) \geq V_{t-1}(y+2) - V_{t-1}(y+1).$$

Using this together with (A-11), we get

$$\begin{aligned}
V_{t+1}(y+1) - V_t(y) &\geq (V_{t-1}(y+2) - V_{t-1}(y+1))(1 - \lambda_P \beta_P(p^\circ) - \lambda_R \beta_R \alpha(p^\circ - d^\circ)) \\
&+ \lambda_P \beta_P(p^\circ)p^\circ + [\lambda_R \beta_R \alpha(p^\circ - d^\circ)(p^\circ - d^\circ)]. \tag{A-12}
\end{aligned}$$

From (A-10) and (A-12), we have

$$V_{t+1}(y+1) - V_t(y) \geq V_t(y+2) - V_{t-1}(y+1).$$

Also, using part (b) for  $(y+1, t-1)$  (note that (ii) has already been verified for  $y+t=k$ ), we have

$$V_t(y+1) - V_{t-1}(y+1) \geq V_{t+1}(y+1) - V_t(y+1).$$

Adding up both inequalities, it follows that

$$V_t(y+1) - V_t(y) \geq V_t(y+2) - V_t(y+1).$$

□



**Lemma 6** Let  $\Pi(p, \Delta) = a\Pi_1(p, \Delta) + b\Pi_2(p, \Delta)$  where  $a \geq 0, b \geq 0$  and  $\Pi_2(p, \Delta)$  and  $\Pi_1(p, \Delta)$  are as defined by (6) and (7). Let  $p_1^*(\Delta) = \inf\{p^* : \Pi_1(p^*, \Delta) \geq \Pi_1(p, \Delta), \forall p\}$ ,  $p_2^*(\Delta) = \inf\{p^* : \Pi_2(p^*, \Delta) \geq \Pi_2(p, \Delta), \forall p\}$  and  $p^*(\Delta) = \inf\{p^* : \Pi(p^*, \Delta) \geq \Pi(p, \Delta), \forall p\}$ . Then,  $p_1^*(\Delta)$ ,  $p_2^*(\Delta)$  and  $p^*(\Delta)$  are all increasing in  $\Delta$ .

**Proof of Lemma 6:** We will first prove the result for  $p_1^*(\Delta)$ . Let  $C := \{(p, \Delta) : \Delta > 0 \text{ and } p > \Delta\}$ . By Theorem 8.1 on p.124 of Porteus (2002), it is sufficient to show that  $\Pi_1(p, \Delta)$  is supermodular on  $C$ . Hence, we need to prove that the following inequality is true for any  $x = (x_1, x_2) \in C$  and  $y = (y_1, y_2) \in C$ :

$$\Pi_1(x \wedge y) + \Pi_1(x \vee y) \geq \Pi_1(x) + \Pi_1(y) \quad (\text{A-13})$$

where  $x \wedge y = (\min(x_1, y_1), \min(x_2, y_2))$  and  $x \vee y = (\max(x_1, y_1), \max(x_2, y_2))$ . We will consider two cases:

Case 1,  $x_1 \geq y_1$  and  $x_2 \geq y_2$ : In this case, the desired inequality holds trivially since  $x \wedge y = y$  and  $x \vee y = x$ .

Case 2,  $x_1 \geq y_1$  and  $x_2 < y_2$ : By substituting  $\Pi_1(p, \Delta) = \beta_P(p)(p - \Delta)$  in (A-13) and after some algebra, one can show that the inequality in (A-13) is equivalent to the following:

$$\Pi_1(x \wedge y) + \Pi_1(x \vee y) - \Pi_1(x) - \Pi_1(y) = (y_2 - x_2) (\beta_P(y_1) - \beta_P(x_1)) \geq 0$$

The above inequality holds under the assumptions of Case 2 due to the fact that  $\beta_P(x)$  is decreasing in  $x$ . This concludes the proof of the supermodularity of  $\Pi_1(p, \Delta)$ 's on  $C$ , which allows us to conclude  $p_1^*(\Delta)$  is increasing in  $\Delta$ .

As for  $p_2^*(\Delta)$ , the proof is similar, and uses the fact that  $\Pi_2(p, \Delta)$  is supermodular. Finally, since  $\Pi_1(p, \Delta)$  and  $\Pi_2(p, \Delta)$  are supermodular,  $\Pi(p, \Delta) = a\Pi_1(p, \Delta) + b\Pi_2(p, \Delta)$  is supermodular (by Lemma 8.3 on p.123 of Porteus (2002)), and this allows us to conclude that  $p^*(\Delta)$  is increasing in  $\Delta$ .  $\square$

## Appendix B - Proofs for Section 4.2

Proposition 5 corresponds to parts (b), (d) and (e) of the following proposition and Theorem 2 corresponds to part (c).

