

# Adversarial Coordination and Public Information Design

## Online Appendix

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### Abstract

This document contains the proof of Examples 2 and 3 in the main text. All numbered items contain the prefix “S.” Any numbered reference without the prefix “S” refers to an item in the main text.

## Proof of Example 2 in Main Text

**Proof of Example 2.** The proof is in two steps. Step 1 characterizes the threshold  $\theta_\sigma^*$  defining the optimal deterministic monotone rule, whereas Step 2 constructs the non-monotone policy that strictly improves over the optimal deterministic monotone one.

**Step 1.** The primitives in this example satisfy the conditions in Theorem 2 in the main text. This means that, given any signal  $s$  disclosed by any policy  $\Gamma$ , MARP is in threshold strategies, which in turn implies that the default outcome is monotone in  $\theta$ .

Next recall that, for any default threshold  $\theta \in [0, 1]$ , the corresponding signal threshold  $x_\sigma^*(\theta)$  is implicitly defined by  $P_\sigma(x_\sigma^*(\theta) | \theta) = \theta$ . Using the fact that, for any  $\theta \in [-K, 1 + K]$  and  $x \in [\theta - \sigma, \theta + \sigma]$ ,  $P_\sigma(x | \theta) = (x - \theta + \sigma) / 2\sigma$ , we have that  $x_\sigma^*(\theta) = (1 + 2\sigma)\theta - \sigma$ .

For any  $\hat{\theta} \in [0, 1]$ , let  $\Gamma^{\hat{\theta}} \equiv \{0, 1\}, \pi^{\hat{\theta}}\}$  be the deterministic monotone policy with cutoff  $\hat{\theta}$ . Next, for any  $\theta \in [\hat{\theta}/(1 + 2\sigma), 1]$ , let  $V_\sigma^{\Gamma^{\hat{\theta}}}(\theta) \equiv U_\sigma^{\Gamma^{\hat{\theta}}}(x_\sigma^*(\theta), 1 | x_\sigma^*(\theta))$  be the expected payoff differential between pledging and not pledging of the marginal investor with signal  $x_\sigma^*(\theta)$ , when each investor pledges if and only if their signal is above  $x_\sigma^*(\theta)$  (and hence default occurs if, and only if, fundamentals are below  $\theta$ ), the quality of the investors' signal is  $\sigma$ , and the policy  $\Gamma^{\hat{\theta}}$  announces that  $s = 1$ , thus revealing that  $\theta \geq \hat{\theta}$ . Note that, for any  $0 \leq \theta < \hat{\theta}/(1 + 2\sigma)$ ,  $x_\sigma^*(\theta) + \sigma < \hat{\theta}$ , which implies that the signal  $x_\sigma^*(\theta)$  is not consistent with the event that fundamentals are above  $\hat{\theta}$ . Equivalently, when  $\theta \geq \hat{\theta}$ , the lowest possible signal that an individual may receive is  $\hat{\theta} - \sigma$ . When each investor pledges if and only if  $x > \hat{\theta} - \sigma$ , default occurs if and only if  $\theta \leq \hat{\theta}/(1 + 2\sigma)$ . Hence, the lowest default threshold that is consistent with the policy  $\Gamma^{\hat{\theta}}$  is  $\hat{\theta}/(1 + 2\sigma)$ . The function  $V_\sigma^{\Gamma^{\hat{\theta}}}(\theta)$  is thus defined only for  $\theta \in [\hat{\theta}/(1 + 2\sigma), 1]$ .

The cutoff  $\theta_\sigma^*$  characterizing the optimal deterministic monotone policy is given by

$$\theta_\sigma^* = \inf\{\hat{\theta} \in [0, 1] : V_\sigma^{\Gamma^{\hat{\theta}}}(\theta) \geq 0 \text{ for all } \theta \in [\hat{\theta}/(1 + 2\sigma), 1]\}. \quad (1)$$

**Claim S2.** For any  $\hat{\theta} \in [0, 1]$ ,  $V_\sigma^{\Gamma^{\hat{\theta}}}(\cdot)$  has a unique minimizer. Letting

$$\theta_\sigma^{\min}(\hat{\theta}) \equiv \arg \min_{\theta \in [\hat{\theta}/(1 + 2\sigma), 1]} V_\sigma^{\Gamma^{\hat{\theta}}}(\theta),$$

we have that  $\theta_\sigma^{\min}(\hat{\theta})$  satisfies  $x_\sigma^*(\theta_\sigma^{\min}(\hat{\theta})) - \sigma = \hat{\theta}$ .

**Proof of Claim S2.** Clearly, for any  $\theta \in [\hat{\theta}/(1 + 2\sigma), \hat{\theta}]$ ,  $V_\sigma^{\Gamma^{\hat{\theta}}}(\theta) = g$ . This is because

when each investor pledges if and only if  $x > x_\sigma^*(\theta)$  default occurs only for fundamentals below  $\theta$ . Hence the announcement that  $\theta > \hat{\theta}$  reveals to the marginal investor with signal  $x_\sigma^*(\theta)$  that default will not occur.

Next, observe that for any  $\theta \in (\hat{\theta}, (\hat{\theta} + 2\sigma)/(1 + 2\sigma)]$ ,  $x_\sigma^*(\theta) - \sigma < \hat{\theta}$ , implying that<sup>1</sup>

$$V_\sigma^{\Gamma^{\hat{\theta}}}(\theta) = g - (g + |b|) \mathbb{P}_\sigma\{\tilde{\theta} \leq \theta | \tilde{\theta} \geq \hat{\theta}; x_\sigma^*(\theta)\} = g - (g + |b|) \frac{\theta - \hat{\theta}}{(1 + 2\sigma)\theta - \hat{\theta}},$$

which is strictly decreasing in  $\theta$ . Finally, note that, for any  $\theta \in ((\hat{\theta} + 2\sigma)/(1 + 2\sigma), 1]$ ,  $x_\sigma^*(\theta) - \sigma > \hat{\theta}$ , implying that

$$V_\sigma^{\Gamma^{\hat{\theta}}}(\theta) = g - (g + |b|) \mathbb{P}_\sigma\{\tilde{\theta} \leq \theta | \tilde{\theta} \geq \hat{\theta}; x_\sigma^*(\theta)\} = g + (g + |b|)(\theta - 1),$$

which is strictly increasing in  $\theta$ . Hence,  $V_\sigma^{\Gamma^{\hat{\theta}}}(\cdot)$  has a single minimizer over  $[\hat{\theta}/(1 + 2\sigma), 1]$ . The latter is equal to  $\theta_\sigma^{\min}(\hat{\theta}) = (\hat{\theta} + 2\sigma)/(1 + 2\sigma)$  and is such that  $x_\sigma^*(\theta_\sigma^{\min}(\hat{\theta})) - \sigma = \hat{\theta}$ .  $\square$

Next, let  $\Gamma^{\theta_\sigma^*} \equiv (\{0, 1\}, \pi^{\theta_\sigma^*})$  be the optimal deterministic monotone policy (with cut-off  $\hat{\theta} = \theta_\sigma^*$ ). Using the characterization of  $\theta_\sigma^*$  in (1), we thus have that, under  $\Gamma^{\theta_\sigma^*}$ , at the point  $\theta_\sigma^{\min}(\theta_\sigma^*)$  at which  $V_\sigma^{\Gamma^{\theta_\sigma^*}}$  reaches its minimum,  $V_\sigma^{\Gamma^{\theta_\sigma^*}}(\theta_\sigma^{\min}(\theta_\sigma^*)) = 0$ . Using the fact that

$$V_\sigma^{\Gamma^{\theta_\sigma^*}}(\theta_\sigma^{\min}(\theta_\sigma^*)) = g - (g + |b|) \frac{\theta_\sigma^{\min}(\theta_\sigma^*) - \theta_\sigma^*}{(1 + 2\sigma)\theta_\sigma^{\min}(\theta_\sigma^*) - \theta_\sigma^*},$$

we then have that  $\theta_\sigma^* = (1 + 2\sigma) \frac{|b|}{g + |b|} - 2\sigma$ . Next, let  $\Gamma_\emptyset$  be the no-disclosure policy and note that, for any  $\theta \in [0, 1]$ ,

$$V_\sigma^{\Gamma_\emptyset}(\theta) = g - (g + |b|) \mathbb{P}_\sigma\{\tilde{\theta} \leq \theta | x_\sigma^*(\theta)\} = g + (g + |b|)(\theta - 1),$$

which is increasing in  $\theta$  and has a unique zero at  $\theta = |b|/(g + |b|) \equiv \theta^{MS}$ .

This means that, in the absence of any disclosure, under the unique rationalizable strategy profile (and hence under MARP), each investor pledges if and only if  $x > x_\sigma^*(\theta^{MS})$ , and default occurs if and only if fundamentals are below  $\theta^{MS}$ . The results above then imply that the optimal deterministic policy  $\Gamma^{\theta_\sigma^*}$  is defined by a threshold  $\theta_\sigma^* = (1 + 2\sigma)\theta^{MS} - 2\sigma = x_\sigma^*(\theta^{MS}) - \sigma$  that coincides with the left end-point of the support of the posterior beliefs of

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<sup>1</sup>The notation  $\mathbb{P}_\sigma\{\tilde{\theta} \leq \theta | \tilde{\theta} \geq \hat{\theta}; x\}$  stands for the probability that an investor with signal  $x$  assigns to the event that  $\tilde{\theta} \leq \theta$  when the quality of his exogenous signal is parametrized by  $\sigma$  and the policy reveals that  $\tilde{\theta} \geq \hat{\theta}$ .

each agent with signal  $x_\sigma^*(\theta^{MS})$ . In fact, for any truncation point  $\hat{\theta} < x_\sigma^*(\theta^{MS}) - \sigma$ , there exists  $\theta$  close to  $\theta^{MS}$  such that  $V_\sigma^{\Gamma^{\hat{\theta}}}(\theta) < 0$  implying that refraining from pledging for all  $x < x_\sigma^*(\theta^{MS})$  is rationalizable in the continuation game following the announcement that  $\theta \geq \hat{\theta}$ , implying that the policy  $\Gamma^{\hat{\theta}}$  fails to satisfy PCP. Similarly, for any truncation point  $\hat{\theta} > x_\sigma^*(\theta^{MS}) - \sigma$ ,  $V_\sigma^{\Gamma^{\hat{\theta}}}(\theta)$  reaches its minimum at  $\theta_\sigma^{\min}(\hat{\theta}) > \theta^{MS}$  and is such that  $V_\sigma^{\Gamma^{\hat{\theta}}}(\theta_\sigma^{\min}(\hat{\theta})) = V_\sigma^{\Gamma^{\theta_0}}(\theta_\sigma^{\min}(\hat{\theta})) > V_\sigma^{\Gamma^{\theta_0}}(\theta^{MS}) = 0$ , where the inequality follows from the monotonicity of  $V_\sigma^{\Gamma^{\theta_0}}(\cdot)$ . Hence,  $\theta_\sigma^* = x_\sigma^*(\theta^{MS}) - \sigma$ .

