

# Appendix

## Proof of lemma 2

It is obvious that  $\Pi_b^b$  is a continuous and non-differentiable function and has a non-differential point at  $q_b^b = \lambda_s q_s^b$ , as the left and right derivatives are not equal, i.e.,  $\frac{\partial \Pi_{b1}^b}{\partial q_b^b} \neq \frac{\partial \Pi_{b2}^b}{\partial q_b^b}$ . The optimal production quantity of the user is obtained by comparing the maximums of the two parts of the profit function separated by this point. We first solve the unconstrained solutions to  $\Pi_{b1}^b$  and  $\Pi_{b2}^b$ . It is simple to show that  $\Pi_{b1}^b$  and  $\Pi_{b2}^b$  are both strictly concave in  $q_b^b$ . Solving the first order conditions yields the unconstrained optimal solutions to  $\Pi_{b1}^b$  and  $\Pi_{b2}^b$ , i.e.,  $q_{b1}^b = \frac{\alpha_b - c_b - r_b - \lambda_b d_b}{2\beta_b}$  and  $q_{b2}^b = \frac{\alpha_b - c_b - p_w - \lambda_b d_b}{2\beta_b}$ , as  $r_b \geq p_w$ ,  $q_{b1}^b \leq q_{b2}^b$ .

The constrained optimal solutions to  $\Pi_b^b$  are determined by comparing the unconstrained ones with the separation point  $\lambda_s q_s^b$ , which are presented as follows.

- (i) If  $q_{b1}^b > \lambda_s q_s^b$ , then  $q_{b1}^b$  satisfies the constraint and is the constrained optimal solution to  $\Pi_{b1}^b$ , while  $q_{b2}^b$  falls to the right of  $\lambda_s q_s^b$  and does not subject to the constraint as  $q_{b1}^b \leq q_{b2}^b$ .  $\Pi_{b2}^b$  increases as  $q_b^b$  approaches  $\lambda_s q_s^b$  from the left, thus obtaining the maximum value at  $q_b^b = \lambda_s q_s^b$ . Since  $\Pi_{b2}^b(\lambda_s q_s^b) = \Pi_{b1}^b(\lambda_s q_s^b) < \Pi_{b1}^b(q_{b1}^b)$  by the definition of optimality and continuity, the optimal production quantity that maximizes the user's profit is  $q_b^{b*}(q_s^b, p_w, \theta) = q_{b1}^b$ .
- (ii) If  $q_{b2}^b < \lambda_s q_s^b$ , then  $q_{b2}^b$  subjects to the constraint and is the constrained optimal solution to  $\Pi_{b2}^b$ , while  $q_{b1}^b$  falls to the left of  $\lambda_s q_s^b$  and is not the constrained optimal solution.  $\Pi_{b1}^b$  increases as  $q_b^b$  approaches  $\lambda_s q_s^b$  from the right, thus obtaining the maximum value at  $q_b^b = \lambda_s q_s^b$ . Since  $\Pi_{b1}^b(\lambda_s q_s^b) = \Pi_{b2}^b(\lambda_s q_s^b) < \Pi_{b2}^b(q_{b2}^b)$  by the definition of optimality and continuity, the the user get the optimal profit at  $q_b^{b*}(q_s^b, p_w, \theta) = q_{b2}^b$ .
- (iii) If  $q_{b1}^b \leq \lambda_s q_s^b \leq q_{b2}^b$ , then neither the unconstrained optimal solutions satisfy the constraints.  $\Pi_{b1}^b$  increases as  $q_b^b$  approaches  $\lambda_s q_s^b$  from the right, and  $\Pi_{b2}^b$  increases as  $q_b^b$  approaches  $\lambda_s q_s^b$  from the left, thus, the constrained optimal solutions to  $\Pi_{b1}^b$  and  $\Pi_{b2}^b$  both fall at the separated point. Therefore the user maximizes his profit at  $q_b^{b*}(q_s^b, p_w, \theta) = \lambda_s q_s^b$ .

## Proof of Lemma 3

$\Pi_s^b$  is separated by  $q_s^b = \frac{q_{b2}^b}{\lambda_s}$  into two parts, and is continuous but non-differentiable at  $q_s^b = \frac{q_{b2}^b}{\lambda_s}$ , as the left and right derivatives are not equal. It is simple to show that  $\Pi_{s1}^b$  and  $\Pi_{s2}^b$  are both strictly concave in  $q_s^b$ , solving the first order conditions give the unconstrained optimal solutions to  $\Pi_{s1}^b$  and  $\Pi_{s2}^b$ , i.e.,  $q_{s1}^b = \frac{\alpha_s - c_s - r_s + \lambda_s p_w}{2\beta_s}$  and  $q_{s2}^b = \frac{\alpha_s - c_s - r_s - \lambda_s d_s}{2\beta_s}$ , respectively, since  $p_w \geq -d_s$ ,  $q_{s1}^b \geq q_{s2}^b$ .

The constrained optimal solutions to  $\Pi_s^b$  are determined by comparing the unconstrained ones with the separation point  $\frac{q_{b2}^b}{\lambda_s}$ . Considering  $q_{s1}^b \geq q_{s2}^b$  and  $q_{b1}^b \leq q_{b2}^b$ , there are six scenarios to be analyzed.

- (i) If  $q_{s2}^b \leq q_{s1}^b < \frac{q_{b1}^b}{\lambda_s} \leq \frac{q_{b2}^b}{\lambda_s}$ , then  $q_{s1}^b$  satisfies the constraint and is the constrained optimal solution to  $\Pi_{s1}^b$ , while  $q_{s2}^b$  falls to the left of  $\frac{q_{b2}^b}{\lambda_s}$  and is not the constrained optimal solution.  $\Pi_{s2}^b$  increases as  $q_s^b$  approaches  $\frac{q_{b2}^b}{\lambda_s}$  from the right and reaches the maximum value at  $\frac{q_{b2}^b}{\lambda_s}$ . Since  $\Pi_s^b$  is a continuous function,  $\Pi_{s1}^b(\frac{q_{b2}^b}{\lambda_s}) = \Pi_{s2}^b(\frac{q_{b2}^b}{\lambda_s})$ , and according to the definition of optimality,  $\Pi_{s1}^b(q_{s1}^b) > \Pi_{s1}^b(\frac{q_{b2}^b}{\lambda_s})$ , so the optimal production quantity that maximizes the generator's profit is  $q_s^{b*}(p_w, \theta) = q_{s1}^b$ . In response, the optimal production quantity of the user is  $q_b^{b*}(p_w, \theta) = q_{b1}^b$ .
- (ii) If  $q_{s2}^b < \frac{q_{b1}^b}{\lambda_s} \leq q_{s1}^b \leq \frac{q_{b2}^b}{\lambda_s}$ , then  $q_{s1}^b$  satisfies the constraint and is the constrained optimal solution to  $\Pi_{s1}^b$ , while  $q_{s2}^b$  falls out of the constrained region and increases as  $q_s^b$  approaches  $\frac{q_{b2}^b}{\lambda_s}$  from the right, thus obtaining the maximum value at  $\frac{q_{b2}^b}{\lambda_s}$ . Since  $\Pi_{s1}^b(\frac{q_{b2}^b}{\lambda_s}) = \Pi_{s2}^b(\frac{q_{b2}^b}{\lambda_s})$  and  $\Pi_{s1}^b(q_{s1}^b) \geq \Pi_{s1}^b(\frac{q_{b2}^b}{\lambda_s})$  by the definition of continuity and optimality, the generator maximizes the profit  $\Pi_s^b$  at  $q_s^{b*}(p_w, \theta) = q_{s1}^b$ . In response, the optimal production quantity of the generator is  $q_s^{b*}(p_w, \theta) = \lambda_s q_{s1}^b$ .