**Proposition 8** *Let  $d_t^*(x, y)$  denote the largest optimal discount corresponding to the smallest optimal price when there are  $t$  time periods to go, there are  $x$  units of the regular product and  $y$  units of the promotional product. Suppose that  $\delta_{11} + \delta_{22} < 1$ . Then,*

(a)  $V_t(x + 1, y) - V_t(x, y) \geq 0$ ,  $V_t(x, y + 1) - V_t(x, y) \geq 0$  for  $x \geq 0$ ,  $y \geq 0$ ,  $t \geq 0$

(b)  $\Delta(x + 1, y + 1, t) \geq \Delta(x, y + 1, t)$  for  $x \geq 0$ ,  $y \geq 0$ ,  $t \geq 1$

(c)  $d_t^*(x, y) > 0$  for  $x \geq 0$ ,  $y \geq 1$ ,  $t \geq 1$ .

(d)  $\Delta(x, y + 2, t) \leq \Delta(x, y + 1, t)$  for  $x \geq 0$ ,  $y \geq 0$ ,  $t \geq 1$

(e)  $\Delta(x, y + 1, t + 1) \geq \Delta(x, y + 1, t)$  for  $x \geq 0$ ,  $y \geq 0$ ,  $t \geq 1$

**Proof of Proposition 8:** We skip the proof of part (a) since it immediately follows from (12) with a simple induction argument. Throughout the proof, recall the definition that

$$\Delta(x, y + 1, t) = V_{t-1}(x, y + 1) - V_{t-1}(x, y) \text{ for any } x, y \geq 0 \text{ and } t \geq 1.$$

In this proof, we utilize Lemma A1 of Netessine, Savin, and Xiao (2006), stated as Lemma 8 at the end of Appendix B.

**Proof of (b):** The proof is by induction. Note that (b) holds trivially when  $t = 1$  or  $y = 0$ . Suppose that for some  $t > 1$ ,

$$\Delta(x + 1, y + 1, t) \geq \Delta(x, y + 1, t) \tag{A-14}$$

We need to prove that

$$\Delta(x + 1, y, t + 1) \geq \Delta(x, y, t + 1)$$

It is sufficient to consider two different cases.

**Case 1:**  $x, y \geq 1$ .

Since  $\delta_{11} + \delta_{22} < 1$ , using (A-14) and Lemma 7, it follows that  $d_t^*(x, y) > 0$ . Therefore, we

can use (12) to write

$$\begin{aligned}
V_t(x+1, y+1) - V_t(x, y+1) = & \\
& \lambda_R \beta_R \left[ V_{t-1}(x, y+1) - V_{t-1}(x-1, y+1) + \max_z \{ \alpha(z) (z + V_{t-1}(x, y) - V_{t-1}(x, y+1)) \} \right. \\
& \quad \left. - \max_z \{ \alpha(z) (z + V_{t-1}(x-1, y) - V_{t-1}(x-1, y+1)) \} \right] \\
& + \lambda_P \left[ V_{t-1}(x+1, y+1) - V_{t-1}(x, y+1) + \max_z \{ \beta_P(z) (z + V_{t-1}(x+1, y) - V_{t-1}(x+1, y+1)) \} \right. \\
& \quad \left. - \max_z \{ \beta_P(z) (z + V_{t-1}(x, y) - V_{t-1}(x, y+1)) \} \right] \\
& + (1 - \lambda_R \beta_R - \lambda_P) (V_{t-1}(x+1, y+1) - V_{t-1}(x, y+1)). \tag{A-15}
\end{aligned}$$

Define

$$g(x) = \max_z \{ \beta_P(z) (z - x) \} \text{ and } h(x) = \max_z \{ \alpha(z) (z - x) \}.$$

Using the above definition, we can rewrite (A-15) as:

$$\begin{aligned}
V_t(x+1, y+1) - V_t(x, y+1) = & \lambda_R \beta_R \left[ \begin{array}{c} V_{t-1}(x, y+1) - V_{t-1}(x-1, y+1) \\ + h(\Delta(x, y+1, t)) - h(\Delta(x-1, y+1, t)) \end{array} \right] \\
& + \lambda_P \left[ \begin{array}{c} V_{t-1}(x+1, y+1) - V_{t-1}(x, y+1) \\ + g(\Delta(x+1, y+1, t)) - g(\Delta(x, y+1, t)) \end{array} \right] \\
& + (1 - \lambda_R \beta_R - \lambda_P) (V_{t-1}(x+1, y+1) - V_{t-1}(x, y+1))
\end{aligned}$$