**Step 2.** Having characterized the optimal deterministic monotone policy  $\Gamma^{\theta_\sigma^*}$ , we now show that, when  $\sigma$  is small, there exists another policy  $\Gamma$  that also satisfies PCP and guarantees no default for a larger set of fundamentals than  $\Gamma^{\theta_\sigma^*}$ .

Let  $\sigma^\# \equiv \frac{\theta^{MS}}{2(1-\theta^{MS})} > 0$ . For any  $\sigma \in (0, \sigma^\#)$ ,  $\theta_\sigma^* = (1+2\sigma)\theta^{MS} - 2\sigma > 0$ . For any  $\sigma, \delta, \gamma > 0$  small, let  $\theta_\sigma''(\delta, \gamma) \equiv x_\sigma^*(\theta^{MS} - \delta) - \sigma = (1+2\sigma)(\theta^{MS} - \delta) - 2\sigma$  and  $\theta_\sigma'(\delta, \gamma) \equiv \theta_\sigma''(\delta, \gamma) - \gamma$ . Note that, for any  $\sigma \in (0, \sigma^\#)$ ,  $\delta > 0$  and  $\gamma > 0$  can be chosen so that  $0 < \theta_\sigma'(\delta, \gamma) < \theta_\sigma''(\delta, \gamma) < \theta_\sigma^*$ .

Consider the non-monotone deterministic policy  $\Gamma_{\delta, \gamma} \equiv \{\{0, 1\}, \pi_{\delta, \gamma}\}$  given by

$$\pi_{\delta, \gamma}(1|\theta) \equiv \mathbf{1} \{ \theta \in [\theta_\sigma'(\delta, \gamma), \theta_\sigma''(\delta, \gamma)] \cup [\theta_\sigma^*, \infty) \}.$$

We show that, for any  $\sigma \in (0, \sigma^\#)$ , there exist  $\delta, \gamma > 0$  such that (i)  $0 \leq \theta_\sigma'(\delta, \gamma) < \theta_\sigma''(\delta, \gamma) < \theta_\sigma^*$ , and (ii)  $V_\sigma^{\Gamma_{\delta, \gamma}}(\theta) \geq 0$  for all  $\theta > \theta_\sigma'(\delta, \gamma)/(1+2\sigma)$ , with  $V_\sigma^{\Gamma_{\delta, \gamma}}(\theta) = 0$  only for  $\theta = \theta^{MS}$ .<sup>2</sup>

First observe that, for any  $\sigma \in (0, \sigma^\#)$ ,  $\delta \in (0, \theta^{MS} - \frac{2\sigma}{1+2\sigma})$  and

$$0 < \gamma \leq (1+2\sigma)(\theta^{MS} - \delta) - 2\sigma \equiv R_0(\delta, \theta^{MS}, \sigma)$$

guarantee that  $0 \leq \theta_\sigma'(\delta, \gamma) < \theta_\sigma''(\delta, \gamma) < \theta_\sigma^*$ .<sup>3</sup>

Next note that, for any  $(\sigma, \delta, \gamma)$  with  $\sigma \in (0, \sigma^\#)$ ,  $\delta \in (0, \theta^{MS} - 2\sigma/(1+2\sigma))$  and  $0 < \gamma \leq R_0(\delta, \theta^{MS}, \sigma)$ ,  $V_\sigma^{\Gamma_{\delta, \gamma}}(\theta) = V_\sigma^{\Gamma^{\theta_\sigma^*}}(\theta)$  for all  $\theta \in [\theta^{MS} - \delta, 1]$ . Indeed, for any  $\theta \in [\theta^{MS} - \delta, 1]$ ,  $x_\sigma^*(\theta) - \sigma > \theta_\sigma''(\delta, \gamma)$  implying that the the posterior beliefs of the marginal investor with signal  $x_\sigma^*(\theta)$  under the policy  $\Gamma_{\delta, \gamma}$  coincide with those under the policy  $\Gamma^{\theta_\sigma^*}$ .

<sup>2</sup>Consistently with the notation above,  $V_\sigma^{\Gamma_{\delta, \gamma}}(\theta)$  is the expected payoff of the marginal investor with signal  $x_\sigma^*(\theta)$  when the policy  $\Gamma_{\delta, \gamma}$  announces that  $s = 1$  and the quality of the agents' exogenous signals is parametrized by  $\sigma$ . For any  $\theta < \theta_\sigma'(\delta, \gamma)/(1+2\sigma)$ ,  $x_\sigma^*(\theta) + \sigma < \theta_\sigma'$ , which implies that the signal  $x_\sigma^*(\theta)$  is not consistent with the event that fundamentals are above  $\theta_\sigma'(\delta, \gamma)$ . Equivalently, because the lowest signal that is consistent with  $\theta \in [\theta_\sigma'(\delta, \gamma), \theta_\sigma''(\delta, \gamma)] \cup [\theta_\sigma^*, \infty)$  is  $\theta_\sigma'(\delta, \gamma) - \sigma$ , the lowest default threshold is  $\theta_\sigma'(\delta, \gamma)/(1+2\sigma)$ .

<sup>3</sup>Observe that  $\sigma \in (0, \sigma^\#)$  implies that  $\theta^{MS} - 2\sigma/(1+2\sigma) > 0$ . In turn,  $\delta \in (0, \theta^{MS} - 2\sigma/(1+2\sigma))$  implies that  $0 < \theta_\sigma''(\delta, \gamma) < \theta_\sigma^*$  and that  $R_0(\delta, \theta^{MS}, \sigma) > 0$ . Finally, that  $0 < \gamma \leq R_0(\delta, \theta^{MS}, \sigma)$  implies that  $0 \leq \theta_\sigma'(\delta, \gamma) < \theta_\sigma''(\delta, \gamma)$ .

Let  $\theta_\sigma^\#(\delta, \gamma)$  be such that  $x_\sigma^*(\theta_\sigma^\#(\delta, \gamma)) - \sigma = \theta'_\sigma(\delta, \gamma)$ . Dropping the arguments of  $\theta_\sigma^\#(\delta, \gamma)$ ,  $\theta'_\sigma(\delta, \gamma)$  and  $\theta''_\sigma(\delta, \gamma)$  to ease the notation, we have that

$$\theta' = \theta'' - \gamma = x_\sigma^*(\theta^{MS} - \delta) - \sigma - \gamma = (1 + 2\sigma)(\theta^{MS} - \delta) - 2\sigma - \gamma.$$

From the definition of  $\hat{\theta}$  we have that  $x_\sigma^*(\hat{\theta}) - \sigma = (1 + 2\sigma)\theta^\# - 2\sigma = \theta'$ . Combining the above two results we obtain that  $\theta^\# = \theta^{MS} - \delta - \gamma / (1 + 2\sigma)$ . Fixing  $\sigma \in (0, \sigma^\#)$ , note that, for  $\delta, \gamma > 0$  small,  $\theta^\# \geq \theta_\sigma^*$ . Specifically, for any  $\sigma \in (0, \sigma^\#)$  and any  $0 < \delta < 2\sigma(1 - \theta^{MS})$ ,  $\theta^\# \geq \theta_\sigma^*$  if and only if

$$\gamma \leq (1 + 2\sigma)(2\sigma(1 - \theta^{MS}) - \delta) \equiv R_1(\delta, \theta^{MS}, \sigma).$$

Next, observe that, for any  $\theta \in [\theta^\#, \theta^{MS} - \delta)$ ,

$$\begin{aligned} V_\sigma^{\Gamma, \delta, \gamma}(\theta) &= g - (g + |b|) \mathbb{P}_\sigma\{\tilde{\theta} \leq \theta | \tilde{\theta} \in [x_\sigma^*(\theta) - \sigma, \theta''] \cup [\theta_\sigma^*, \infty); x_\sigma^*(\theta)\} \\ &= g - (g + |b|) (\theta'' - \theta_\sigma^* + 2\sigma(1 - \theta)) / (\theta'' - \theta_\sigma^* + 2\sigma), \end{aligned}$$

which is strictly increasing in  $\theta$ . Similarly, for any  $\theta \in [\theta_\sigma^*, \theta^\#)$ ,

$$\begin{aligned} V_\sigma^{\Gamma, \delta, \gamma}(\theta) &= g - (g + |b|) \mathbb{P}_\sigma\{\tilde{\theta} \leq \theta | \tilde{\theta} \in [\theta', \theta''] \cup [\theta_\sigma^*, \infty); x_\sigma^*(\theta)\} \\ &= g - (g + |b|) \frac{\theta - \theta_\sigma^* + \gamma}{x_\sigma^*(\theta) + \sigma - \theta_\sigma^* + \gamma} = g - (g + |b|) \frac{\theta - \theta_\sigma^* + \gamma}{(1 + 2\sigma)\theta - \theta_\sigma^* + \gamma}, \end{aligned}$$

which is strictly decreasing for any  $\gamma \leq \theta_\sigma^*$ . Note that  $\theta' \geq 0$  requires that  $\gamma \leq \theta_\sigma^*$ . Next, note that, for  $\theta \in [\theta'', \theta_\sigma^*)$ ,

$$\begin{aligned} V_\sigma^{\Gamma, \delta, \gamma}(\theta) &= g - (g + |b|) \mathbb{P}_\sigma\{\tilde{\theta} \leq \theta | \tilde{\theta} \in [\theta', \theta''] \cup [\theta_\sigma^*, \infty); x_\sigma^*(\theta)\} \\ &= g - (g + |b|) \frac{\gamma}{x_\sigma^*(\theta) + \sigma - \theta_\sigma^* + \gamma} = g - (g + |b|) \frac{\gamma}{(1 + 2\sigma)\theta - \theta_\sigma^* + \gamma}, \end{aligned}$$

and, therefore,  $V_\sigma^{\Gamma, \delta, \gamma}(\cdot)$  is increasing over the range  $[\theta'', \theta_\sigma^*)$ . Finally, for  $\theta \in [\theta', \theta'')$ , we have that

$$\begin{aligned} V_\sigma^{\Gamma, \delta, \gamma}(\theta) &= g - (g + |b|) \mathbb{P}_\sigma\{\tilde{\theta} \leq \theta | \tilde{\theta} \in [\theta', \theta''] \cup [\theta_\sigma^*, \infty); x_\sigma^*(\theta)\} \\ &= g - (g + |b|) \frac{\theta - \theta'}{x_\sigma^*(\theta) + \sigma - \theta_\sigma^* + \gamma} = g - (g + |b|) \frac{\theta - \theta'}{(1 + 2\sigma)\theta - \theta_\sigma^* + \gamma}. \end{aligned}$$

Hence  $V_\sigma^{\Gamma, \delta, \gamma}(\cdot)$  is decreasing over  $[\theta', \theta'')$  if  $(1 + 2\sigma)\theta' = x_\sigma^*(\theta') + \sigma > \theta_\sigma^*$ . Using the fact that  $\theta' = \theta'' - \gamma$ , together with the fact that  $\theta'' = x_\sigma^*(\theta^{MS} - \delta) - \sigma$  and  $\theta_\sigma^* = (1 + 2\sigma)\theta^{MS} - 2\sigma$ ,

we have that  $(1 + 2\sigma)\theta' > \theta_\sigma^*$  if

$$\gamma < 2\sigma [(1 + 2\sigma)\theta^{MS} - 2\sigma] / (1 + 2\sigma) - (1 + 2\sigma)\delta \equiv R_2(\delta, \theta^{MS}, \sigma).$$

Lastly, observe that, for any  $\theta \in [\theta'/(1 + 2\sigma), \theta']$ ,  $V_\sigma^{\Gamma, \delta, \gamma}(\theta) = g$ .