- (iii) If  $\frac{q_{b1}^b}{\lambda_s} < q_{s2}^b \leq q_{s1}^b < \frac{q_{b2}^b}{\lambda_s}$ , then  $q_{s1}^b$  is the constrained optimal solution to  $\Pi_{s1}^b$ , while  $q_{s2}^b$  does not subject to the constraint. Obviously, this scenario is the same as (ii), as a result, the generator maximizes the profit  $\Pi_s^b$  at  $q_s^{b*}(p_w, \theta) = q_{s1}^b$ , and the user achieves the maximum profit at  $q_b^{b*}(p_w, \theta) = \lambda_s q_{s1}^b$ .
- (iv) If  $q_{s2}^b < \frac{q_{b1}^b}{\lambda_s} \leq \frac{q_{b2}^b}{\lambda_s} < q_{s1}^b$ , then neither the unconstrained optimal solutions to  $\Pi_{s1}^b$  and  $\Pi_{s2}^b$  satisfy the constraints. Therefore,  $\Pi_{s1}^b$  increases as  $q_s^b$  approaches  $\frac{q_{b2}^b}{\lambda_s}$  from left and  $\Pi_{s2}^b$  increases as  $q_s^b$  approaches  $\frac{q_{b2}^b}{\lambda_s}$  from right, and both reach the maximum values at  $\frac{q_{b2}^b}{\lambda_s}$ . Since  $\Pi_s^b$  is a continuous function,  $\Pi_{s1}^b(\frac{q_{b2}^b}{\lambda_s}) = \Pi_{s2}^b(\frac{q_{b2}^b}{\lambda_s})$ . Hence, the generator gets the optimal profit at  $q_s^{b*}(p_w, \theta) = \frac{q_{b2}^b}{\lambda_s}$ . In response, the user achieves the optimal profit at  $q_b^{b*}(p_w, \theta) = q_{b2}^b$ .
- (v) If  $\frac{q_{b1}^b}{\lambda_s} \leq q_{s2}^b \leq \frac{q_{b2}^b}{\lambda_s} < q_{s1}^b$ , then neither the unconstrained optimal solutions to  $\Pi_{s1}^b$  and  $\Pi_{s2}^b$  satisfy the constraints. The scenario is same as (iv), consequently, the optimal production quantity of the generator is  $q_s^{b*}(p_w, \theta) = \frac{q_{b2}^b}{\lambda_s}$ , and the optimal production quantity of the user is  $q_b^{b*}(p_w, \theta) = q_{b2}^b$ .
- (vi) If  $\frac{q_{b1}^b}{\lambda_s} \leq \frac{q_{b2}^b}{\lambda_s} < q_{s2}^b \leq q_{s1}^b$ , then  $q_{s2}^b$  satisfies the constraint and is the constrained optimal solution to  $\Pi_{s2}^b$ , while  $q_{s1}^b$  falls out of the constrained region and reaches the maximum value at  $\frac{q_{b2}^b}{\lambda_s}$ . Since  $\Pi_{s1}^b(\frac{q_{b2}^b}{\lambda_s}) = \Pi_{s2}^b(\frac{q_{b2}^b}{\lambda_s})$  and  $\Pi_{s2}^b(q_{s2}^b) > \Pi_{s2}^b(\frac{q_{b2}^b}{\lambda_s})$  by the definition of continuity and optimality, the generator maximizes the profit at  $q_s^{b*}(p_w, \theta) = q_{s2}^b$ . In response, the user reaches the maximum profit at  $q_b^{b*}(p_w, \theta) = q_{b2}^b$ .

From the above analysis, we can see there are four pairs solutions to  $\Pi_s^b$  and  $\Pi_b^b$ , i.e., (i) When  $0 < q_{s1}^b < \frac{q_{b1}^b}{\lambda_s}$ , the optimal production quantities of the generator and user are  $(q_{s1}^b, q_{b1}^b)$ ; (ii) When  $\frac{q_{b1}^b}{\lambda_s} \leq q_{s1}^b \leq \frac{q_{b2}^b}{\lambda_s}$ , the optimal production quantities of the generator and user are  $(q_{s1}^b, \lambda_s q_{s1}^b)$ ; (iii) When  $q_{s2}^b \leq \frac{q_{b2}^b}{\lambda_s} < q_{s1}^b$ , the optimal production quantities of the generator and user are  $(\frac{q_{b2}^b}{\lambda_s}, q_{b2}^b)$ ; (iv) When  $q_{s2}^b > \frac{q_{b2}^b}{\lambda_s}$ , the optimal production quantities of the generator and user are  $(q_{s2}^b, q_{b2}^b)$ . The four pairs solutions are characterized by  $p_w < p_{w1}$ ,  $p_{w1} \leq p_w \leq p_{w2}$ ,  $p_{w2} < p_w \leq p_{w3}$ , and  $p_w > p_{w3}$ , where  $p_{w1}$ ,  $p_{w2}$ , and  $p_{w3}$  are values of  $p_w$  such that  $q_{s1}^b = \frac{q_{b1}^b}{\lambda_s}$ ,  $q_{s1}^b = \frac{q_{b2}^b}{\lambda_s}$ , and  $q_{s2}^b = \frac{q_{b2}^b}{\lambda_s}$ , respectively. Specifically,  $p_{w1} = \frac{\beta_s(\alpha_b - c_b - r_b - \lambda_b d_b) - \beta_b \lambda_s(\alpha_s - c_s - r_s)}{\beta_b \lambda_s^2}$ ,  $p_{w2} = \frac{\beta_s(\alpha_b - c_b - \lambda_b d_b) - \beta_b \lambda_s(\alpha_s - c_s - r_s)}{\beta_s + \beta_b \lambda_s^2}$ , and  $p_{w3} = \alpha_b - c_b - \lambda_b d_b - \frac{\beta_b \lambda_s(\alpha_s - c_s - r_s - \lambda_s d_s)}{\beta_s}$ , respectively.

Since  $p_w$  subjects to  $[-d_s, r_b]$ , separately comparing  $p_{w1}$ ,  $p_{w2}$ , and  $p_{w3}$  with the constraints, we find  $p_{w1} \geq -d_s$  (i.e.,  $\beta_s(\alpha_b - c_b - r_b - \lambda_b d_b) \geq \beta_b \lambda_s(\alpha_s - c_s - r_s - \lambda_s d_s)$ ) and  $p_{w3} \leq r_b$  (i.e.,  $\beta_s(\alpha_b - c_b - r_b - \lambda_b d_b) \leq \beta_b \lambda_s(\alpha_s - c_s - r_s - \lambda_s d_s)$ ) are contradictory. Therefore, if  $p_{w1} \in [-d_s, r_b]$ , i.e.,  $\beta_b \lambda_s(\alpha_s - c_s - r_s + \lambda_s r_b) \geq \beta_s(\alpha_b - c_b - r_b - \lambda_b d_b) \geq \beta_b \lambda_s(\alpha_s - c_s - r_s - \lambda_s d_s)$ , then  $p_{w3} > r_b \geq p_{w2} > p_{w1} \geq -d_s$  hold; if  $p_{w3} \in [-d_s, r_b]$ , i.e.,  $\beta_s(\alpha_b - c_b - r_b - \lambda_b d_b) \leq \beta_b \lambda_s(\alpha_s - c_s - r_s - \lambda_s d_s) \leq \beta_s(\alpha_b - c_b + d_s - \lambda_b d_b)$ , then  $r_b \geq p_{w3} > p_{w2} \geq -d_s > p_{w1}$  hold. Hence, there exist two scenarios with respect to  $p_w$ , i.e.,  $-d_s \leq p_{w1} < p_{w2} \leq r_b < p_{w3}$  and  $p_{w1} < -d_s \leq p_{w2} < p_{w3} \leq r_b$ , and the optimal production decisions of the generator and user are presented according to the two scenarios.