It now follows from Lemma 8 and (A-14) that

$$\begin{aligned}
V_t(x+1, y+1) - V_t(x, y+1) \geq & \lambda_R \beta_R \left[ \begin{array}{c} V_{t-1}(x, y+1) - V_{t-1}(x-1, y+1) \\ - \Delta(x, y+1, t) + \Delta(x-1, y+1, t) \end{array} \right] \\
& + \lambda_P \left[ \begin{array}{c} V_{t-1}(x+1, y+1) - V_{t-1}(x, y+1) \\ - \Delta(x+1, y+1, t) + \Delta(x, y+1, t) \end{array} \right] \\
& + (1 - \lambda_R \beta_R - \lambda_P) (V_{t-1}(x+1, y) - V_{t-1}(x, y))
\end{aligned}$$

Using the definition of  $\Delta(x, y, t)$ , we can simplify the inequality above:

$$\begin{aligned}
V_t(x+1, y+1) - V_t(x, y+1) \geq & \lambda_R \beta_R (V_{t-1}(x, y) - V_{t-1}(x-1, y)) \\
& + \lambda_P (V_{t-1}(x+1, y) - V_{t-1}(x, y)) \\
& + (1 - \lambda_R \beta_R - \lambda_P) (V_{t-1}(x+1, y) - V_{t-1}(x, y))
\end{aligned}$$

Applying again Lemma 8 and (A-14), we can write

$$\begin{aligned}
V_t(x+1, y+1) - V_t(x, y+1) &\geq \lambda_R \beta_R \left[ + \begin{array}{c} V_{t-1}(x, y) - V_{t-1}(x-1, y) \\ h(\Delta(x, y, t)) - h(\Delta(x-1, y, t)) \end{array} \right] \\
&\quad + \lambda_P \left[ + \begin{array}{c} V_{t-1}(x+1, y) - V_{t-1}(x, y) \\ g(\Delta(x+1, y, t)) - g(\Delta(x, y, t)) \end{array} \right] \\
&\quad + (1 - \lambda_R \beta_R - \lambda_P)(V_{t-1}(x+1, y) - V_{t-1}(x, y)) \\
&= V_t(x+1, y) - V_t(x, y),
\end{aligned}$$

where the last equality follows from (12). Thus, we have shown that

$$V_t(x+1, y+1) - V_t(x, y+1) \geq V_t(x+1, y) - V_t(x, y),$$

which is equivalent to  $\Delta(x+1, y, t+1) \geq \Delta(x, y, t+1)$ .

**Case 2:**  $x = 0, y \geq 1$ .

As in Case 1, since  $\delta_{11} + \delta_{22} < 1$ , we have  $d_t^*(x, y) > 0$ , and we can write

$$\begin{aligned}
V_t(1, y+1) - V_t(0, y+1) &= \\
&\lambda_R \beta_R \left[ V_{t-1}(0, y+1) + \max_z \{ \alpha(z) (z + V_{t-1}(0, y) - V_{t-1}(0, y+1)) \} \right] \\
&+ \lambda_P \left[ V_{t-1}(1, y+1) - V_{t-1}(0, y+1) + \max_z \{ \beta_P(z) (z + V_{t-1}(1, y) - V_{t-1}(1, y+1)) \} \right] \\
&\quad - \max_z \{ \beta_P(z) (z + V_{t-1}(0, y) - V_{t-1}(0, y+1)) \} \right] \\
&+ (1 - \lambda_R \beta_R - \lambda_P)(V_{t-1}(1, y+1) - V_{t-1}(0, y+1)),
\end{aligned}$$

which we can rewrite as

$$\begin{aligned}
V_t(1, y+1) - V_t(0, y+1) &= \lambda_R \beta_R [V_{t-1}(0, y+1) + h(\Delta(0, y+1, t))] \\
&\quad + \lambda_P \left[ + \begin{array}{c} V_{t-1}(1, y+1) - V_{t-1}(0, y+1) \\ g(\Delta(1, y+1, t)) - g(\Delta(0, y+1, t)) \end{array} \right] \\
&\quad + (1 - \lambda_R \beta_R - \lambda_P)(V_{t-1}(1, y+1) - V_{t-1}(0, y+1))
\end{aligned}$$