We thus have that the function  $V_\sigma^{\Gamma, \delta, \gamma}$  is such that (1)  $V_\sigma^{\Gamma, \delta, \gamma}(\theta) \geq 0$  for all  $\theta \geq \theta'/(1 + 2\sigma)$ , and (2)  $V_\sigma^{\Gamma, \delta, \gamma}(\theta) = 0$  only if  $\theta = \theta^{MS}$ , if and only if the following conditions hold: (a)  $V_\sigma^{\Gamma, \delta, \gamma}(\theta^\#) > 0$ , and (b)  $V_\sigma^{\Gamma, \delta, \gamma}(\theta'') > 0$ . Requiring that  $V_\sigma^{\Gamma, \delta, \gamma}(\theta^\#) > 0$  is equivalent to

$$\begin{aligned} g - (g + |b|) (\theta^\# - \theta_\sigma^* + \gamma) / (x_\sigma^*(\theta^\#) + \sigma - \theta_\sigma^* + \gamma) &> 0 \\ \Leftrightarrow \theta^{MS}(\theta_\sigma^* - \gamma) - ((1 + 2\sigma)\theta^{MS} - 2\sigma)\theta^\# &> 0. \end{aligned}$$

Recall that  $\theta_\sigma^* = (1 + 2\sigma)\theta^{MS} - 2\sigma$ . Using the fact that  $\theta^\# = \theta^{MS} - \delta - \frac{\gamma}{1 + 2\sigma}$ , we conclude that a sufficient condition for  $V_\sigma^{\Gamma, \delta, \gamma}(\theta^\#) > 0$  is that

$$\begin{aligned} (\theta^{MS} - \delta - \gamma/(1 + 2\sigma))\theta_\sigma^* &< \theta^{MS}(\theta_\sigma^* - \gamma) \\ \Leftrightarrow \gamma &< \delta(1 + 2\sigma)((1 + 2\sigma)\theta^{MS} - 2\sigma) / (2\sigma) \equiv R_3(\delta, \theta^{MS}, \sigma). \end{aligned}$$

Next, observe that  $V_\sigma^{\Gamma, \delta, \gamma}(\theta'') > 0$  is equivalent to

$$\begin{aligned} \gamma &< (1 - \theta^{MS})((1 + 2\sigma)\theta'' - \theta_\sigma^* + \gamma) \\ \Leftrightarrow \gamma &< \left(\frac{1 - \theta^{MS}}{\theta^{MS}}\right)((1 + 2\sigma)[(1 + 2\sigma)(\theta^{MS} - \delta) - 2\sigma] - (1 + 2\sigma)\theta^{MS} + 2\sigma) \equiv R_4(\delta, \theta^{MS}, \sigma). \end{aligned}$$

We conclude that, for any  $\sigma \in (0, \sigma^\#)$ , (i)  $0 \leq \theta'_\sigma(\delta, \gamma) < \theta''_\sigma(\delta, \gamma) < \theta_\sigma^*$ , and (ii)  $V_\sigma^{\Gamma, \delta, \gamma}(\theta) \geq 0$  for all  $\theta > \theta'_\sigma(\delta, \gamma)/(1 + 2\sigma)$ , with  $V_\sigma^{\Gamma, \delta, \gamma}(\theta) = 0$  only for  $\theta = \theta^{MS}$ , if

$$0 < \delta < \min \left\{ \theta^{MS} - \frac{2\sigma}{1 + 2\sigma}, 2\sigma(1 - \theta^{MS}), \frac{2\sigma[(1 + 2\sigma)\theta^{MS} - 2\sigma]}{(1 + 2\sigma)^2}, \frac{2\sigma}{1 + 2\sigma} \left[ \theta^{MS} - \frac{2\sigma}{1 + 2\sigma} \right] \right\} \equiv \varsigma(\theta^{MS}, \sigma)$$

and  $0 < \gamma < \min_{0 \leq i \leq 4} R_i(\delta, \theta^{MS}, \sigma)$ . Note that  $\sigma < \sigma^\#$  implies that  $\varsigma(\theta^{MS}, \sigma) > 0$ , whereas  $\delta < \varsigma(\theta^{MS}, \sigma)$  implies that  $\min_{0 \leq i \leq 4} R_i(\delta, \theta^{MS}, \sigma) > 0$ . Finally note that, for any  $\sigma \in (0, \sigma^\#)$ , and any  $\theta \geq \theta'_\sigma(\delta, \gamma)$ , the payoff  $V_\sigma^{\Gamma, \delta, \gamma}(\theta)$  is continuous in the threshold  $\theta_\sigma^*$ . Hence there exists a policy  $\Gamma$  whose rule  $\pi$  is given by  $\pi(1|\theta) \equiv \mathbf{1}\{\theta \in [\theta'_\sigma(\delta, \gamma), \theta''_\sigma(\delta, \gamma)] \cup [\theta_\sigma^* + \varepsilon, \infty)\}$  with  $\varepsilon > 0$  arbitrarily small, such that  $\Gamma$  strictly improves over  $\Gamma^{\theta_\sigma^*}$  and is such that  $V_\sigma^\Gamma(\theta) > 0$  for all  $\theta > \theta'_\sigma(\delta, \gamma)/(1 + 2\sigma)$ , implying that  $\Gamma$  satisfies PCP. Q.E.D.

## Proof of Example 3 in Main Text

**Preliminaries.** For any  $\theta \in (0, 1)$ , any  $\sigma \in \mathbb{R}_+$ , note that, in this example  $x_\sigma^*(\theta) \equiv \theta + \sigma\Phi^{-1}(\theta)$ , where  $\Phi$  is the cdf of the standard Normal distribution and  $\phi$  its density. Also let  $x_\sigma^*(0) \equiv -\infty$  and  $x_\sigma^*(1) \equiv +\infty$ . For any  $(\theta_0, \hat{\theta}, \sigma) \in (0, 1) \times \mathbb{R} \times \mathbb{R}_+$ , let  $\psi(\theta_0, \hat{\theta}, \sigma)$  denote the payoff from pledging of an investor with private signal  $x_\sigma^*(\theta_0)$ , when default occurs if and only if  $\theta \leq \theta_0$ , the policy reveals that  $\theta \geq \hat{\theta}$ , and the precision of private information is  $\sigma^{-2}$ . Then let  $\hat{\sigma} \equiv \inf \{\sigma \in \mathbb{R}_+ : \psi(\theta_0, 0, \sigma) > 0 \text{ all } \theta_0 \in (0, 1)\}$  if  $\{\sigma \in \mathbb{R}_+ : \psi(\theta_0, 0, \sigma) > 0 \text{ all } \theta_0 \in (0, 1)\} \neq \emptyset$  and else  $\hat{\sigma} = +\infty$ .<sup>4</sup> Then let  $\Psi(\sigma) \equiv \inf_{\theta_0 \in (0, 1)} \psi(\theta_0, 0, \sigma)$  and note that  $\lim_{\sigma \rightarrow 0^+} \Psi(\sigma) < 0$ , implying that  $\hat{\sigma} > 0$ . For any  $\sigma \in \mathbb{R}_+$  for which  $\psi(\theta_0, 0, \sigma) > 0$  for all  $\theta_0 \in (0, 1)$ , the policy maker can avoid default for every  $\theta > 0$  by using the monotone rule  $\pi(\theta) = \mathbf{1}\{\theta > 0\}$ . This case is uninteresting. Hereafter, we thus confine attention to the case in which  $\sigma < \hat{\sigma}$ .

Let  $U_\sigma^\Gamma(x, 1|x)$  denote the payoff from pledging of an agent with signal  $x$  who expects all other agents to pledge if and only if their signal exceeds  $x$ , when the precision of private information is  $\sigma^{-2}$ , and the policy  $\Gamma$  announces that  $s = 1$ . Also let

$$U_\sigma^\Gamma(x_\sigma^*(0), 1|x_\sigma^*(0)) \equiv \lim_{x \rightarrow -\infty} U_\sigma^\Gamma(x, 1|x)$$

and

$$U_\sigma^\Gamma(x_\sigma^*(1), 1|x_\sigma^*(1)) \equiv \lim_{x \rightarrow +\infty} U_\sigma^\Gamma(x, 1|x).$$

Now let  $\mathbb{G}_\sigma$  denote the set of deterministic binary policies  $\Gamma = (\{0, 1\}, \pi)$  such that,  $\pi(\theta) = 0$  for all  $\theta \leq 0$ ,  $\pi(\theta) = 1$  for all  $\theta > 1$  and  $U_\sigma^\Gamma(x, 1|x) \geq 0$  for all  $x \in \mathbb{R}$ . From the proofs of Theorems 1 and 2, observe that, given any  $\sigma$ , any deterministic binary policy  $\Gamma$  satisfying PCP and such that  $\pi(\theta) = 0$  for all  $\theta \leq 0$  and  $\pi(\theta) = 1$  for all  $\theta > 1$  belongs in  $\mathbb{G}_\sigma$ . However,  $\mathbb{G}_\sigma$  contains also policies that do not satisfy PCP.<sup>5</sup>

**Proof Structure.** The proof is in four steps. Step 1 establishes that, when  $\sigma$  is small, under any policy  $\Gamma = (\{0, 1\}, \pi) \in \mathbb{G}_\sigma$ , any interval  $(\theta', \theta''] \subset (0, \theta^{MS}]$  receiving a pass grade

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<sup>4</sup>Recall that, when the announcement that  $s = 1$  reveals that  $\theta \geq 0$ , the unique rationalizable profile features all agents pledging, irrespective of  $x$ , if and only if  $\psi(\theta_0, 0, \sigma) > 0$  for all  $\theta_0 \in (0, 1)$ . This follows directly from Lemma 1 in the main text.

<sup>5</sup>These are those for which there exists  $x$  such that  $U_\sigma^\Gamma(x, 1|x) = 0$ ; when  $\Gamma$  announces  $s = 1$ , in addition to the rationalizable profile under which all agents pledge, there also exists a rationalizable profile under which each agent pledges if and only if his signal exceeds  $x$ .

(i.e., such that  $\pi(\theta) = 1$  for all  $\theta \in (\theta', \theta'']$ ) has a sufficiently small Lebesgue measure, with the measure vanishing as  $\sigma \rightarrow 0^+$ .

Step 2 then considers an *auxiliary game*  $G_\sigma$  in which the agents play less aggressively than under MARP. Namely,  $G_\sigma$  is the game in which (i) the policy maker's choice set is  $\mathbb{G}_\sigma$  and (ii) given *any* policy  $\Gamma \in \mathbb{G}_\sigma$ , all agents pledge after receiving the signal  $s = 1$  and refrain from pledging after receiving the signal  $s = 0$ .<sup>6</sup> We show that, when  $\sigma$  is small, given any policy  $\Gamma \in \mathbb{G}_\sigma$  that gives a fail grade to an interval  $(\theta', \theta''] \subseteq (\underline{\theta}, \theta^{MS}]$  of large Lebesgue measure, there exists another policy  $\Gamma^\# \in \mathbb{G}_\sigma$  that gives a pass grade to a  $F$ -positive measure subset of  $(\theta', \theta'']$ , has a mesh smaller than  $\Gamma$ , and is such that, when agents play as in  $G_\sigma$ , the probability of default under  $\Gamma^\#$  is strictly smaller than under  $\Gamma$ .