### Proof of corollary 2

- (i) When  $-d_s \leq p_{w1} < p_{w2} \leq r_b < p_{w3}$ , for  $-d_s \leq p_w \leq p_{w2}$ ,  $\Pi_s^{b*}(p_w, \theta) + \theta F - \Pi_s^{N*} = \frac{(\alpha_s - c_s - r_s + \lambda_s p_w)^2}{4\beta_s} - \frac{(\alpha_s - c_s - r_s - \lambda_s d_s)^2}{4\beta_s}$ , since  $p_w \geq -d_s$ ,  $\Pi_s^{b*}(p_w, \theta) + \theta F \geq \Pi_s^{N*}$ . For  $p_{w2} < p_w \leq r_b$ ,  $\Pi_s^{b*}(p_w, \theta) + \theta F - \Pi_s^{N*} = \frac{(\alpha_b - c_b - p_w - \lambda_b d_b)[2\beta_b \lambda_s(\alpha_s - c_s - r_s + \lambda_s p_w) - \beta_s(\alpha_b - c_b - p_w - \lambda_b d_b)]}{4\beta_b^2 \lambda_s^2} - \frac{(\alpha_s - c_s - r_s - \lambda_s d_s)^2}{4\beta_s} > \beta_b \lambda_s(\alpha_s - c_s - r_s - \lambda_s d_s) * \frac{[\beta_s(\alpha_b - c_b - p_w - \lambda_b d_b) - \beta_b \lambda_s(\alpha_s - c_s - r_s - \lambda_s d_s)]}{4\beta_s \beta_b^2 \lambda_s^2} + \frac{(\alpha_b - c_b - p_w - \lambda_b d_b)[\beta_b \lambda_s(\alpha_s - c_s - r_s + \lambda_s p_w) - \beta_s(\alpha_b - c_b - p_w - \lambda_b d_b)]}{4\beta_b^2 \lambda_s^2}$ , since  $q_{s1}^b > \frac{q_{b2}^b}{\lambda_s} > q_{s2}^b$ ,  $\Pi_s^{b*}(p_w, \theta) + \theta F > \Pi_s^{N*}$ . In summarize,  $\Pi_s^{b*}(p_w, \theta) + \theta F - \Pi_s^{N*} \geq 0$  regardless of  $p_w$ . For  $-d_s \leq p_w < p_{w1}$ ,  $\Pi_b^{b*}(p_w, \theta) + (1 - \theta)F - \Pi_b^{N*} = \frac{\lambda_s(r_b - p_w)(\alpha_s - c_s - r_s + \lambda_s p_w)}{2\beta_s}$ , as  $r_b \geq p_w$ ,  $\Pi_b^{b*}(p_w, \theta) + (1 - \theta)F \geq \Pi_b^{N*}$ . For  $p_{w1} \leq p_w \leq p_{w2}$ ,  $\Pi_b^{b*}(p_w, \theta) + (1 - \theta)F - \Pi_b^{N*} = \frac{\lambda_s(\alpha_s - c_s - r_s + \lambda_s p_w)[2\beta_s(\alpha_b - c_b - p_w - \lambda_b d_b) - \beta_b \lambda_s(\alpha_s - c_s - r_s + \lambda_s p_w)]}{4\beta_s^2} - \frac{(\alpha_b - c_b - r_b - \lambda_b d_b)^2}{4\beta_b} > \beta_s(\alpha_b - c_b - r_b - \lambda_b d_b) * \frac{[\beta_b \lambda_s(\alpha_s - c_s - r_s + \lambda_s p_w) - \beta_s(\alpha_b - c_b - r_b - \lambda_b d_b)]}{4\beta_b \beta_s^2} + \frac{\lambda_s(\alpha_s - c_s - r_s + \lambda_s p_w)[\beta_s(\alpha_b - c_b - p_w - \lambda_b d_b) - \beta_b \lambda_s(\alpha_s - c_s - r_s + \lambda_s p_w)]}{4\beta_s^2}$ , as  $\frac{q_{b1}^b}{\lambda_s} \leq$

$q_{s1}^b \leq \frac{q_{b2}^b}{\lambda_s}$ ,  $\Pi_b^{b*}(p_w, \theta) + (1 - \theta)F > \Pi_b^{N*}$ . For  $p_{w2} < p_w \leq r_b$ ,  $\Pi_b^{b*}(p_w, \theta) + (1 - \theta)F - \Pi_b^{N*} = \frac{(\alpha_b - c_b - p_w - \lambda_b d_b)^2}{4\beta_b} - \frac{(\alpha_b - c_b - r_b - \lambda_b d_b)^2}{4\beta_b}$ , since  $r_b \geq p_w$ ,  $\Pi_b^{b*}(p_w, \theta) + (1 - \theta)F \geq \Pi_b^{N*}$ . Therefore,  $\Pi_b^{b*}(p_w, \theta) + \theta F \geq \Pi_b^{N*}$  regardless of  $p_w$ .

(ii) When  $p_{w1} < -d_s \leq p_{w2} < p_{w3} \leq r_b$ , for  $-d_s \leq p_w \leq p_{w2}$ , we can see from above that  $\Pi_s^{b*}(p_w, \theta) + \theta F \geq \Pi_s^{N*}$  and  $\Pi_b^{b*}(p_w, \theta) + (1 - \theta)F > \Pi_b^{N*}$ . For  $p_{w2} < p_w \leq p_{w3}$ , similar to the results described above,  $\Pi_s^{b*}(p_w, \theta) + \theta F > \Pi_s^{N*}$  and  $\Pi_b^{b*}(p_w, \theta) + (1 - \theta)F \geq \Pi_b^{N*}$  hold. For  $p_{w3} < p_w \leq r_b$ ,  $\Pi_s^{b*}(p_w, \theta) + \theta F - \Pi_s^{N*} = \frac{(p_w + d_s)(\alpha_b - c_b - p_w - \lambda_b d_b)}{2\beta_b} > 0$ , so  $\Pi_s^{b*}(p_w, \theta) + \theta F > \Pi_s^{N*}$ .  $\Pi_b^{b*}(p_w, \theta) + (1 - \theta)F - \Pi_b^{N*} = \frac{(\alpha_b - c_b - p_w - \lambda_b d_b)^2}{4\beta_b} - \frac{(\alpha_b - c_b - r_b - \lambda_b d_b)^2}{4\beta_b}$ , since  $r_b \geq p_w$ ,  $\Pi_b^{b*}(p_w, \theta) + (1 - \theta)F \geq \Pi_b^{N*}$ . Therefore,  $\Pi_s^{b*}(p_w, \theta) + \theta F \geq \Pi_s^{N*}$  and  $\Pi_b^{b*}(p_w, \theta) + \theta F \geq \Pi_s^{N*}$  regardless of  $p_w$ .