Since  $V_{t-1}(0, y+1) \geq V_{t-1}(0, y)$  (by Proposition 8(a)),  $\Delta(1, y+1, t) \geq \Delta(0, y+1, t)$  (by (A-14)) and  $\Delta(0, y+1, t) \leq \Delta(0, y, t)$  (by Proposition 3(b)), we obtain:

$$\begin{aligned}
V_t(1, y+1) - V_t(0, y+1) &\geq \lambda_R \beta_R [V_{t-1}(0, y) + h(\Delta(0, y, t))] \\
&\quad + \lambda_P (V_t(1, y) - V_t(0, y)) \\
&\quad + (1 - \lambda_R \beta_R - \lambda_P)(V_{t-1}(1, y+1) - V_{t-1}(0, y+1))
\end{aligned}$$

Since  $\Delta(1, y, t) \geq \Delta(0, y, t)$  (by (A-14)), we can write

$$\begin{aligned}
V_t(1, y+1) - V_t(0, y+1) &\geq \lambda_R \beta_R [V_{t-1}(0, y) + h(\Delta(0, y, t))] \\
&\quad + \lambda_P \left[ \begin{array}{l} V_{t-1}(1, y) - V_{t-1}(0, y) \\ + g(\Delta(1, y, t)) - g(\Delta(0, y, t)) \end{array} \right] \\
&\quad + (1 - \lambda_R \beta_R - \lambda_P)(V_{t-1}(1, y+1) - V_{t-1}(0, y+1)) \\
&= V_t(1, y) - V_t(0, y), \tag{A-16}
\end{aligned}$$

where the last equality follows from (12). Thus, we have shown that

$$V_t(1, y+1) - V_t(0, y+1) \geq V_t(1, y) - V_t(0, y),$$

which is equivalent to  $\Delta(1, y+1, t+1) \geq \Delta(0, y+1, t+1)$ , concluding the proof of (b).

**Proof of (c):** Part (c) follows from part (b) of the proposition and Lemma 7.

**Proof of (d):** The proof is by induction. Note that part (d) holds trivially when  $t = 1$ . Furthermore, part (d) holds for  $x = 0$  by Proposition 3(b). Suppose that for some  $t \geq 1$ ,

$$\Delta(x, y+2, t) \leq \Delta(x, y+1, t) \text{ for any } x \geq 1, y \geq 0 \tag{A-17}$$

We need to prove that

$$\Delta(x, y+2, t+1) \geq \Delta(x, y+1, t+1) \text{ for any } x \geq 1, y \geq 0$$

Part (c) of the proposition along with (12) allow us to write:

$$\begin{aligned}
\Delta(x, y+2, t+1) &= V_t(x, y+2) - V_t(x, y+1) = \\
&\lambda_R \beta_R \left[ \begin{array}{l} V_{t-1}(x-1, y+2) - V_{t-1}(x-1, y+1) \\ + \max_z \{ \alpha(z) (z + V_{t-1}(x-1, y+1) - V_{t-1}(x-1, y+2)) \} \\ - \max_z \{ \alpha(z) (z + V_{t-1}(x-1, y) - V_{t-1}(x-1, y+1)) \} \end{array} \right] \\
&+ \lambda_P \left[ \begin{array}{l} V_{t-1}(x, y+2) - V_{t-1}(x, y+1) \\ + \max_z \{ \beta_P(z) (z + V_{t-1}(x, y+1) - V_{t-1}(x, y+2)) \} \\ - \max_z \{ \beta_P(z) (z + V_{t-1}(x, y) - V_{t-1}(x, y+1)) \} \end{array} \right] \\
&+ (1 - \lambda_R \beta_R - \lambda_P)(V_{t-1}(x, y+2) - V_{t-1}(x, y+1)). \tag{A-18}
\end{aligned}$$

Using the definitions of  $h(x)$ ,  $g(x)$  and  $\Delta(x, y, t)$ , we can simplify the equality above:

$$\begin{aligned}
\Delta(x, y+2, t+1) &= \\
&\lambda_R \beta_R [\Delta(x-1, y+2, t) + h(\Delta(x-1, y+2, t)) - h(\Delta(x-1, y+1, t))] \\
&+ \lambda_P [\Delta(x, y+2, t) + g(\Delta(x, y+2, t)) - g(\Delta(x, y+1, t))] \\
&+ (1 - \lambda_R \beta_R - \lambda_P) \Delta(x, y+2, t).
\end{aligned}$$

Using Lemma 8 and (A-17), we obtain from the above equality

$$\Delta(x, y + 2, t + 1) \leq \lambda_R \beta_R \Delta(x - 1, y + 1, t) + \lambda_P \Delta(x, y + 1, t) + (1 - \lambda_R \beta_R - \lambda_P) \Delta(x, y + 2, t)$$