Step 3 then combines the results from Steps 1 and 2 to show that, when  $\sigma$  is small, given any policy  $\Gamma \in \mathbb{G}_\sigma$  for which the mesh  $M(\Gamma)$  of  $(0, \theta^{MS}]$  is larger than  $\varepsilon$ , there exists another policy  $\Gamma' \in \mathbb{G}_\sigma$  with a mesh  $M(\Gamma')$  smaller than  $\varepsilon$  such that, when agents play as in  $G_\sigma$ , the probability of default is strictly smaller under  $\Gamma'$  than under  $\Gamma$ . Starting from  $\Gamma' \in \mathbb{G}_\sigma$  one can then construct a “nearby” policy  $\Gamma^* \in \mathbb{G}_\sigma$  such that the probability of default under  $\Gamma^*$  is arbitrarily close to that under  $\Gamma'$  (and hence strictly smaller than under the original policy  $\Gamma$ ) and such that  $U_\sigma^{\Gamma^*}(x, 1|x) > 0$  for all  $x$ . The last property implies that  $\Gamma^*$  satisfies PCP also when agents play according to MARP. The policy  $\Gamma^*$  thus strictly improves upon  $\Gamma$  also in the original game, as claimed in the main text.

Finally, step 4 closes the proof by showing how to construct the function  $\mathcal{E}$  relating the noise  $\sigma$  in the agents' exogenous private information to the bound  $\mathcal{E}(\sigma)$  on the mesh of the policies.

**Step 1.** We start with the following result:

**Lemma S3-A.** *For any  $\varepsilon \in \mathbb{R}_{++}$ , there exists  $\sigma(\varepsilon) \in \mathbb{R}_{++}$  such that, for any  $\sigma \in (0, \sigma(\varepsilon)]$ , the following is true: for any policy  $\Gamma = (\{0, 1\}, \pi) \in \mathbb{G}_\sigma$  and any cell  $(\theta', \theta''] \in D^\Gamma$  with  $|\theta'' - \theta'| > \varepsilon$ , necessarily  $\pi(\theta) = 0$ .*

**Proof of Lemma S3-A.** We first show (Property S3-A below) that, for any  $\sigma > 0$ , if the policy maker were to replace  $\Gamma$  with the cutoff policy  $\Gamma^{\theta'}$ , then for any  $\theta \leq \theta''$ ,  $U_\sigma^{\Gamma^{\theta'}}(x_\sigma^*(\theta), 1|x_\sigma^*(\theta)) \geq U_\sigma^\Gamma(x_\sigma^*(\theta), 1|x_\sigma^*(\theta))$ .<sup>7</sup> Next, we show (Property S3-B below) that, for

<sup>6</sup>The agent's behavior is consistent with MARP only for those  $\Gamma \in \mathbb{G}_\sigma$  for which, for all  $x$ ,  $U_\sigma^\Gamma(x, 1|x) > 0$ .

<sup>7</sup>For any  $\hat{\theta} \in [0, 1]$ ,  $\Gamma^{\hat{\theta}} = (\{0, 1\}, \pi^{\hat{\theta}})$  is the deterministic monotone policy with cut-off  $\hat{\theta}$ .



any  $\theta > \theta'$ , as  $\sigma$  goes to zero,  $U_{\sigma}^{\Gamma^{\theta'}}(x_{\sigma}^*(\theta), 1|x_{\sigma}^*(\theta))$  converges uniformly to  $\int_0^1 u(\theta, A)dA$ . Because  $\int_0^1 u(\theta, A)dA < 0$  for  $\theta < \theta^{MS}$ , the above two properties imply that, for  $\sigma$  small,  $U_{\sigma}^{\Gamma}(x_{\sigma}^*(\theta), 1|x_{\sigma}^*(\theta)) < 0$  for some  $\theta \in (\theta', \theta'']$ , and hence that  $\Gamma \notin \mathbb{G}_{\sigma}$ . The result in the lemma then follows by contrapositive.

**Property S3-A.** For any policy  $\Gamma = (\{0, 1\}, \pi) \in \mathbb{G}_{\sigma}$  and any cell  $(\theta', \theta''] \in D^{\Gamma}$  such that  $\pi(\theta) = 1$  for all  $\theta \in (\theta', \theta'']$ ,  $U_{\sigma}^{\Gamma}(x_{\sigma}^*(\theta), 1|x_{\sigma}^*(\theta)) \leq U_{\sigma}^{\Gamma^{\theta'}}(x_{\sigma}^*(\theta), 1|x_{\sigma}^*(\theta))$  for all  $\theta \leq \theta''$ .

**Proof of Property S3-A.** The proof follows from Results S3-A-1 and S3-A-2 below.

**Result S3-A-1.** Pick any policy  $\Gamma = (\{0, 1\}, \pi) \in \mathbb{G}_{\sigma}$ . Given the partition  $D^{\Gamma} \equiv \{d_i = (\underline{\theta}_i, \bar{\theta}_i] : i = 1, \dots, N\}$  of  $(0, \theta^{MS}]$  induced by  $\Gamma$ , take any cell  $d_i = (\underline{\theta}_i, \bar{\theta}_i]$  for which  $\pi(\theta) = 1$  for all  $\theta \in d_i$ . Let  $\Gamma_L^i = \{\{0, 1\}, \pi_L^i\} \in \mathbb{G}_{\sigma}$  be the policy constructed as follows: (a)  $\pi_L^i(\theta) = 0$  for all  $\theta \leq \underline{\theta}_i$ ; and (b)  $\pi_L^i(\theta) = \pi(\theta)$  for all  $\theta > \underline{\theta}_i$ . Then, for all  $\theta \in [0, 1]$ ,  $U_{\sigma}^{\Gamma_L^i}(x_{\sigma}^*(\theta), 1|x_{\sigma}^*(\theta)) \geq U_{\sigma}^{\Gamma}(x_{\sigma}^*(\theta), 1|x_{\sigma}^*(\theta))$ .

**Proof of Result S3-A-1.** Note that, under the new policy,  $\pi_L^i(\theta) = \pi(\theta) \times \mathbf{1}\{\theta > \underline{\theta}_i\}$ . The posterior beliefs  $\Lambda_{\sigma}^{\Gamma_L^i}(\cdot|x, 1)$  about  $\theta$  of an agent with exogenous signal  $x$  and endogenous signal  $s = 1$  under the new policy  $\Gamma_L^i$  thus dominate, in the FOSD sense, the analogous beliefs  $\Lambda_{\sigma}^{\Gamma}(\cdot|x, 1)$  under the original policy  $\Gamma$ .<sup>8</sup> The result then follows from the fact that, given any default threshold  $\theta$ , the payoff from pledging when the fundamentals are equal to  $\tilde{\theta}$  and default occurs if and only if  $\tilde{\theta} \leq \theta$  is nondecreasing in  $\tilde{\theta}$ . *End of Proof of Result S3-A-1.*

**Result S3-A-2.** Pick any policy  $\Gamma = \{\{0, 1\}, \pi\} \in \mathbb{G}_{\sigma}$ . Given the partition  $D^{\Gamma} \equiv \{d_i = (\underline{\theta}_i, \bar{\theta}_i] : i = 1, \dots, N\}$  of  $(0, \theta^{MS}]$  induced by  $\Gamma$ , take any cell  $d_i = (\underline{\theta}_i, \bar{\theta}_i]$ ,  $i \geq 2$ , for which  $\pi(\theta) = 1$  for all  $\theta \in d_i$ . Let  $\Gamma_R^i = \{\{0, 1\}, \pi_R^i\} \in \mathbb{G}_{\sigma}$  be the policy constructed from  $\Gamma$  as follows: (a)  $\pi_R^i(\theta) = \pi(\theta)$  for all  $\theta \leq \underline{\theta}_i$ ; and (b)  $\pi_R^i(\theta) = 1$  for all  $\theta > \underline{\theta}_i$ . Then, for all  $\theta \leq \bar{\theta}_i$ ,  $U_{\sigma}^{\Gamma_R^i}(x_{\sigma}^*(\theta), 1|x_{\sigma}^*(\theta)) \geq U_{\sigma}^{\Gamma}(x_{\sigma}^*(\theta), 1|x_{\sigma}^*(\theta))$ .

**Proof of Result S3-A-2.** Let  $\Theta^1 \equiv \{\theta \in \Theta : \pi(\theta) = 1\}$  and  $\Theta_i^0 \equiv \{\theta \in (\underline{\theta}_i, 1] : \pi(\theta) = 0\}$ .

<sup>8</sup>No matter the shape of the beliefs  $\Lambda_{\sigma}^{\Gamma}(\cdot|x, 1)$ , the announcement that  $\theta > \underline{\theta}_i$  is always “good news” in the sense of Milgrom (1981) and hence  $\Lambda_{\sigma}^{\Gamma_L^i}(\cdot|x, 1) \succ_{FOSD} \Lambda_{\sigma}^{\Gamma}(\cdot|x, 1)$ .

For any  $\theta^\# \leq \bar{\theta}_i$ , and any  $x$ ,

$$\begin{aligned}\Lambda_{\sigma}^{\Gamma^i}(\theta^\#|x, 1) &= \Pr\{\theta \leq \theta^\# | x, \theta \in (\Theta^1 \cup \Theta_i^0)\} = \frac{\Pr\{\theta \leq \theta^\# \wedge \theta \in (\Theta^1 \cup \Theta_i^0) | x\}}{\Pr\{\theta \in (\Theta^1 \cup \Theta_i^0) | x\}} \\ &= \frac{\Pr\{\theta \leq \theta^\# \wedge \theta \in \Theta^1 | x\}}{\Pr\{\theta \in (\Theta^1 \cup \Theta_i^0) | x\}} + \frac{\Pr\{\theta \leq \theta^\# \wedge \theta \in \Theta_i^0 | x\}}{\Pr\{\theta \in (\Theta^1 \cup \Theta_i^0) | x\}} = \frac{\Pr\{\theta \leq \theta^\# \wedge \theta \in \Theta^1 | x\}}{\Pr\{\theta \in (\Theta^1 \cup \Theta_i^0) | x\}} \\ &\leq \Pr\{\theta \leq \theta^\# | x, \theta \in \Theta^1\} = \Lambda_{\sigma}^{\Gamma}(\theta^\# | x, 1).\end{aligned}$$

Given the above inequality, and the fact that  $b < 0 < g$ , we then have that, for any  $\theta \leq \bar{\theta}_i$ ,

$$\begin{aligned}U_{\sigma}^{\Gamma}(x_{\sigma}^*(\theta), 1 | x_{\sigma}^*(\theta)) &= b \cdot \Lambda_{\sigma}^{\Gamma}(\theta | x_{\sigma}^*(\theta), 1) + g \cdot (1 - \Lambda_{\sigma}^{\Gamma}(\theta | x_{\sigma}^*(\theta), 1)) \\ &\leq b \cdot \Lambda_{\sigma}^{\Gamma^i}(\theta | x_{\sigma}^*(\theta), 1) + g \cdot (1 - \Lambda_{\sigma}^{\Gamma^i}(\theta | x_{\sigma}^*(\theta), 1)) = U_{\sigma}^{\Gamma^i}(x_{\sigma}^*(\theta), 1 | x_{\sigma}^*(\theta)).\end{aligned}$$