### Proof of Theorem 2

When  $-d_s \leq p_{w1} < p_{w2} \leq r_b < p_{w3}$ , for  $-d_s \leq p_w \leq p_{w2}$ ,  $\frac{\partial q_s^{b*}(p_w, \theta)}{\partial p_w} = \frac{\lambda_s}{2\beta_s} > 0$ , for  $p_{w2} < p_w \leq r_b$ ,  $\frac{\partial q_s^{b*}(p_w, \theta)}{\partial p_w} = -\frac{1}{2\beta_b \lambda_s} < 0$ . For  $-d_s \leq p_w < p_{w1}$ ,  $\frac{\partial q_b^{b*}(p_w, \theta)}{\partial p_w} = 0$ , for  $p_{w1} \leq p_w \leq p_{w2}$ ,  $\frac{\partial q_b^{b*}(p_w, \theta)}{\partial p_w} = \frac{\lambda_s^2}{2\beta_s} > 0$ , and for  $p_{w2} < p_w \leq r_b$ ,  $\frac{\partial q_b^{b*}(p_w, \theta)}{\partial p_w} = \frac{-1}{2\beta_b} < 0$ .

When  $p_{w1} < -d_s \leq p_{w2} < p_{w3} \leq r_b$ , for  $-d_s \leq p_w \leq p_{w2}$ ,  $\frac{\partial q_s^{b*}(p_w, \theta)}{\partial p_w} = \frac{\lambda_s}{2\beta_s} > 0$ , for  $p_{w2} < p_w \leq p_{w3}$ ,  $\frac{\partial q_s^{b*}(p_w, \theta)}{\partial p_w} = -\frac{1}{2\beta_b \lambda_s} < 0$ , for  $p_{w3} < p_w \leq r_b$ ,  $\frac{\partial q_s^{b*}(p_w, \theta)}{\partial p_w} = 0$ . For  $-d_s \leq p_w \leq p_{w2}$ ,  $\frac{\partial q_b^{b*}(p_w, \theta)}{\partial p_w} = \frac{\lambda_s^2}{2\beta_s} > 0$ , for  $p_{w2} < p_w \leq r_b$ ,  $\frac{\partial q_b^{b*}(p_w, \theta)}{\partial p_w} = \frac{-1}{2\beta_b} < 0$ .

### Proof of Theorem 3

When  $-d_s \leq p_{w1} < p_{w2} \leq r_b < p_{w3}$ ,  $\Delta \Pi_T^b$  is a continuous function of  $p_w$ , since values of  $\Delta \Pi_T^b$  at the separated points  $p_w = p_{w1}$  and  $p_w = p_{w2}$  are equal, respectively. For  $-d_s \leq p_w < p_{w1}$ ,  $\frac{\partial \Delta \Pi_T^b}{\partial p_w} = \frac{\lambda_s^2(r_b - p_w)}{2\beta_s} > 0$ ,  $\frac{\partial^2 \Delta \Pi_T^b}{\partial p_w^2} = \frac{-\lambda_s^2}{2\beta_s} < 0$ ,  $\frac{\partial \Delta \Pi_T^b}{\partial r_b} = \frac{\lambda_s(\alpha_s - c_s - r_s + \lambda_s p_w)}{2\beta_s} > 0$ , and  $\frac{\partial \Delta \Pi_T^b}{\partial d_s} = \frac{\lambda_s(\alpha_s - c_s - r_s - \lambda_s d_s)}{2\beta_s} > 0$ , so  $\Delta \Pi_T^b$  is an increasingly concave function of  $p_w$  and an increasing function of  $r_b$  and  $d_s$ , respectively. For  $p_{w1} \leq p_w \leq p_{w2}$ ,  $\frac{\partial \Delta \Pi_T^b}{\partial p_w} = \frac{\lambda_s^2(\alpha_b - c_b - p_w - \lambda_b d_b)}{2\beta_s} - \frac{\beta_b \lambda_s^3(\alpha_s - c_s - r_s + \lambda_s p_w)}{2\beta_s^2} > 0$ ,  $\frac{\partial^2 \Delta \Pi_T^b}{\partial p_w^2} = \frac{-\lambda_s^2(\beta_s + \beta_b \lambda_s^2)}{2\beta_s} < 0$ ,  $\frac{\partial \Delta \Pi_T^b}{\partial r_b} = \frac{\alpha_b - c_b - r_b - \lambda_b d_b}{2\beta_b} > 0$ , and  $\frac{\partial \Delta \Pi_T^b}{\partial d_s} = \frac{\lambda_s(\alpha_s - c_s - r_s - \lambda_s d_s)}{2\beta_s} > 0$ , so  $\Delta \Pi_T^b$  is an increasingly concave function of  $p_w$  and an increasing function of  $r_b$  and  $d_s$ , respectively. For  $p_{w2} < p_w \leq r_b$ ,  $\frac{\partial \Delta \Pi_T^b}{\partial p_w} = \frac{\beta_s(\alpha_b - c_b - p_w - \lambda_b d_b) - \lambda_s \beta_b(\alpha_s - c_s - r_s + \lambda_s p_w)}{2\lambda_s^2 \beta_b^2} < 0$ ,  $\frac{\partial^2 \Delta \Pi_T^b}{\partial p_w^2} = \frac{-(\beta_s + \beta_b \lambda_s^2)}{2\lambda_s^2 \beta_b^2} < 0$ ,  $\frac{\partial \Delta \Pi_T^b}{\partial r_b} = \frac{\alpha_b - c_b - r_b - \lambda_b d_b}{2\beta_b} > 0$ , and  $\frac{\partial \Delta \Pi_T^b}{\partial d_s} = \frac{\lambda_s(\alpha_s - c_s - r_s - \lambda_s d_s)}{2\beta_s} > 0$ , so  $\Delta \Pi_T^b$  is a decreasingly concave function of  $p_w$  and an increasing function of  $r_b$  and  $d_s$ , respectively.

When  $p_{w1} < -d_s \leq p_{w2} < p_{w3} \leq r_b$ ,  $\Delta \Pi_T^b$  is a continuous function of  $p_w$  since values of  $\Delta \Pi_T^b$  at the separated points  $p_w = p_{w2}$  and at  $p_w = p_{w3}$  are equal, respectively. For  $-d_s \leq p_w \leq p_{w2}$ , from above, we can see that  $\Delta \Pi_T^b$  is an increasingly concave function of  $p_w$ ,  $r_b$  and  $d_s$ , respectively. For  $p_{w2} < p_w \leq p_{w3}$ , from above, we can see that  $\Delta \Pi_T^b$  is a decreasingly concave function of  $p_w$  and an increasingly concave function of  $r_b$  and  $d_s$ , respectively. For  $p_{w3} < p_w \leq r_b$ ,  $\frac{\partial \Delta \Pi_T^b}{\partial p_w} = \frac{-(p_w + d_s)}{2\beta_b} < 0$ ,  $\frac{\partial^2 \Delta \Pi_T^b}{\partial p_w^2} = \frac{-1}{2\beta_b} < 0$ ,  $\frac{\partial \Delta \Pi_T^b}{\partial r_b} = \frac{\alpha_b - c_b - r_b - \lambda_b d_b}{2\beta_b} > 0$ , and  $\frac{\partial \Delta \Pi_T^b}{\partial d_s} = \frac{\alpha_b - c_b - p_w - \lambda_b d_b}{2\beta_b} > 0$ , so  $\Delta \Pi_T^b$  is a decreasingly concave function of  $p_w$  and an increasing function of  $r_b$  and  $d_s$ , respectively.