Since  $\Delta(x - 1, y + 1, t) \leq \Delta(x - 1, y, t)$ ,  $\Delta(x, y + 1, t) \leq \Delta(x, y, t)$  and  $\Delta(x, y + 2, t) \leq \Delta(x, y + 1, t)$  (all by (A-17)), we can write

$$\begin{aligned} \Delta(x, y + 2, t + 1) &\leq \\ &\lambda_R \beta_R [\Delta(x - 1, y + 1, t) + h(\Delta(x - 1, y + 1, t)) - h(\Delta(x - 1, y, t))] \\ &+ \lambda_P [\Delta(x, y + 1, t) + g(\Delta(x, y + 1, t)) - g(\Delta(x, y, t))] \\ &+ (1 - \lambda_R \beta_R - \lambda_P) \Delta(x, y + 1, t). \\ &= V_t(x, y + 1) - V_t(x, y) = \Delta(x, y + 1, t + 1), \end{aligned}$$

where the next to last equality follows from (12). This concludes the proof of (d).

**Proof of (e):** Given the result proven in part (d) of the proposition, the proof of part (e) follows the same line of reasoning as the proof of Proposition 3(a).  $\square$

Before we state and prove Lemma 7, we first introduce new notation. For  $x, y \geq 1$ , we can rewrite (12) as

$$\begin{aligned} V_t(x, y) &= (1 - \lambda_R \beta_R) V_{t-1}(x, y) + \lambda_R \beta_R V_{t-1}(x - 1, y) \\ &+ \max_{p \geq d \geq 0} \{ \lambda_R \beta_R \alpha(p - d)(p - d - \Delta(x - 1, y, t)) + \lambda_P \beta_P(p)(p - \Delta(x, y, t)) \} \end{aligned}$$

Let  $\Pi_1(p, \Delta)$  and  $\Pi_2(p, \Delta)$  be as defined by (6) and (7), that is:

$$\Pi_1(p, \Delta) = \beta_P(p)(p - \Delta) \text{ and } \Pi_2(p, \Delta) = \alpha(p)(p - \Delta).$$

Then, in period  $t$ , a firm with  $x$  units of regular product and  $y$  units of the promotional product is solving the following single-stage optimization problem:

$$\max_{p \geq d \geq 0} \{ \lambda_P \Pi_1(p, \Delta(x, y, t)) + \lambda_R \beta_R \Pi_2(p - d, \Delta(x - 1, y, t)) \}.$$

Finally, define

$$z_1^*(\Delta) = \inf \{ p^* : \Pi_1(p^*, \Delta) \geq \Pi_1(p, \Delta) \ \forall p \}$$

and

$$z_2^*(\Delta) = \inf \{ p^* : \Pi_2(p^*, \Delta) \geq \Pi_2(p, \Delta) \ \forall p \}.$$

**Lemma 7** *Suppose that for some fixed  $t > 0$ , we have*

$$\Delta(x + 1, y + 1, t) \geq \Delta(x, y + 1, t) \quad (\text{A-19})$$

for all  $x, y \geq 0$ . Then, if  $\delta_{11} + \delta_{22} < 1$ , we have  $z_1^*(\Delta(x, y, t)) > z_2^*(\Delta(x - 1, y, t))$  and  $d_t^*(x, y) = z_1^*(\Delta(x, y, t)) - z_2^*(\Delta(x - 1, y, t)) > 0$  for all  $x \geq 1$  and  $y \geq 0$ .

**Proof of Lemma 7:** First, we note that  $z_1^*(\Delta(x - 1, y, t)) > z_2^*(\Delta(x - 1, y, t))$ . (The proof of this result follows as in Lemma 4 and is therefore omitted.) It now remains to show that  $z_1^*(\Delta(x, y, t)) \geq z_1^*(\Delta(x - 1, y, t))$ . Given (A-19), it is easy to verify that

$$\frac{d}{dp} [\Pi_1(p, \Delta(x, y, t)) - \Pi_1(p, \Delta(x - 1, y, t))] = \frac{d\beta_P(p)}{dp} (\Delta(x - 1, y, t) - \Delta(x, y, t)) \geq 0. \quad (\text{A-20})$$

From the definition of  $z_1^*(\Delta(x - 1, y, t))$ ,

$$\Pi_1(z_1^*(\Delta(x, y, t)), \Delta(x - 1, y, t)) \leq \Pi_1(z_1^*(\Delta(x - 1, y, t)), \Delta(x - 1, y, t)).$$

Now, suppose for contradiction that  $z_1^*(\Delta(x, y, t)) < z_1^*(\Delta(x - 1, y, t))$ . Then, from (A-20), it follows that

$$\Pi_1(z_1^*(\Delta(x, y, t)), \Delta(x, y, t)) \leq \Pi_1(z_1^*(\Delta(x - 1, y, t)), \Delta(x, y, t)),$$

which is a contradiction to the definition of  $z_1^*(\Delta(x, y, t))$ . Hence,  $z_1^*(\Delta(x, y, t)) \geq z_1^*(\Delta(x - 1, y, t))$  and the result follows.  $\square$