*End of Proof of Result S3-A-2.*

Property S3-A follows from Results S3-A-1 and S3-A-2, by taking the cell  $d_i = (\theta', \theta'']$ .  $\square$

Now, fix  $\varepsilon \in (0, \theta^{MS})$ . For any  $\theta^* \in [0, \theta^{MS} - \varepsilon]$ , let  $\Gamma^{\theta^*}$  be the monotone rule with cut-off  $\theta^*$ . For any  $\theta^* \in [0, \theta^{MS} - \varepsilon]$ , any  $\sigma \in \mathbb{R}_{++}$ , let

$$H_{\sigma}(\theta^*; \varepsilon) \equiv \inf_{\theta \in [\theta^*, \theta^* + \varepsilon]} U_{\sigma}^{\Gamma^{\theta^*}}(x_{\sigma}^*(\theta), 1 | x_{\sigma}^*(\theta)).$$

Note that  $U_{\sigma}^{\Gamma^{\theta^*}}(x_{\sigma}^*(\theta), 1 | x_{\sigma}^*(\theta))$  is continuous in  $(\theta^*, \theta, \sigma)$  over  $[0, 1]^2 \times (0, \hat{\sigma}]$ . From Berge's Maximum Theorem,  $H_{\sigma}(\theta^*; \varepsilon)$  is thus continuous in  $(\theta^*, \sigma)$  over  $[0, \theta^{MS} - \varepsilon] \times (0, \hat{\sigma}]$ .

For all  $\theta^* \in [0, \theta^{MS} - \varepsilon]$ , all  $\theta \in (\theta^*, \theta^* + \varepsilon]$ ,  $\lim_{\sigma \rightarrow 0^+} U_{\sigma}^{\Gamma^{\theta^*}}(x_{\sigma}^*(\theta), 1 | x_{\sigma}^*(\theta)) = \int_0^1 u(\theta, A) dA$ . Because  $\int_0^1 u(\theta, A) dA$  is strictly increasing in  $\theta$  and equal to zero at  $\theta = \theta^{MS}$ , for any  $\theta^* \in [0, \theta^{MS} - \varepsilon]$ ,  $H_{0^+}(\theta^*; \varepsilon) \equiv \lim_{\sigma \rightarrow 0^+} H_{\sigma}(\theta^*; \varepsilon) = \lim_{\sigma \rightarrow 0^+} \lim_{\theta \rightarrow \theta^* +} U_{\sigma}^{\Gamma^{\theta^*}}(x_{\sigma}^*(\theta), 1 | x_{\sigma}^*(\theta)) = \int_0^1 u(\theta^*, A) dA$ . We show next that  $H_{\sigma}(\cdot; \varepsilon)$  converges uniformly to the limit function  $H_{0^+}(\cdot; \varepsilon)$  over  $[0, \theta^{MS} - \varepsilon]$ .

**Property S3-B.** Fix  $\varepsilon \in (0, \theta^{MS})$ . For any  $\epsilon < \varepsilon$ , there exists  $\sigma'(\epsilon) > 0$  such that, for any  $\sigma \leq \sigma'(\epsilon)$ , and any  $\theta^* \in [0, \theta^{MS} - \varepsilon]$ ,  $|H_{\sigma}(\theta^*; \varepsilon) - H_{0^+}(\theta^*; \varepsilon)| < \epsilon$ .

**Proof of Property S3-B.** The limit function  $H_{0^+}(\cdot; \varepsilon)$  is *uniformly* continuous over  $[0, \theta^{MS} - \varepsilon]$ . As a consequence, there exists  $\delta > 0$  such that for any  $\theta, \tilde{\theta} \in [0, \theta^{MS} - \varepsilon]$ , with  $|\tilde{\theta} - \theta| \leq \delta$ , necessarily  $|H_{0^+}(\tilde{\theta}; \varepsilon) - H_{0^+}(\theta; \varepsilon)| < \epsilon/2$ . Next, let  $D_{\delta} \equiv \{(\underline{\theta}_i, \bar{\theta}_i) : i = 1, \dots, N\}$ ,  $N \in \mathbb{N}$ , be any interval partition of  $(0, \theta^{MS} - \varepsilon]$  with the property that every cell  $(\underline{\theta}_i, \bar{\theta}_i] \in D_{\delta}$  is

such that  $|\bar{\theta}_i - \underline{\theta}_i| \leq \delta$ . For any  $i = 1, \dots, N$ , any  $\sigma > 0$ , let  $\hat{\theta}_\sigma^i \equiv \sup\{\arg \max_{\theta \in [\underline{\theta}_i, \bar{\theta}_i]} H_\sigma(\theta; \varepsilon)\}$ . That  $H_\sigma(\theta; \varepsilon)$  is continuous in  $(\sigma, \theta)$  implies that the hypothesis of Berge's Maximum Theorem hold and, hence, the correspondence  $\arg \max_{\theta \in [\underline{\theta}_i, \bar{\theta}_i]} H_\sigma(\theta; \varepsilon)$  is compact-valued and upper hemi-continuous in  $\sigma$ . As a result, for any  $\sigma > 0$ ,  $\hat{\theta}_\sigma^i = \max\{\arg \max_{\theta \in [\underline{\theta}_i, \bar{\theta}_i]} H_\sigma(\theta; \varepsilon)\}$ . Moreover,  $\lim_{\sigma \rightarrow 0^+} H_\sigma(\hat{\theta}_\sigma^i; \varepsilon) = H_{0^+}(\hat{\theta}_{0^+}^i; \varepsilon)$ , where  $\hat{\theta}_{0^+}^i \equiv \lim_{\sigma \rightarrow 0^+} \hat{\theta}_\sigma^i$ .

For any  $\theta^* \in [0, \theta^{MS} - \varepsilon]$ , let  $(\underline{\theta}_j, \bar{\theta}_j] \in D_\delta$  be the partition cell containing  $\theta^*$ . Then,

$$\begin{aligned} & H_\sigma(\theta^*; \varepsilon) - H_{0^+}(\theta^*; \varepsilon) \leq H_\sigma(\hat{\theta}_\sigma^j; \varepsilon) - H_{0^+}(\theta^*; \varepsilon) \\ & = H_\sigma(\hat{\theta}_\sigma^j; \varepsilon) - H_{0^+}(\hat{\theta}_{0^+}^j; \varepsilon) + H_{0^+}(\hat{\theta}_{0^+}^j; \varepsilon) - H_{0^+}(\theta^*; \varepsilon) < H_\sigma(\hat{\theta}_\sigma^j; \varepsilon) - H_{0^+}(\hat{\theta}_{0^+}^j; \varepsilon) + \epsilon/2 < \epsilon \end{aligned}$$

for all  $\sigma < \bar{\sigma}_j(\epsilon)$ , for some  $\bar{\sigma}_j(\epsilon) > 0$ . The first inequality is by definition of  $\hat{\theta}_\sigma^j$ . The second inequality follows from the fact that  $|\hat{\theta}_{0^+}^j - \theta^*| < \delta$ . The last inequality follows from the fact that  $\lim_{\sigma \rightarrow 0^+} H_\sigma(\hat{\theta}_\sigma^j) = H_{0^+}(\hat{\theta}_{0^+}^j)$ . Similar arguments imply that  $H_\sigma(\theta^*; \varepsilon) - H_{0^+}(\theta^*; \varepsilon) > -\epsilon$  for all  $\sigma < \underline{\sigma}_j(\epsilon)$ , for some  $\underline{\sigma}_j(\epsilon) > 0$ .

Now let  $\sigma'(\epsilon) \equiv \min\{\min_{i \in N} \{\bar{\sigma}_i(\epsilon)\}, \min_{i \in N} \{\underline{\sigma}_i(\epsilon)\}\}$ . For any  $\sigma \leq \sigma'(\epsilon)$ , and any  $\theta^* \in [0, \theta^{MS} - \varepsilon]$ , we thus have that  $|H_\sigma(\theta^*; \varepsilon) - H_{0^+}(\theta^*; \varepsilon)| < \epsilon$ , thus proving that  $H_\sigma(\cdot; \varepsilon)$  converges uniformly to  $H_{0^+}(\cdot; \varepsilon)$  as  $\sigma \rightarrow 0^+$ . This completes the proof of Property S3-B.  $\square$

Next, given  $\varepsilon \in (0, \theta^{MS})$ , pick an arbitrary  $\eta \in (\int_0^1 u(\theta^{MS} - \varepsilon, A) dA, 0)$ . Because  $H_{0^+}(\theta^*; \varepsilon) \leq \eta$  for all  $\theta^* \in [0, \theta^{MS} - \varepsilon]$ , and because  $H_\sigma(\cdot; \varepsilon)$  converges uniformly to  $H_{0^+}(\cdot; \varepsilon)$ , there exists  $\sigma(\varepsilon) > 0$  such that, for any  $\sigma < \sigma(\varepsilon)$ , and any  $\theta^* \in [0, \theta^{MS} - \varepsilon]$ ,  $H_\sigma(\theta^*; \varepsilon) \leq \eta < 0$ . Therefore, for any  $\sigma < \sigma(\varepsilon)$ , and any monotone policy  $\Gamma^{\theta^*}$  with cut-off  $\theta^* \in [0, \theta^{MS} - \varepsilon]$ , there exists  $\theta \in [\theta^*, \theta^* + \varepsilon]$  such that  $U_\sigma^{\Gamma^{\theta^*}}(x_\sigma^*(\theta), 1|x_\sigma^*(\theta)) \leq \eta$ .

Together, Properties S3-A and S3-B then imply that, for any  $\sigma < \sigma(\varepsilon)$ , and any policy  $\Gamma$  such that  $\pi(\theta) = 1$  for all  $\theta \in (\theta', \theta'']$  for some  $(\theta', \theta''] \in D^\Gamma$  with  $|\theta'' - \theta'| > \varepsilon$ , necessarily  $U_\sigma^\Gamma(x_\sigma^*(\theta), 1|x_\sigma^*(\theta)) < 0$  for some  $\theta \in (\theta', \theta'']$ . Hence  $\Gamma \notin \mathbb{G}_\sigma$ . The claim in Lemma S3-A then follows by contrapositive. This completes the proof of Lemma S3-A.  $\blacksquare$

**Step 2.** Next, we show that, for any policy  $\Gamma = (\{0, 1\}, \pi) \in \mathbb{G}_\sigma$  that gives a fail grade to an interval  $(\theta', \theta''] \subseteq (0, \theta^{MS}]$  of large Lebesgue measure, there exists another policy  $\Gamma^\# \in \mathbb{G}_\sigma$  with a mesh  $M(\Gamma^\#) < M(\Gamma)$  such that, when agents play as in  $G_\sigma$ , the probability of default under  $\Gamma^\#$  is strictly smaller than under  $\Gamma$ . The result follows from Lemmas S3-B, S3-C and S3-D below.