### Proof of theorem 4

When  $-d_s \leq p_{w1} < p_{w2} \leq r_b < p_{w3}$ , for  $-d_s \leq p_w < p_{w1}$ ,  $\frac{\partial \Delta E_T^b}{\partial p_w} = \frac{-[e_r + e_p(1 - \delta)]\lambda_s^2}{2\beta_s} < 0$ , so  $\Delta E_T^b$  decreases with  $p_w$ , and get the minimum value at  $p_w = p_{w1}$ . For  $p_{w1} \leq p_w \leq p_{w2}$ ,  $\frac{\partial \Delta E_T^b}{\partial p_w} = \frac{\delta e_p \lambda_s^2}{2\beta_s} > 0$ , so  $\Delta E_T^b$  increases with  $p_w$ , and get the minimum value at  $p_w = p_{w1}$ . For  $p_{w2} < p_w \leq r_b$ ,  $\frac{\partial \Delta E_T^b}{\partial p_w} = \frac{-\delta e_p}{2\beta_b} < 0$ , so  $\Delta E_T^b$  decreases with  $p_w$ , and get the minimum value when  $p_w = r_b$ .

When  $p_{w1} < -d_s \leq p_{w2} < p_{w3} \leq r_b$ , we can see from above that for  $-d_s \leq p_w \leq p_{w2}$ ,  $\Delta E_T^b$  increases with  $p_w$ , for  $p_{w2} < p_w \leq p_{w3}$ ,

$\Delta E_T^b$  decreases with  $p_w$ . For  $p_{w3} < p_w \leq r_b$ ,  $\frac{\partial \Delta E_T^b}{\partial p_w} = \frac{e_d - \delta e_p}{2\beta_b}$ , if  $e_d - \delta e_p < 0$ , then  $\Delta E_T^b$  decreases with  $p_w$ , and get the minimum value when  $p_w = r_b$ , otherwise,  $\Delta E_T^b$  increases with  $p_w$ , and get the minimum value when  $p_w = p_{w3}$ .

#### Proof of theorem 5

When  $-d_s \leq p_{w1} < p_{w2} \leq r_b \leq p_{w3}$ , for  $-d_s \leq p_w < p_{w1}$ ,  $\frac{\partial \Delta E_T^b}{\partial d_s} = \frac{e_d \lambda_s^2}{2\beta_s} > 0$ ,  $\frac{\partial \Delta E_T^b}{\partial r_b} = 0$ , for  $p_{w1} \leq p_w \leq p_{w2}$ ,  $\frac{\partial \Delta E_T^b}{\partial d_s} = \frac{e_d \lambda_s^2}{2\beta_s} > 0$ ,  $\frac{\partial \Delta E_T^b}{\partial r_b} = \frac{e_r + e_p}{2\beta_b} > 0$ , for  $p_{w2} < p_w \leq r_b$ ,  $\frac{\partial \Delta E_T^b}{\partial d_s} = \frac{e_d \lambda_s^2}{2\beta_s} > 0$ ,  $\frac{\partial \Delta E_T^b}{\partial r_b} = \frac{e_r + e_p}{2\beta_b} > 0$ .

When  $p_{w1} < -d_s \leq p_{w2} < p_{w3} \leq r_b$ , for  $-d_s \leq p_w \leq p_{w2}$ ,  $\frac{\partial \Delta E_T^b}{\partial d_s} = \frac{e_d \lambda_s^2}{2\beta_s} > 0$ ,  $\frac{\partial \Delta E_T^b}{\partial r_b} = \frac{e_r + e_p}{2\beta_b} > 0$ , for  $p_{w2} < p_w \leq p_{w3}$ ,  $\frac{\partial \Delta E_T^b}{\partial d_s} = \frac{e_d \lambda_s^2}{2\beta_s} > 0$ ,  $\frac{\partial \Delta E_T^b}{\partial r_b} = \frac{e_r + e_p}{2\beta_b} > 0$ , for  $p_{w3} < p_w \leq r_b$ ,  $\frac{\partial \Delta E_T^b}{\partial d_s} = 0$ ,  $\frac{\partial \Delta E_T^b}{\partial r_b} = \frac{e_r + e_p}{2\beta_b} > 0$ .

#### Proof of theorem 6

When  $-d_s \leq p_{w1} < p_{w2} \leq r_b < p_{w3}$ , from theorem 3, we see that  $\Delta \Pi_T^b$  is an increasing concave function for  $p_w \in [-d_s, p_{w2}]$ , and a decreasing concave function for  $p_w \in (p_{w2}, r_b]$ . Therefore,  $\Delta \Pi_T^b$  achieves the maximum value at  $p_w = p_{w2}$ , and the minimum value at either  $p_w = -d_s$  or  $p_w = r_b$ . Since  $\Delta \Pi_T^b(r_b) = \Delta \Pi_T^b(p_{w1}) > \Delta \Pi_T^b(-d_s)$ , so  $\Delta \Pi_T^b$  obtains the minimum value at  $p_w = -d_s$ .

When  $p_{w1} < -d_s \leq p_{w2} < p_{w3} \leq r_b$ ,  $\Delta \Pi_T^b$  is an increasing concave function for  $p_w \in [-d_s, p_{w2}]$ , and a decreasing concave function for  $p_w \in (p_{w2}, r_b]$ . Therefore,  $\Delta \Pi_T^b$  achieves the maximum value at  $p_w = p_{w2}$ , and the minimum value at either  $p_w = -d_s$  or  $p_w = r_b$ . Since  $\Delta \Pi_T^b(-d_s) = \Delta \Pi_T^b(p_{w3}) > \Delta \Pi_T^b(r_b)$ , so  $\Delta \Pi_T^b$  obtains the minimum value at  $p_w = r_b$ .

#### Proof of Lemma 4

Considering the complexity of the Nash product in our model, we calculate the waste trading price  $p_w$  and share of the fixed investment cost  $\theta$  by maximizing the logarithm of  $G(p_w, \theta)$ . Let  $\ln G(p_w, \theta) = \ln G_1$  when  $-d_s \leq p_w < p_{w1}$ ,  $\ln G(p_w, \theta) = \ln G_2$  when  $p_{w1} \leq p_w \leq p_{w2}$  or  $-d_s \leq p_w \leq p_{w2}$ ,  $\ln G(p_w, \theta) = \ln G_3$  when  $p_{w2} < p_w \leq r_b$  or  $p_{w2} < p_w \leq p_{w3}$ , and  $\ln G(p_w, \theta) = \ln G_4$  when  $p_{w3} < p_w \leq r_b$ .