**Proof of Proposition 6:** First, use Lemma 6 to observe that  $z_1^*(\Delta)$  and  $z_2^*(\Delta)$  are both increasing in  $\Delta$ . Now, by Proposition 8(b) and Lemma 7, note that in period  $t$  with  $x$  units of regular inventory and  $y$  units of promotional inventory, the optimal announced price will be given by  $z_1^*(\Delta(x, y, t))$  and the optimal upsell price will be given by  $z_1^*(\Delta(x, y, t))$ . The monotonicity results now follow from Proposition 8(b)–(d).  $\square$

**Lemma 8 (Netessine, Savin, and Xiao)** *Let  $g(x) = \max_{z \in A} \{\theta(z)(z - x)\}$  where  $\theta(z)$  is non-decreasing in  $z$  with  $0 \leq \theta(z) \leq 1$ . Then, for any  $x_1, x_2 \geq 0$ ,*

$$\max(0, x_2 - x_1) \geq g(x_1) - g(x_2) \geq \min(0, x_2 - x_1).$$

**Proof of Lemma 8:** For a proof, see Netessine, Savin, and Xiao (2006).  $\square$

## Appendix C - Model details and proofs for Section 4.3

We will require some changes to the notation defined earlier. The following is a list of the notation to be used in this section:

- $r_k$  : exogenously fixed price of regular product  $k$
- $p_t$  : announced price of the promotional product in period  $t$
- $d_{tk}$  : discount offered on the promotional product in the upsell stage to a customer who purchased regular product  $k$  in the initial stage of period  $t$
- $\lambda_k$  : the probability that a potential customer for regular item  $k$  will arrive in a given period
- $\lambda_P$  : the probability that a potential customer for the promotional item will arrive in a given period
- $F_{ik}$  : the cdf of the reservation price of segment  $i$  customers for regular item  $k$
- $F_{Pi}$  : the cdf of the reservation price of segment  $i$  customers for the promotional item
- $q_{ik}$  : the probability that a customer belongs to segment  $i$  for regular item  $k$
- $q_{Pi}$  : the probability that a customer belongs to segment  $i$  for the promotional item
- $\widehat{q}_{Pik}$  : the probability that a customer belongs to segment  $i$  for the promotional item given that the customer bought regular item  $k$  in the initial stage
- $\delta_{ijk}$  : the probability that a customer belongs to segment  $j$  for the promotional item given that she belongs to segment  $i$  for regular item  $k$
- $\beta_k$  : the probability that a potential customer for regular product  $k$  will buy it at the exogenously-fixed price
- $\beta_P(x)$  : the probability that a potential customer for the promotional product will buy it at price  $x$
- $\alpha_k(x)$  : given that a customer bought regular item  $k$  in the initial stage, the probability that the customer will buy the promotional item in the upsell stage at price  $x$

Given the notation above,  $\beta_k$ ,  $\beta_P(x)$  and  $\alpha_k(x)$  are obtained from  $q_{ik}$ ,  $q_{Pi}$  and  $\widehat{q}_{Pik}$  as before. The optimality equations, given by (5) for the single regular product case, can now be



modified as follows:

$$\begin{aligned}
V_t(y) &= V_{t-1}(y) + \max_{p_t, d_{kt}: p_t \geq d_{kt} \geq 0} \left\{ \lambda_P \beta_P(p_t)(p_t + V_{t-1}(y-1) - V_{t-1}(y)) \right. \\
&\quad \left. + \sum_{k=1}^n \lambda_k \beta_k \alpha_k(p_t - d_{tk})(p_t - d_{tk} + V_{t-1}(y-1) - V_{t-1}(y)) \right\}, \\
&\quad y > 0, t = 1, \dots, T
\end{aligned} \tag{A-21}$$

Once again, define  $\Delta(y, t) = V_{t-1}(y) - V_{t-1}(y-1)$  and

$$\begin{aligned}
\Pi_1(p, \Delta) &= \beta_P(p)(p - \Delta), \\
\Pi_{2k}(p, \Delta) &= \alpha_k(p)(p - \Delta), k = 1, \dots, n.
\end{aligned}$$

Using the definitions above, we can write the optimality equations in (A-21) in the following alternative form:

$$\max_p \left\{ \lambda_P \Pi_1(p, \Delta(y, t)) + \sum_{k=1}^n \max_{0 \leq p_k \leq p} \{ \lambda_k \beta_k \Pi_{2k}(p_k, \Delta(y, t)) \} \right\} \tag{A-22}$$