**Lemma S3-B.** For any  $\Gamma = (\{0, 1\}, \pi) \in \mathbb{G}_\sigma$  such that  $\inf_{\theta \in [0, 1]} U_\sigma^\Gamma(x_\sigma^*(\theta), 1|x_\sigma^*(\theta)) > 0$ , there exists another policy  $\hat{\Gamma} = (\{0, 1\}, \hat{\pi}) \in \mathbb{G}_\sigma$ , with  $M(\hat{\Gamma}) \leq M(\Gamma)$ , such that, in the auxiliary game  $G_\sigma$ , the probability of default under  $\hat{\Gamma}$  is strictly smaller than under  $\Gamma$ .

**Proof of Lemma S3-B.** That  $\inf_{\theta \in [0, 1]} U_\sigma^\Gamma(x_\sigma^*(\theta), 1|x_\sigma^*(\theta)) > 0$  implies that, starting from  $\Gamma = (\{0, 1\}, \pi)$ , one can construct another policy  $\hat{\Gamma} = (\{0, 1\}, \hat{\pi})$  sufficiently close to  $\Gamma$  (in the  $L_1$  norm) and such that  $\hat{\pi}(\theta) \geq \pi(\theta)$  for all  $\theta$ , with the inequality strict over some positive  $F$ -measure set  $(\tilde{\theta}', \tilde{\theta}'') \subseteq (0, 1]$ , and such that (a)  $\hat{\pi}(\theta) = 0$  for all  $\theta \leq 0$ , (b)  $\hat{\pi}(\theta) = 1$  for all  $\theta > 1$ , (c)  $U_\sigma^{\hat{\Gamma}}(x, 1|x) \geq 0$  all  $x$ , and (d)  $M(\hat{\Gamma}) \leq M(\Gamma)$ . By definition of  $\mathbb{G}_\sigma$ ,  $\hat{\Gamma} \in \mathbb{G}_\sigma$ . That, in the auxiliary game  $G_\sigma$ , the probability of default under  $\hat{\Gamma}$  is strictly smaller than under  $\Gamma$ , then follows from the fact that all agents pledge when they receive the signal  $s = 1$ . This completes the proof of Lemma S3-B. ■

For any  $\sigma > 0$ , and any policy  $\Gamma = (\{0, 1\}, \pi) \in \mathbb{G}_\sigma$ ,  $U_\sigma^\Gamma(x_\sigma^*(\cdot), 1|x_\sigma^*(\cdot))$  is continuous over  $[0, 1]$ . Hence  $\inf_{\theta \in [0, 1]} U_\sigma^\Gamma(x_\sigma^*(\theta), 1|x_\sigma^*(\theta)) = \min_{\theta \in [0, 1]} U_\sigma^\Gamma(x_\sigma^*(\theta), 1|x_\sigma^*(\theta))$ .

**Lemma S3-C.** Let  $\Gamma = (\{0, 1\}, \pi) \in \mathbb{G}_\sigma$  be such that  $\min_{\theta \in [0, 1]} U_\sigma^\Gamma(x_\sigma^*(\theta), 1|x_\sigma^*(\theta)) = 0$ . For any  $\theta_\sigma^\# \in \arg \min_{\theta \in [0, 1]} U_\sigma^\Gamma(x_\sigma^*(\theta), 1|x_\sigma^*(\theta))$ , there exists  $\gamma_\sigma^\Gamma > 0$  such that  $\pi(\theta) = 1$  for  $F$ -almost all  $\theta \in (\theta_\sigma^\# - \gamma_\sigma^\Gamma, \theta_\sigma^\#)$ .

**Proof of Lemma S3-C.** The proof is by contraposition. Suppose there exists  $\delta > 0$  such that  $\pi(\theta) = 0$  for  $F$ -almost all  $\theta \in (\theta_\sigma^\# - \delta, \theta_\sigma^\#)$ . Observe that the sign of

$$U_\sigma^\Gamma(x_\sigma^*(\theta_\sigma^\# - \delta), 1|x_\sigma^*(\theta_\sigma^\# - \delta))$$

is the same as the sign of

$$b \int_{-\infty}^{\theta_\sigma^\# - \delta} \phi((x_\sigma^*(\theta_\sigma^\# - \delta) - \theta) / \sigma) \pi(\theta) dF(\theta) + g \int_{\theta_\sigma^\# - \delta}^{+\infty} \phi((x_\sigma^*(\theta_\sigma^\# - \delta) - \theta) / \sigma) \pi(\theta) dF(\theta).$$

Next observe that

$$\begin{aligned}
0 &= U_\sigma^\Gamma \left( x_\sigma^* \left( \theta_\sigma^\# \right), 1 | x_\sigma^* \left( \theta_\sigma^\# \right) \right) \int_{-\infty}^{+\infty} \phi \left( \left( x_\sigma^* \left( \theta_\sigma^\# \right) - \theta \right) / \sigma \right) \pi(\theta) dF(\theta) \\
&= \int_{-\infty}^{+\infty} \left( b \mathbf{1} \left\{ \theta \leq \theta_\sigma^\# \right\} + g \mathbf{1} \left\{ \theta > \theta_\sigma^\# \right\} \right) \phi \left( \left( x_\sigma^* \left( \theta_\sigma^\# \right) - \theta \right) / \sigma \right) \pi(\theta) dF(\theta) \\
&> \int_{-\infty}^{+\infty} \left( b \mathbf{1} \left\{ \theta \leq \theta_\sigma^\# \right\} + g \mathbf{1} \left\{ \theta > \theta_\sigma^\# \right\} \right) \phi \left( \left( x_\sigma^* \left( \theta_\sigma^\# - \delta \right) - \theta \right) / \sigma \right) \pi(\theta) dF(\theta) \\
&= \int_{-\infty}^{+\infty} \left( b \mathbf{1} \left\{ \theta \leq \theta_\sigma^\# - \delta \right\} + g \mathbf{1} \left\{ \theta > \theta_\sigma^\# - \delta \right\} \right) \phi \left( \left( x_\sigma^* \left( \theta_\sigma^\# - \delta \right) - \theta \right) / \sigma \right) \pi(\theta) dF(\theta) \\
&= U_\sigma^\Gamma \left( x_\sigma^* \left( \theta_\sigma^\# - \delta \right), 1 | x_\sigma^* \left( \theta_\sigma^\# - \delta \right) \right) \int_{-\infty}^{+\infty} \phi \left( \left( x_\sigma^* \left( \theta_\sigma^\# - \delta \right) - \theta \right) / \sigma \right) \pi(\theta) dF(\theta)
\end{aligned}$$

The first equality follows from the assumptions of the lemma. The second equality follows from the definition of the function  $U_\sigma^\Gamma \left( x_\sigma^* \left( \theta_\sigma^\# \right), 1 | x_\sigma^* \left( \theta_\sigma^\# \right) \right)$ . The inequality follows from the monotonicity of  $x_\sigma^* \left( \cdot \right)$ , the fact that  $\phi \left( \left( x - \theta \right) / \sigma \right)$  is log-supermodular in  $(x, \theta)$ , and Property SCB in the proof of Theorem 2 in the main text. The third equality follows from the fact that  $\pi(\theta) = 0$  for  $F$ -almost all  $\theta \in \left( \theta_\sigma^\# - \delta, \theta_\sigma^\# \right)$ . The last equality follows from the definition of the function  $U_\sigma^\Gamma \left( x_\sigma^* \left( \theta_\sigma^\# - \delta \right), 1 | x_\sigma^* \left( \theta_\sigma^\# - \delta \right) \right)$ . Hence,  $U_\sigma^\Gamma \left( x_\sigma^* \left( \theta_\sigma^\# - \delta \right), 1 | x_\sigma^* \left( \theta_\sigma^\# - \delta \right) \right) < 0$ , thus contradicting the assumption that  $\Gamma \in \mathbb{G}_\sigma$ . This completes the proof of Lemma S3-C. ■

**Lemma S3-D.** *For any  $\varepsilon > 0$ , there exists  $\sigma^\#(\varepsilon) \in (0, \hat{\sigma})$  such that, for any  $\sigma \in (0, \sigma^\#(\varepsilon))$ , and any policy  $\Gamma = (\{0, 1\}, \pi) \in \mathbb{G}_\sigma$  for which there exists  $(\theta', \theta'') \in D^\Gamma$  such that (a)  $|\theta'' - \theta'| > \varepsilon$  and (b)  $\pi(\theta) = 0$  for all  $\theta \in (\theta', \theta'')$ , there exists another policy  $\Gamma^\# = (\{0, 1\}, \pi^\#) \in \mathbb{G}_\sigma$ , with  $M(\Gamma^\#) \leq M(\Gamma)$ , such that, in the auxiliary game  $G_\sigma$ , the probability of default under  $\Gamma^\#$  is strictly smaller than under  $\Gamma$ .*

**Proof of Lemma S3-D.** For any  $\theta \in (0, 1)$ ,  $\lim_{\sigma \rightarrow 0^+} x_\sigma^*(\theta) \equiv x_{0^+}^*(\theta) = \theta$ . Furthermore, for any  $\varepsilon \in (0, \min\{\theta^{MS}, 1 - \theta^{MS}\})$ , the function  $x_{0^+}^* : \left[ \frac{\varepsilon}{4}, 1 - \frac{\varepsilon}{4} \right] \rightarrow \mathbb{R}$  is uniformly continuous. Hence, for any  $\delta < \varepsilon/4$ , there exists  $\tilde{\sigma}(\delta) > 0$  such that, for any  $\sigma \in (0, \tilde{\sigma}(\delta))$ , and any  $\theta \in \left[ \frac{\varepsilon}{4}, 1 - \frac{\varepsilon}{4} \right]$ , we have that  $|x_\sigma^*(\theta) - \theta| \leq \delta$ .<sup>9</sup> In turn, this implies that, for any  $\varepsilon > 0$  small, there exists  $\sigma^\#(\varepsilon) \in (0, \hat{\sigma}]$  such that, for any  $\sigma \in (0, \sigma^\#(\varepsilon))$ , and any  $(\theta', \theta'') \in D^\Gamma$  such that  $|\theta'' - \theta'| > \varepsilon$ , we have that, for any  $\theta \in [\theta'', 1 - \frac{\varepsilon}{4}]$ ,  $|\theta - x_\sigma^*(\theta)| < |(\theta' + \theta'')/2 - x_\sigma^*(\theta)|$ . Likewise, for any  $\theta \in [\varepsilon/4, \theta']$ , and any  $\hat{\theta} \geq \theta''$ , we have that  $|\theta - x_\sigma^*(\theta)| < |x_\sigma^*(\theta) - \hat{\theta}|$  when

<sup>9</sup>The proof for the existence of a sequence  $\{x_{\sigma_n}^*(\cdot)\}_n$  with domain  $\left[ \frac{\varepsilon}{4}, 1 - \frac{\varepsilon}{4} \right]$  converging uniformly to its limit function  $x_{0^+}^*(\cdot)$  follows from the same arguments that establish the uniform convergence of  $\{H_{\sigma_n}(\cdot)\}_n$  to  $H_{0^+}(\cdot)$  in Step 1.