To ensure that  $\Pi_s^{b*}(p_w, \theta) - \Pi_s^N > 0$  and  $\Pi_b^{b*}(p_w, \theta) - \Pi_b^N > 0$ , we assume  $(\alpha_s - c_s - r_s + \lambda_s p_w)^2 - (\alpha_s - c_s - r_s - \lambda_s d_s)^2 + 2\lambda_s(r_b - p_w)(\alpha_s - c_s - r_s + \lambda_s p_w) - 4\beta_s F > 0$ . Taking the first-order and second-order partial derivatives of  $\ln G_1$  with regard to  $p_w$  and  $\theta$ , we have

$$\begin{aligned} \frac{\partial \ln G_1}{\partial p_w} &= \frac{2\xi_s \lambda_s (\alpha_s - c_s - r_s + \lambda_s p_w)}{(\alpha_s - c_s - r_s + \lambda_s p_w)^2 - (\alpha_s - c_s - r_s - \lambda_s d_s)^2 - 4\beta_s \theta F} \\ &\quad + \frac{\xi_b [\lambda_s^2 (r_b - p_w) - \lambda_s (\alpha_s - c_s - r_s + \lambda_s p_w)]}{\lambda_s (r_b - p_w) (\alpha_s - c_s - r_s + \lambda_s p_w) - 2\beta_s (1 - \theta) F}, \\ \frac{\partial^2 \ln G_1}{\partial p_w^2} &= \frac{-2\xi_s \lambda_s^2 [(\alpha_s - c_s - r_s + \lambda_s p_w)^2 + (\alpha_s - c_s - r_s - \lambda_s d_s)^2 + 4\beta_s \theta F]}{[(\alpha_s - c_s - r_s + \lambda_s p_w)^2 - (\alpha_s - c_s - r_s - \lambda_s d_s)^2 - 4\beta_s \theta F]^2} \\ &\quad - \frac{2\xi_b \lambda_s^2}{\lambda_s (r_b - p_w) (\alpha_s - c_s - r_s + \lambda_s p_w) - 2\beta_s (1 - \theta) F} \\ &\quad - \frac{\xi_b \lambda_s^2 [\lambda_s (r_b - p_w) - (\alpha_s - c_s - r_s + \lambda_s p_w)]^2}{[\lambda_s (r_b - p_w) (\alpha_s - c_s - r_s + \lambda_s p_w) - 2\beta_s (1 - \theta) F]^2}, \\ \frac{\partial \ln G_1}{\partial \theta} &= \frac{-4\xi_s \beta_s F}{(\alpha_s - c_s - r_s + \lambda_s p_w)^2 - (\alpha_s - c_s - r_s - \lambda_s d_s)^2 - 4\beta_s \theta F} \\ &\quad + \frac{2\xi_b \beta_s F}{\lambda_s (r_b - p_w) (\alpha_s - c_s - r_s + \lambda_s p_w) - 2\beta_s (1 - \theta) F}, \end{aligned}$$

$$\frac{\partial^2 \ln G_1}{\partial \theta^2} = \frac{-16\xi_s \beta_s^2 F^2}{[(\alpha_s - c_s - r_s + \lambda_s p_w)^2 - (\alpha_s - c_s - r_s - \lambda_s d_s)^2 - 4\beta_s \theta F]^2} - \frac{4\xi_b \beta_s^2 F^2}{[\lambda_s(r_b - p_w)(\alpha_s - c_s - r_s + \lambda_s p_w) - 2\beta_s(1 - \theta)F]^2},$$

$$\frac{\partial^2 \ln G_1}{\partial \theta \partial p_w} = \frac{8\xi_s \lambda_s \beta_s F(\alpha_s - c_s - r_s + \lambda_s p_w)}{[(\alpha_s - c_s - r_s + \lambda_s p_w)^2 - (\alpha_s - c_s - r_s - \lambda_s d_s)^2 - 4\beta_s \theta F]^2} - \frac{2\xi_b \lambda_s \beta_s F[\lambda_s(r_b - p_w) - (\alpha_s - c_s - r_s + \lambda_s p_w)]}{[\lambda_s(r_b - p_w)(\alpha_s - c_s - r_s + \lambda_s p_w) - 2\beta_s(1 - \theta)F]^2}.$$

Let  $A = \alpha_s - c_s - r_s + \lambda_s p_w$ ,  $B = \alpha_s - c_s - r_s - \lambda_s d_s$ ,  $C = \alpha_b - c_b - p_w - \lambda_b d_b$ ,  $D = \alpha_b - c_b - r_b - \lambda_b d_b$ .

It is obvious  $\frac{\partial^2 \ln G_1}{\partial p_w^2} < 0$  and  $\frac{\partial^2 \ln G_1}{\partial \theta^2} < 0$ , then  $\ln G_1$  is a strictly concave function both in  $p_w$  for a given  $\theta$ , and in  $\theta$  for a given  $p_w$ . Therefore, we can first calculate the optimal  $\theta$  for a given  $p_w$  and then substitute  $\theta$ , which is a function of  $p_w$ , into  $\ln G_1$  to search the optimal  $p_w$ . Solving  $\frac{\partial \ln G_1}{\partial \theta} = 0$ , we can get  $\theta_1(p_w) = \frac{\xi_b(A^2 - B^2) - 2\xi_s[\lambda_s(r_b - p_w)A - 2\beta_s F]}{4\beta_s F(\xi_s + \xi_b)}$ .

Substituting  $\theta_1(p_w)$  into  $\frac{d \ln G_1}{dp_w}$ , we can get  $\frac{d \ln G_1}{dp_w} = \frac{2\lambda_s^2(\xi_s + \xi_b)(r_b - p_w)}{A^2 - B^2 + 2[\lambda_s(r_b - p_w)A - 2\beta_s F]}$ , then  $\frac{d \ln G_1}{dp_w} > 0$  is always true according to the assumption. Therefore,  $\ln G_1$ , constrained by  $-d_s \leq p_w < p_{w1}$ , is an increasing function in  $p_w$ , and reaches the maximum value at  $p_w^* = p_{w1}$ , correspondingly, the optimal share of cost is  $\theta^* = \theta_1(p_{w1})$ .

Taking the first-order and second-order partial derivatives of  $\ln G_2$  with regard to  $p_w$  and  $\theta$ , we can see that  $\ln G_2$  is a strictly concave function both in  $p_w$  for a given  $\theta$ , and in  $\theta$  for a given  $p_w$ , because  $\frac{\partial^2 \ln G_2}{\partial p_w^2} < 0$  and  $\frac{\partial^2 \ln G_2}{\partial \theta^2} < 0$ . Therefore, just like above, we can first calculate the optimal  $\theta$  for a given  $p_w$  and then substitute  $\theta$ , which is a function of  $p_w$ , into  $\ln G_2$  to search the optimal  $p_w$ . Solving  $\frac{\partial \ln G_2}{\partial \theta} = 0$ , we can get  $\theta_2(p_w) = \frac{\xi_b \beta_s \beta_b(A^2 - B^2) - \xi_s[2\lambda_s \beta_s \beta_b AC - \lambda_s^2 \beta_b^2 A^2 - \beta_s^2 D^2 - 4\beta_b \beta_s^2 F]}{4\beta_s \beta_s^2 F(\xi_s + \xi_b)}$ .

Substituting  $\theta_2(p_w)$  into  $\frac{d \ln G_2}{dp_w}$ , we can get  $\frac{d \ln G_2}{dp_w} = \frac{2\beta_b(\xi_s + \xi_b)\lambda_s^2(\beta_s C - \lambda_s \beta_b A)}{\beta_s \beta_b(A^2 - B^2) + 2\lambda_s \beta_s \beta_b AC - \lambda_s^2 \beta_b^2 A^2 - \beta_s^2 D^2 - 4\beta_b \beta_s^2 F}$ , then  $\frac{d \ln G_2}{dp_w} > 0$  is constantly true, since  $\Pi_s^{b*}(p_w, \theta) > \Pi_s^{N*}$  and  $\Pi_b^{b*}(p_w, \theta) > \Pi_b^{N*}$  hold, and  $\beta_s C > \lambda_s \beta_b A$  when  $p_{w1} \leq p_w \leq p_{w2}$ . Therefore,  $\ln G_2$  increases with  $p_w$ , and reaches the maximum value at  $p_w^* = p_{w2}$ , at which  $\theta^* = \theta_2(p_{w2})$ .