As in the single regular product case, define  $z_1^*(y, t)$  and  $z_{2k}^*(t, y)$  as:

$$\begin{aligned}
z_1^*(y, t) &= \inf \{ p^* : \Pi_1(p^*, \Delta(y, t)) \geq \Pi_1(p, \Delta(y, t)), \forall p \} \\
z_{2k}^*(t, y) &= \inf \{ p^* : \Pi_{2k}(p^*, \Delta(y, t)) \geq \Pi_{2k}(p, \Delta(y, t)), \forall p \}
\end{aligned}$$

In addition, let  $p_t^*(y)$  denote the optimal announced price of the promotional item with  $y$  units in inventory and  $t$  periods to go. (We pick the smallest maximizer when multiple maximizers exist.) It is given by

$$\begin{aligned}
p_t^*(y) &= \inf \left\{ p^* : \lambda_P \Pi_1(p^*, \Delta(y, t)) + \sum_{k=1}^n \max_{0 \leq p_k \leq p^*} \{ \lambda_k \beta_k \Pi_{2k}(p_k, \Delta(y, t)) \} \right. \\
&\quad \left. \geq \lambda_P \Pi_1(p, \Delta(y, t)) + \sum_{k=1}^n \max_{0 \leq p_k \leq p} \{ \lambda_k \beta_k \Pi_{2k}(p_k, \Delta(y, t)) \}, \forall p \right\}
\end{aligned}$$

In order to prove Theorem 3, we will first prove a number of lemmas. The following lemma is analogous to Lemma 4, and it goes through as before:

**Lemma 9** *For a given  $k \in \{1, \dots, n\}$ , suppose that  $\delta_{11k} + \delta_{22k} < (>)(=)1$ . Then,  $z_1^*(y, t) > (<)(=)z_{2k}^*(y, t)$ .*

Next, we prove three new lemmas:

**Lemma 10** For any  $i, j \in \{1, \dots, n\}$ , if  $\widehat{q}_{P1i} - \widehat{q}_{P1j} > (<)(=)0$ , then  $\frac{d(\Pi_{2i}(p, \Delta(y, t)) - \Pi_{2j}(p, \Delta(y, t)))}{dp} > (<)(=)0$  for  $p \in (\eta_2(y, t), \eta_1(y, t))$ .

**Proof of Lemma 10:** We have

$$\Pi_{2i}(p, \Delta(y, t)) - \Pi_{2j}(p, \Delta(y, t)) = (\alpha_i(p) - \alpha_j(p))(p - \Delta(y, t)).$$

Then,

$$\begin{aligned} \frac{d(\Pi_{2i}(p, \Delta(y, t)) - \Pi_{2j}(p, \Delta(y, t)))}{dp} &= (\alpha_i(p) - \alpha_j(p)) + (\alpha'_i(p) - \alpha'_j(p))(p - \Delta(y, t)) \\ &= (\widehat{q}_{P1i} - \widehat{q}_{P1j})(\overline{F}_{P1}(p) - (p - \Delta(y, t))f_{P1}(p)) \\ &\quad + (\widehat{q}_{P2i} - \widehat{q}_{P2j})(\overline{F}_{P2}(p) - (p - \Delta(y, t))f_{P2}(p)) \\ &= (\widehat{q}_{P1i} - \widehat{q}_{P1j}) \\ &\quad \times [\overline{F}_{P1}(p) - (p - \Delta(y, t))f_{P1}(p) - \overline{F}_{P2}(p) + (p - \Delta(y, t))f_{P2}(p)] \end{aligned}$$

The result follows from the fact that the term in brackets above is strictly positive for  $p \in (\eta_2(y, t), \eta_1(y, t))$  (by Lemma 1(b)).  $\square$

**Lemma 11** For any  $i, j \in \{1, \dots, n\}$ , if  $\widehat{q}_{P1i} - \widehat{q}_{P1j} > (<)(=)0$ , then  $z_{2i}^*(y, t) > (<)(=)z_{2j}^*(y, t)$ .