$\sigma \in (0, \sigma^\#(\varepsilon)]$ .

Next, pick any policy  $\Gamma = (\{0, 1\}, \pi) \in \mathbb{G}_\sigma$  for which there exists  $d \equiv (\theta', \theta'') \in D^\Gamma$  such that (a)  $|\theta'' - \theta'| > \varepsilon$  and (b)  $\pi(\theta) = 0$  for all  $\theta \in (\theta', \theta'')$ . If  $\min_{\theta \in [0, 1]} U_\sigma^\Gamma(x_\sigma^*(\theta), 1|x_\sigma^*(\theta)) > 0$ , the result follows directly from Lemma S3-B. Thus assume that  $\min_{\theta \in [0, 1]} U_\sigma^\Gamma(x_\sigma^*(\theta), 1|x_\sigma^*(\theta)) = 0$ .

Suppose that  $\min_{\theta \in [\theta', 1]} U_\sigma^\Gamma(x_\sigma^*(\theta), 0|x_\sigma^*(\theta)) > 0$ . By Lemma S3-C,  $U_\sigma^\Gamma(x_\sigma^*(\theta), 1|x_\sigma^*(\theta)) > 0$  for all  $\theta \in (\theta', \theta'')$ . Hence,  $\min_{\theta \in [\theta', 1]} U_\sigma^\Gamma(x_\sigma^*(\theta), 1|x_\sigma^*(\theta)) > 0$ .

Below we show that, starting from  $\Gamma$ , we can then construct a policy  $\Gamma^\eta \in \mathbb{G}_\sigma$ , with  $M(\Gamma^\eta) \leq M(\Gamma)$  such that, when agents play as in  $G_\sigma$ , the probability of default under  $\Gamma^\eta$  is strictly smaller than under  $\Gamma$ .  $\Gamma^\eta$  is obtained from  $\Gamma$  by giving a pass grade to a positive-measure interval of types in the middle of  $(\theta', \theta'')$ . Formally, take  $\eta \in (0, (\theta'' - \theta')/2)$  and let  $\Gamma^\eta = (\{0, 1\}, \pi^\eta)$  be the policy whose rule  $\pi^\eta$  is given by (a)  $\pi^\eta(\theta) = \pi(\theta)$  for all  $\theta \notin [(\theta' + \theta'')/2, (\theta' + \theta'')/2 + \eta]$ , and (b)  $\pi^\eta(\theta) = 1$  for all  $\theta \in [(\theta' + \theta'')/2, (\theta' + \theta'')/2 + \eta]$ . Below we show that  $U^{\Gamma^\eta}(x_\sigma^*(\theta), 1|x_\sigma^*(\theta)) \geq 0$  for all  $\theta \in [0, 1]$ . To see this, let  $\Theta^1 \equiv \{\theta \in \Theta : \pi(\theta) = 1\}$  be the collection of fundamentals receiving a pass grade under the original policy  $\Gamma$ . For any  $\theta \in [0, \theta']$ , and any  $x$ ,

$$\begin{aligned} \Lambda_\sigma^{\Gamma^\eta}(\theta|x, 1) &= Pr\{\tilde{\theta} \leq \theta|x, \tilde{\theta} \in (\Theta^1 \cup [(\theta' + \theta'')/2, (\theta' + \theta'')/2 + \eta])\} \\ &= \frac{Pr\{\tilde{\theta} \leq \theta \wedge \tilde{\theta} \in \Theta^1|x\}}{Pr\{\tilde{\theta} \in (\Theta^1 \cup [(\theta' + \theta'')/2, (\theta' + \theta'')/2 + \eta])|x\}} \leq Pr\{\tilde{\theta} \leq \theta|x, \tilde{\theta} \in \Theta^1\} = \Lambda_\sigma^\Gamma(\theta|x, 1). \end{aligned}$$

The first equality follows from the fact that, under  $\Gamma^\eta$ , the signal  $s = 1$  carries the same information as the announcement that  $\tilde{\theta} \in (\Theta^1 \cup [(\theta' + \theta'')/2, (\theta' + \theta'')/2 + \eta])$ . The inequality follows from the fact that  $Pr\{\tilde{\theta} \in (\Theta^1 \cup [(\theta' + \theta'')/2, (\theta' + \theta'')/2 + \eta])|x\} > Pr\{\tilde{\theta} \in \Theta^1|x\}$ . The last equality follows from fact that, under the original policy  $\Gamma$ , the signal  $s = 1$  carries the same information as the announcement that  $\tilde{\theta} \in \Theta^1$ .

Given the above inequality, and the fact that,  $b < 0 < g$ , we then have that, for any  $\theta \in [0, \theta']$ ,

$$\begin{aligned} U_\sigma^\Gamma(x_\sigma^*(\theta), 1|x_\sigma^*(\theta)) &= b \cdot \Lambda_\sigma^\Gamma(\theta|x_\sigma^*(\theta), 1) + g \cdot [1 - \Lambda_\sigma^\Gamma(\theta|x_\sigma^*(\theta), 1)] \\ &\leq b \cdot \Lambda_\sigma^{\Gamma^\eta}(\theta|x_\sigma^*(\theta), 1) + g \cdot [1 - \Lambda_\sigma^{\Gamma^\eta}(\theta|x_\sigma^*(\theta), 1)] = U_\sigma^{\Gamma^\eta}(x_\sigma^*(\theta), 1|x_\sigma^*(\theta)). \end{aligned}$$

Hence  $U^{\Gamma^\eta}(x_\sigma^*(\theta), 1|x_\sigma^*(\theta)) \geq 0$ , all  $\theta \leq \theta'$ . That  $\min_{\theta \in [\theta', 1]} U_\sigma^\Gamma(x_\sigma^*(\theta), 1|x_\sigma^*(\theta)) > 0$ , along with the continuity of  $U_\sigma^{\Gamma^\eta}(x_\sigma^*(\theta), 1|x_\sigma^*(\theta))$  in  $\eta$  implies that  $\min_{\theta \in [0, 1]} U_\sigma^{\Gamma^\eta}(x_\sigma^*(\theta), 1|x_\sigma^*(\theta)) \geq 0$  for  $\eta$  small. Hence  $\Gamma^\eta \in \mathbb{G}_\sigma$ .

Next, consider the more interesting case in which  $\min_{\theta \in [\theta'', 1]} U_{\sigma}^{\Gamma}(x_{\sigma}^*(\theta), 0|x_{\sigma}^*(\theta)) = 0$ . Let  $\theta_{\sigma}^{\#} \equiv \inf \{ \theta \geq \theta'' : U_{\sigma}^{\Gamma}(x_{\sigma}^*(\theta), 1|x_{\sigma}^*(\theta)) = 0 \}$ . An implication of Lemma S3-C is that that  $\theta_{\sigma}^{\#} > \theta''$ . Also let  $(\theta''', \theta'''' ) \subset [0, 1]$  be the first interval to the immediate right of  $(\theta', \theta'')$  such that  $\pi(\theta) = 1$  for all  $\theta \in (\theta''', \theta'''' )$  and let  $\hat{\theta} = \min \{ \theta'''' , \theta_{\sigma}^{\#} \}$ .<sup>10</sup>

Now, pick  $\xi > 0$  small and let  $\delta(\xi)$  be implicitly defined by

$$F((\theta' + \theta'')/2 + \xi) - F((\theta' + \theta'')/2) = F((\theta''' + \hat{\theta})/2 + \delta(\xi)) - F((\theta''' + \hat{\theta})/2). \quad (\text{S7})$$

Consider the policy  $\Gamma^{\xi} = (\{0, 1\}, \pi^{\xi})$  defined by (a)  $\pi^{\xi}(\theta) = \pi(\theta)$  for all  $\theta \notin [(\theta' + \theta'')/2, (\theta' + \theta'')/2 + \xi] \cup [(\theta''' + \hat{\theta})/2, (\theta''' + \hat{\theta})/2 + \delta(\xi)]$ , (b)  $\pi^{\xi}(\theta) = 1$  for all  $\theta \in [(\theta' + \theta'')/2, (\theta' + \theta'')/2 + \xi]$ , and (c)  $\pi^{\xi}(\theta) = 0$  for all  $\theta \in [(\theta''' + \hat{\theta})/2, (\theta''' + \hat{\theta})/2 + \delta(\xi)]$ . Below we establish that, when  $\xi > 0$  is small, such a policy is such that  $\min_{\theta \in [0, 1]} U^{\Gamma^{\xi}}(x_{\sigma}^*(\theta), 1|x_{\sigma}^*(\theta)) > 0$  and hence  $\Gamma^{\xi} \in \mathbb{G}_{\sigma}$ . To see this, for any arbitrary policy  $\tilde{\Gamma} = (\{0, 1\}, \tilde{\pi})$ , any  $\theta \in [0, 1]$ , let

$$V_{\sigma}^{\tilde{\Gamma}}(\theta) \equiv U_{\sigma}^{\tilde{\Gamma}}(x_{\sigma}^*(\theta), 1|x_{\sigma}^*(\theta)) p_{\sigma}^{\tilde{\Gamma}}(x_{\sigma}^*(\theta), 1),$$

where, for any  $x$ ,  $p_{\sigma}^{\tilde{\Gamma}}(x, 1) \equiv \int_{\Theta} \tilde{\pi}(\theta) p_{\sigma}(x|\theta) dF(\theta)$ , with  $p_{\sigma}(x|\theta) \equiv \frac{1}{\sigma} \phi((x - \theta)/\sigma)$ .