Taking the first-order and second-order partial derivatives of  $\ln G_3$  with regard to  $p_w$  and  $\theta$ , we can see that  $\ln G_3$  is a strictly concave function both in  $p_w$  for a given  $\theta$ , and in  $\theta$  for a given  $p_w$ , since  $\frac{\partial^2 \ln G_3}{\partial p_w^2} < 0$  and  $\frac{\partial^2 \ln G_3}{\partial \theta^2} < 0$ . Therefore, we can first calculate the optimal  $\theta$  for a given  $p_w$  and then substitute  $\theta$ , which is a function of  $p_w$ , into  $\ln G_3$  to search the optimal  $p_w$ . Solving  $\frac{\partial \ln G_3}{\partial \theta} = 0$ , we can get  $\theta_3(p_w) = \frac{\xi_b(2\lambda_s \beta_s \beta_b AC - \beta_s^2 C^2 - \lambda_s^2 \beta_b^2 D^2) - \xi_s \beta_s \beta_b \lambda_s^2(C^2 - D^2 - 4\beta_b F)}{4\beta_s \beta_b^2 \lambda_s^2(\xi_s + \xi_b)F}$ .

Substituting  $\theta_3(p_w)$  into  $\frac{d \ln G_3}{dp_w}$ , we can get  $\frac{d \ln G_3}{dp_w} = \frac{2\beta_s(\xi_s + \xi_b)(\beta_s C - \lambda_s \beta_b A)}{2\lambda_s \beta_s \beta_b AC - \beta_s^2 C^2 - \lambda_s^2 \beta_b^2 D^2 + \beta_s \beta_b \lambda_s^2(C^2 - D^2 - 4\beta_b F)}$ , then  $\frac{d \ln G_3}{dp_w} < 0$  is constantly true, as  $\Pi_s^{b*}(p_w, \theta) > \Pi_s^{N*}$  and  $\Pi_b^{b*}(p_w, \theta) > \Pi_b^{N*}$  hold, and  $\beta_s C < \lambda_s \beta_b A$  when  $p_{w2} < p_w \leq r_b$ . Therefore,  $\ln G_3$  decreases with  $p_w$ , and reaches the maximum value at  $p_w^* = p_{w2}$ , correspondingly, the optimal share of cost is  $\theta^* = \theta_3(p_{w2})$ .

Taking the first-order and second-order partial derivatives of  $\ln G_4$  with regard to  $p_w$  and  $\theta$ , we can see that  $\ln G_4$  is a strictly concave function both in  $p_w$  for a given  $\theta$ , and in  $\theta$  for a given  $p_w$ , since  $\frac{\partial^2 \ln G_4}{\partial p_w^2} < 0$  and  $\frac{\partial^2 \ln G_4}{\partial \theta^2} < 0$ . Therefore, we can first calculate the optimal  $\theta$  for a given  $p_w$  and then substitute  $\theta$ , which is a function of  $p_w$ , into  $\ln G_4$  to search the optimal  $p_w$ . Solving  $\frac{\partial \ln G_4}{\partial \theta} = 0$ , we can get  $\theta_4(p_w) = \frac{2\xi_b(p_w + d_s)C - \xi_s(C^2 - D^2 - 4\beta_b F)}{4\beta_b(\xi_s + \xi_b)F}$ .

Substituting  $\theta_4(p_w)$  into  $\frac{d \ln G_4}{dp_w}$ , we can get  $\frac{d \ln G_4}{dp_w} = \frac{-2(\xi_b + \beta_s \xi_s)(p_w + d_s)}{2(p_w + d_s)C + C^2 - D^2 - 4\beta_b F}$ , then  $\frac{d \ln G_4}{dp_w} < 0$  is constantly true according to the assumption. Therefore,  $\ln G_4$ , constrained by  $p_{w3} < p_w \leq r_b$ , is a decreasing function of  $p_w$ , and reaches the maximum value at  $p_w^* = p_{w3}$ , at which  $\theta^* = \theta_4(p_{w3})$ .

### Proof of theorem 7

As  $p_w$  is independent of  $\xi_s$  and  $\xi_b$ , we investigate the effects of  $\xi_s$  and  $\xi_b$  on  $\theta$  by using  $\theta(p_w)$ . When  $-d_s \leq p_{w1} < p_{w2} \leq r_b < p_{w3}$ , for  $-d_s \leq p_w < p_{w1}$ ,  $\frac{\partial \theta_1(p_w)}{\partial \xi_s} = \frac{-\xi_b[A^2 - B^2 + 2\lambda_s(r_b - p_w)A - 4\beta_s F]}{4\beta_s F(\xi_s + \xi_b)^2} < 0$  and  $\frac{\partial \theta_1(p_w)}{\partial \xi_b} = \frac{\xi_s[A^2 - B^2 + 2\lambda_s(r_b - p_w)A - 4\beta_s F]}{4\beta_s F(\xi_s + \xi_b)^2} > 0$  according to  $\Pi_s^{b*}(p_w, \theta) > \Pi_s^N$  and  $\Pi_b^{b*}(p_w, \theta) > \Pi_b^N$ . For  $p_{w1} \leq p_w \leq p_{w2}$ ,  $\frac{\partial \theta_2(p_w)}{\partial \xi_s} = \frac{-\xi_b[\beta_s \beta_b(A^2 - B^2) + 2\beta_s \beta_b \lambda_s AC - \beta_b^2 \lambda_s^2 A^2 - \beta_s^2 D^2 - 4\beta_s^2 \beta_b F]}{4\beta_s^2 \beta_b F(\xi_s + \xi_b)^2} < 0$  and  $\frac{\partial \theta_2(p_w)}{\partial \xi_b} = -\frac{\xi_s}{\xi_b} * \frac{\partial \theta_2(p_w)}{\partial \xi_s} > 0$  according to  $\Pi_s^{b*}(p_w, \theta) > \Pi_s^N$  and  $\Pi_b^{b*}(p_w, \theta) > \Pi_b^N$ . For  $p_{w2} < p_w \leq r_b$ ,  $\frac{\partial \theta_3(p_w)}{\partial \xi_b} = -\frac{\xi_s}{\xi_b} * \frac{\partial \theta_3(p_w)}{\partial \xi_s} > 0$  and  $\frac{\partial \theta_3(p_w)}{\partial \xi_s} = \frac{-\xi_b[\beta_s \beta_b \lambda_s^2(C^2 - D^2 - 4\beta_b F) + 2\beta_s \beta_b \lambda_s AC - \beta_s^2 C^2 - \beta_b^2 \lambda_s^2 D^2]}{4\beta_s \beta_b^2 \lambda_s^2 F(\xi_s + \xi_b)^2} < 0$  according to  $\Pi_s^{b*}(p_w, \theta) > \Pi_s^N$  and  $\Pi_b^{b*}(p_w, \theta) > \Pi_b^N$ .