**Proof of Lemma 11:** We will prove the result for the case where  $\widehat{q}_{P1i} - \widehat{q}_{P1j} > 0$ . The other cases follow similarly. Suppose, for a contradiction, that  $\widehat{q}_{P1i} - \widehat{q}_{P1j} > 0$  and  $z_{2i}^*(y, t) < z_{2j}^*(y, t)$ . From the optimality of  $z_{2i}^*(y, t)$  for  $\Pi_{2i}(\cdot, \Delta(y, t))$ , we have

$$\Pi_{2i}(z_{2i}^*(y, t), \Delta(y, t)) \geq \Pi_{2i}(z_{2j}^*(y, t), \Delta(y, t)).$$

Furthermore, from Lemma 10, we have

$$\Pi_{2i}(z_{2i}^*(y, t), \Delta(y, t)) - \Pi_{2j}(z_{2i}^*(y, t), \Delta(y, t)) < \Pi_{2i}(z_{2j}^*(y, t), \Delta(y, t)) - \Pi_{2j}(z_{2j}^*(y, t), \Delta(y, t)).$$

The last two inequalities together yield  $\Pi_{2j}(z_{2i}^*(y, t), \Delta(y, t)) > \Pi_{2j}(z_{2j}^*(y, t), \Delta(y, t))$ , which is a contradiction to the optimality of  $z_{2j}^*(y, t)$  for  $\Pi_{2j}(\cdot, \Delta(y, t))$ . Hence, if  $\widehat{q}_{P1i} - \widehat{q}_{P1j} > 0$ , then we must have  $z_{2i}^*(y, t) \geq z_{2j}^*(y, t)$ . It remains to show that  $z_{2i}^*(y, t) \neq z_{2j}^*(y, t)$ .

Now, suppose for contradiction that  $\widehat{q}_{P1i} - \widehat{q}_{P1j} > 0$  and  $z_{2i}^*(y, t) = z_{2j}^*(y, t)$ . From the optimality of  $z_{2i}^*(y, t)$  for  $\Pi_{2i}(\cdot, \Delta(y, t))$ , we must have  $\left. \frac{d\Pi_{2i}(p, \Delta(y, t))}{dp} \right|_{z_{2i}^*(y, t)} = 0$ . Therefore, from Lemma 10, we have  $\left. \frac{d\Pi_{2j}(p, \Delta(y, t))}{dp} \right|_{z_{2j}^*(y, t)} < 0$ , which is a contradiction to the optimality of  $z_{2j}^*(y, t)$  for  $\Pi_{2j}(\cdot, \Delta(y, t))$ . Hence, if  $\widehat{q}_{P1i} - \widehat{q}_{P1j} > 0$ , then we cannot have  $z_{2i}^*(y, t) = z_{2j}^*(y, t)$ . Therefore, it must be that  $z_{2i}^*(y, t) > z_{2j}^*(y, t)$ .  $\square$

**Lemma 12** *Suppose the regular products are indexed so that  $\widehat{q}_{P11} \leq \dots \leq \widehat{q}_{P1n}$ . Then:*

(a)  $z_{21}^*(y, t) \leq \dots \leq z_{2n}^*(y, t)$ ,

(b)  $z_1^*(y, t) \leq p_t^*(y)$ .

**Proof of Lemma 12:** Observe that, by Lemma 11 and our assumption that  $\widehat{q}_{P11} \leq \dots \leq \widehat{q}_{P1n}$ , we have  $z_{21}^*(y, t) \leq \dots \leq z_{2n}^*(y, t)$ . To see why  $p_t^*(y) \geq z_1^*(y, t)$ , note that  $\max_{0 \leq p_k \leq p} \{\lambda_k \beta_k \Pi_{2k}(p_k, \Delta(y, t))\}$  is increasing in  $p$ . Hence, we notice from the optimality equations in (A-22) that we must have  $p_t^*(y) \geq z_1^*(y, t)$ .  $\square$

**Proof of Theorem 3:**

**Proof of (a):** First,  $z_1^*(y, t) > z_{2k}^*(y, t)$  follows from Lemma 9. This along with  $p_t^*(y) \geq z_1^*(y, t)$  (by Lemma 12(b)) yield  $p_t^*(y) > z_{2k}^*(y, t)$ . Therefore, when  $p = p_t^*(y)$ , setting  $p_k = z_{2k}^*(y, t)$  is feasible for the optimization problem in (A-22), and it is optimal to do so since  $z_{2k}^*(y, t)$  maximizes  $\Pi_{2k}(p)$ . The result follows.

**Proof of (b):** Since there is at least one dissimilar regular product, there exists  $j \in \{1, \dots, n\}$  such that  $z_1^*(y, t) \leq z_{2j}^*(y, t)$  by Lemma 9. Then, it follows from Lemma 12(a) and (b) that there exists  $m \in \{1, \dots, n\}$  such that  $p_t^*(y) > z_{2k}^*(y, t)$  for  $k \leq m$ . Therefore, as in part (a), when  $p = p_t^*(y)$ , setting  $p_k = z_{2k}^*(y, t)$  for  $k \leq m$  is feasible for the optimization problem in (A-22), and it is optimal to do so. The result follows.  $\square$