When  $p_{w1} < -d_s \leq p_{w2} < p_{w3} \leq r_b$ , for  $-d_s \leq p_w \leq p_{w2}$  and  $p_{w2} < p_w \leq p_{w3}$ ,  $\theta_2(p_w)$  and  $\theta_3(p_w)$  change with  $\xi_s$  and  $\xi_b$  in the same direction as in the case  $-d_s \leq p_{w1} < p_{w2} \leq r_b < p_{w3}$ . For  $p_{w3} < p_w \leq r_b$ ,  $\frac{\partial \theta_4(p_w)}{\partial \xi_s} = \frac{-\xi_b[C^2 - D^2 - 4\beta_b F + 2(p_w + d_s)]}{4\beta_b F(\xi_s + \xi_b)^2} < 0$  and  $\frac{\partial \theta_4(p_w)}{\partial \xi_b} = -\frac{\xi_s}{\xi_b} * \frac{\partial \theta_4(p_w)}{\partial \xi_s} > 0$  according to  $\Pi_s^{b*}(p_w, \theta) > \Pi_s^N$  and  $\Pi_b^{b*}(p_w, \theta) > \Pi_b^N$ . Since the optimal profits of the generator and user decrease and increase with  $\theta$ , respectively, it is easy to get the effect of  $\xi_s$  and  $\xi_b$  on  $\Pi_s^{b*}$  and  $\Pi_b^{b*}$ .

### Proof of theorem 8

When  $-d_s \leq p_{w1} < p_{w2} \leq r_b < p_{w3}$ , from theorem 3, 4, we see that  $E_T^{b*}(p_w, \theta)$  decreases and  $\Pi_T^{b*}(p_w, \theta)$  increases with  $p_w$  for  $-d_s \leq p_w < p_{w1}$ , thus the optimal economic and environmental performance are simultaneously achieved at  $p_w = p_{w1}$ , and the two goals align. For  $p_{w1} \leq p_w \leq r_b$ ,  $E_T^{b*}(p_w, \theta)$  and  $\Pi_T^{b*}(p_w, \theta)$  both first increase then decrease with  $p_w$ , hence, the optimal economic performance and worst environmental performance are both achieved at  $p_w = p_{w2}$ , and there exist conflicts between the two goals.

When  $p_{w1} < -d_s \leq p_{w2} < p_{w3} \leq r_b$ , from theorem 3, 4, we see that for  $p_{w1} \leq p_w \leq p_{w3}$ , both  $E_T^{b*}(p_w, \theta)$  and  $\Pi_T^{b*}(p_w, \theta)$  first increase then decrease with  $p_w$ , thus the optimal economic and environmental goals can not be achieved simultaneously. For  $p_{w3} < p_w \leq r_b$ ,  $\Pi_T^{b*}(p_w, \theta)$  decreases with  $p_w$  while  $E_T^{b*}(p_w, \theta)$  increases with  $p_w$  when  $e_d > \delta e_p$ , then the optimal economic and environmental performance align if  $e_d > \delta e_p$ , otherwise, the two goals conflict.

### Proof of theorem 9

When  $-d_s \leq p_{w1} < p_{w2} \leq r_b < p_{w3}$ , for  $-d_s \leq p_w < p_{w1}$ , the optimal waste trading price is  $p_w = p_{w1}$ ,  $E_T^{b*}(p_{w1}) - E_T^{N*} = \frac{(\delta - 1)e_p(\alpha_b - c_b - r_b - \lambda_b d_b)}{2\beta_b} - \frac{\lambda_s e_d(\alpha_s - c_s - r_s - \lambda_s d_s)}{2\beta_s} - \frac{e_r(\alpha_b - c_b - r_b - \lambda_b d_b)}{2\beta_b} < 0$ , as  $0 < \delta < 1$ , therefore, the interfirm waste utilization is more environmentally preferable than the benchmark case. For  $p_{w1} \leq p_w \leq p_{w2}$  and  $p_{w2} < p_w \leq r_b$ , the optimal waste trading price is  $p_w = p_{w2}$ ,  $E_T^{b*}(p_{w2}) - E_T^{N*} = \frac{e_p[\delta \lambda_s[\alpha_s - c_s - r_s + \lambda_s(\alpha_b - c_b - r_b - \lambda_b d_b)] - \beta_s(\beta_s + \beta_b \lambda_s^2)(\alpha_b - c_b - r_b - \lambda_b d_b)]}{2\beta_s \beta_b(\beta_s + \beta_b \lambda_s^2)} - \frac{e_r(\alpha_b - c_b - r_b - \lambda_b d_b)}{2\beta_b} - \frac{\lambda_s e_d(\alpha_s - c_s - r_s - \lambda_s d_s)}{2\beta_s}$ , if  $\delta < \frac{\beta_s(\beta_s + \beta_b \lambda_s^2)(\alpha_b - c_b - r_b - \lambda_b d_b)}{\lambda_s[\alpha_s - c_s - r_s + \lambda_s(\alpha_b - c_b - r_b - \lambda_b d_b)]}$ , then  $E_T^{b*}(p_{w2}) < E_T^{N*}$ , otherwise, the interfirm waste utilization is environmentally superior to the benchmark case if  $e_p < \frac{(\beta_s + \beta_b \lambda_s^2)[\beta_b \lambda_s e_d(\alpha_s - c_s - r_s - \lambda_s d_s) + \beta_s e_r(\alpha_b - c_b - r_b - \lambda_b d_b)]}{\delta \lambda_s[\alpha_s - c_s - r_s + \lambda_s(\alpha_b - c_b - r_b - \lambda_b d_b)] - \beta_s(\beta_s + \beta_b \lambda_s^2)(\alpha_b - c_b - r_b - \lambda_b d_b)} = \tilde{\Omega}(e_d, e_r)$ .

When  $p_{w1} < -d_s \leq p_{w2} < p_{w3} \leq r_b$ , from above we can see that for  $p_{w1} \leq p_w \leq p_{w2}$  and  $p_{w2} < p_w \leq p_{w3}$ , the interfirm waste utilization is more environmentally preferable than the benchmark case if  $\delta < \frac{\beta_s(\beta_s + \beta_b \lambda_s^2)(\alpha_b - c_b - r_b - \lambda_b d_b)}{\lambda_s[\alpha_s - c_s - r_s + \lambda_s(\alpha_b - c_b - r_b - \lambda_b d_b)]}$ , otherwise, it is environmentally superior if  $e_p < \tilde{\Omega}(e_d, e_r)$ . For  $p_{w3} < p_w \leq r_b$ , the optimal waste trading price is  $p_w = p_{w3}$ ,  $E_T^{b*}(p_{w3}) - E_T^{N*} = \frac{e_p[\delta \beta_b \lambda_s(\alpha_s - c_s - r_s - \lambda_s d_s) - \beta_s(\alpha_b - c_b - r_b - \lambda_b d_b)]}{2\beta_s \beta_b} - \frac{\lambda_s e_d(\alpha_s - c_s - r_s - \lambda_s d_s)}{2\beta_s} - \frac{e_r(\alpha_b - c_b - r_b - \lambda_b d_b)}{2\beta_b}$ , if  $\delta < \frac{\beta_s(\alpha_b - c_b - r_b - \lambda_b d_b)}{\beta_b \lambda_s(\alpha_s - c_s - r_s - \lambda_s d_s)}$ , then  $E_T^{b*}(p_{w3}) < E_T^{N*}$ , otherwise, it is environmentally superior to the benchmark case if  $e_p < \frac{\beta_b \lambda_s e_d(\alpha_s - c_s - r_s - \lambda_s d_s) + \beta_s e_r(\alpha_b - c_b - r_b - \lambda_b d_b)}{\delta \beta_b \lambda_s(\alpha_s - c_s - r_s - \lambda_s d_s) - \beta_s(\alpha_b - c_b - r_b - \lambda_b d_b)} = \hat{\Omega}(e_d, e_r)$ .