By definition of  $\theta_{\sigma}^{\#}$ , we must have that, for all  $\theta$ ,  $0 = V_{\sigma}^{\Gamma}(\theta_{\sigma}^{\#}) \leq V_{\sigma}^{\Gamma}(\theta)$ . Next, for any  $\xi > 0$ , define  $\varphi_R(\xi) \equiv \min_{\theta \in [\theta'', 1]} V_{\sigma}^{\Gamma^{\xi}}(\theta)$ . Let  $\bar{u}(\tilde{\theta}, \theta) \equiv g\mathbf{1}\{\tilde{\theta} > \theta\} + b\mathbf{1}\{\tilde{\theta} \leq \theta\}$  and note that, for any  $\theta$ ,

$$V_{\sigma}^{\Gamma^{\xi}}(\theta) = V_{\sigma}^{\Gamma}(\theta) + \int_{(\theta' + \theta'')/2}^{(\theta' + \theta'')/2 + \xi} \bar{u}(\tilde{\theta}, \theta) p_{\sigma}(x_{\sigma}^*(\theta) | \tilde{\theta}) dF(\tilde{\theta}) - \int_{(\theta''' + \hat{\theta})/2}^{(\theta''' + \hat{\theta})/2 + \delta(\xi)} \bar{u}(\tilde{\theta}, \theta) p_{\sigma}(x_{\sigma}^*(\theta) | \tilde{\theta}) dF(\tilde{\theta}).$$

Using the envelope theorem, we have that, for any  $\theta_{\sigma}^{\xi} \in \arg \min_{\theta \in [\theta'', 1]} V_{\sigma}^{\Gamma^{\xi}}(\theta)$ ,

$$\begin{aligned} \varphi'_R(\xi) &= f((\theta' + \theta'')/2 + \xi) \bar{u}((\theta' + \theta'')/2 + \xi, \theta_{\sigma}^{\xi}) p_{\sigma}(x_{\sigma}^*(\theta_{\sigma}^{\xi}) | (\theta' + \theta'')/2 + \xi) \\ &\quad - f((\theta''' + \hat{\theta})/2 + \delta(\xi)) \bar{u}((\theta''' + \hat{\theta})/2 + \delta(\xi), \theta_{\sigma}^{\xi}) p_{\sigma}(x_{\sigma}^*(\theta_{\sigma}^{\xi}) | (\theta''' + \hat{\theta})/2 + \delta(\xi)) \delta'(\xi) \\ &= f((\theta' + \theta'')/2 + \xi) [\bar{u}((\theta' + \theta'')/2 + \xi, \theta_{\sigma}^{\xi}) p_{\sigma}(x_{\sigma}^*(\theta_{\sigma}^{\xi}) | (\theta' + \theta'')/2 + \xi) \\ &\quad - \bar{u}((\theta''' + \hat{\theta})/2 + \delta(\xi), \theta_{\sigma}^{\xi}) p_{\sigma}(x_{\sigma}^*(\theta_{\sigma}^{\xi}) | (\theta''' + \hat{\theta})/2 + \delta(\xi))], \end{aligned}$$

where the second equality uses the implicit function theorem applied to (S7) to obtain that

<sup>10</sup>The existence of such an interval follows from the fact that  $\pi(\theta) = 1$  in a left neighborhood of  $\theta_{\sigma}^{\#}$  by virtue of Lemma S3-C. Also observe that, when  $\theta'' < \theta^{MS}$ , such an interval is adjacent to  $(\theta', \theta'')$  and hence  $\theta''' = \theta''$ .

$\delta'(\xi) = f((\theta' + \theta'')/2 + \xi) / f((\theta''' + \hat{\theta})/2 + \delta(\xi))$ . As a consequence,

$$\begin{aligned} \lim_{\xi \rightarrow 0^+} \varphi'_R(\xi) &= f((\theta' + \theta'')/2) [\bar{u}((\theta' + \theta'')/2, \theta_{\sigma}^{\#}) p_{\sigma}(x_{\sigma}^*(\theta_{\sigma}^{\#}) | (\theta' + \theta'')/2) \\ &\quad - \bar{u}((\theta''' + \hat{\theta})/2, \theta_{\sigma}^{\#}) p_{\sigma}(x_{\sigma}^*(\theta_{\sigma}^{\#}) | (\theta''' + \hat{\theta})/2)]. \end{aligned} \quad (\text{S8})$$

That  $\sigma < \sigma^{\#}(\varepsilon)$  implies that  $|x_{\sigma}^*(\theta_{\sigma}^{\#}) - (\theta''' + \hat{\theta})/2| < |x_{\sigma}^*(\theta_{\sigma}^{\#}) - (\theta' + \theta'')/2|$ . That  $p_{\sigma}(x|\theta)$  is single-peaked in turn implies that  $p_{\sigma}(x_{\sigma}^*(\theta_{\sigma}^{\#}) | (\theta' + \theta'')/2) < p_{\sigma}(x_{\sigma}^*(\theta_{\sigma}^{\#}) | (\theta''' + \hat{\theta})/2)$  and hence that

$$\begin{aligned} &\bar{u}((\theta' + \theta'')/2, \theta_{\sigma}^{\#}) p_{\sigma}(x_{\sigma}^*(\theta_{\sigma}^{\#}) | (\theta' + \theta'')/2) - \bar{u}((\theta''' + \hat{\theta})/2, \theta_{\sigma}^{\#}) p_{\sigma}(x_{\sigma}^*(\theta_{\sigma}^{\#}) | (\theta''' + \hat{\theta})/2) \\ &= b \times \left( p_{\sigma}(x_{\sigma}^*(\theta_{\sigma}^{\#}) | (\theta' + \theta'')/2) - p_{\sigma}(x_{\sigma}^*(\theta_{\sigma}^{\#}) | (\theta''' + \hat{\theta})/2) \right) > 0. \end{aligned}$$

Thus,  $\lim_{\xi \rightarrow 0^+} \varphi'_R(\xi) > 0$ . By continuity of  $U_{\sigma}^{\Gamma^{\xi}}(x_{\sigma}^*(\theta), 1|x_{\sigma}^*(\theta))$  in  $\xi$ , we then have that, for  $\xi > 0$  small,  $\min_{\theta \in [\theta'', 1]} U_{\sigma}^{\Gamma^{\xi}}(x_{\sigma}^*(\theta), 1|x_{\sigma}^*(\theta)) > 0$ .

Next, we prove that, under the policy  $\Gamma^{\xi}$ ,  $\min_{\theta \in [0, \theta'']} U_{\sigma}^{\Gamma^{\xi}}(x_{\sigma}^*(\theta), 1|x_{\sigma}^*(\theta)) > 0$ . For any  $\xi > 0$ , define  $\varphi_L(\xi) \equiv \min_{\theta \in [0, \theta'] } V_{\sigma}^{\Gamma^{\xi}}(\theta)$ . Arguments similar to those used above to compute  $\lim_{\xi \rightarrow 0^+} \varphi'_R(\xi)$  imply that, for any  $\theta_{\sigma}^{\#\#} \in \arg \min_{\theta \in [0, \theta'] } V_{\sigma}^{\Gamma}(\theta)$ , when  $\sigma \leq \sigma^{\#}(\varepsilon)$ ,

$$\begin{aligned} \lim_{\xi \rightarrow 0^+} \varphi'_L(\xi) &= f((\theta' + \theta'')/2) [\bar{u}((\theta' + \theta'')/2, \theta_{\sigma}^{\#\#}) p_{\sigma}(x_{\sigma}^*(\theta_{\sigma}^{\#\#}) | (\theta' + \theta'')/2) \\ &\quad - \bar{u}((\theta''' + \hat{\theta})/2, \theta_{\sigma}^{\#\#}) p_{\sigma}(x_{\sigma}^*(\theta_{\sigma}^{\#\#}) | (\theta''' + \hat{\theta})/2)] \\ &= f((\theta' + \theta'')/2) g \left[ p_{\sigma}(x_{\sigma}^*(\theta_{\sigma}^{\#\#}) | (\theta' + \theta'')/2) - p_{\sigma}(x_{\sigma}^*(\theta_{\sigma}^{\#\#}) | (\theta''' + \hat{\theta})/2) \right] > 0. \end{aligned}$$

The first equality follows from steps analogous to those used to establish (S8). The second equality follows from the fact that, by assumption  $\theta_{\sigma}^{\#\#} \leq \theta'$ . The inequality is a consequence of the fact that, for  $\sigma \leq \sigma^{\#}(\varepsilon)$ ,  $|x_{\sigma}^*(\theta_{\sigma}^{\#\#}) - (\theta' + \theta'')/2| < |x_{\sigma}^*(\theta_{\sigma}^{\#\#}) - (\theta''' + \hat{\theta})/2|$ , which, together with the fact that the noise distribution is single-peaked, implies that

$$p_{\sigma}(x_{\sigma}^*(\theta_{\sigma}^{\#\#}) | (\theta' + \theta'')/2) > p_{\sigma}(x_{\sigma}^*(\theta_{\sigma}^{\#\#}) | (\theta''' + \hat{\theta})/2).$$

Hence, for  $\xi > 0$  small,  $U_{\sigma}^{\Gamma^{\xi}}(x_{\sigma}^*(\theta), 1|x_{\sigma}^*(\theta)) > 0$  for all  $\theta \in [0, \theta']$ . Furthermore, by Lemma S3-C,  $U_{\sigma}^{\Gamma}(x_{\sigma}^*(\theta), 1|x_{\sigma}^*(\theta)) > 0$  for all  $\theta \in (\theta', \theta'']$ . Hence, provided that  $\xi$  is small, the continuity of  $U_{\sigma}^{\Gamma^{\xi}}(x_{\sigma}^*(\theta), 1|x_{\sigma}^*(\theta))$  in  $\xi$  implies that  $U_{\sigma}^{\Gamma^{\xi}}(x_{\sigma}^*(\theta), 1|x_{\sigma}^*(\theta)) > 0$  also for  $\theta \in (\theta', \theta'']$ . Combining all the properties above, we thus conclude that, for  $\xi > 0$  small,  $\min_{\theta \in [0, 1]} U_{\sigma}^{\Gamma^{\xi}}(x_{\sigma}^*(\theta), 1|x_{\sigma}^*(\theta)) > 0$ . Hence  $\Gamma^{\xi} \in \mathbb{G}_{\sigma}$ .

By construction,  $M(\Gamma^{\xi}) < M(\Gamma)$ . Furthermore, when agents play according to  $G_{\sigma}$ , the



probability of default under  $\Gamma^\xi$  is the same as under  $\Gamma$ . Lemma S3-B then implies that, starting from  $\Gamma^\xi$ , one can construct a policy  $\Gamma^\# \in \mathbb{G}_\sigma$ , close to  $\Gamma^\xi$  in the  $L_1$  norm, such that (1)  $M(\Gamma^\#) \leq M(\Gamma^\xi)$  and (2), when agents play according to  $G_\sigma$ , the probability of default under  $\Gamma^\#$  is strictly smaller than under  $\Gamma$ . This completes the proof of Lemma S3-D. ■

**Step 3.** Steps 1 and 2 imply that there exists a function  $\bar{\sigma} : (0, \min\{\theta^{MS}, 1 - \theta^{MS}\}) \rightarrow \mathbb{R}_{++}$ , with  $\bar{\sigma}(\varepsilon) \leq \min\{\sigma(\varepsilon), \sigma^\#(\varepsilon)\}$  for all  $\varepsilon \in (0, \min\{\theta^{MS}, 1 - \theta^{MS}\})$  and with  $\bar{\sigma}(\varepsilon) \rightarrow 0^+$  as  $\varepsilon \rightarrow 0^+$ , such that the following is true: For any  $\varepsilon \in (0, \min\{\theta^{MS}, 1 - \theta^{MS}\})$ , any  $\sigma \in (0, \bar{\sigma}(\varepsilon)]$ , and any policy  $\Gamma = (\{0, 1\}, \pi) \in \mathbb{G}_\sigma$  with  $M(\Gamma) > \varepsilon$ , there exists another policy  $\Gamma' = (\{0, 1\}, \pi') \in \mathbb{G}_\sigma$  with  $M(\Gamma') \leq \varepsilon$  such that, when the agents play as in the auxiliary game  $G_\sigma$ , the probability of default under  $\Gamma'$  is strictly smaller than under  $\Gamma$ .<sup>11</sup>

Furthermore, the arguments establishing Lemma S3-D reveal that the policy  $\Gamma'$  can be constructed so that  $U_\sigma^{\Gamma'}(x, 1|x) > 0$  for all  $x$ . The policy  $\Gamma'$  thus satisfies PCP also when agents play according to MARP. The claim in the Example then follows by taking  $\Gamma^* = \Gamma'$  with  $\Gamma'$  satisfying the above properties.

**Step 4.** We now complete the proof by showing how to construct the function  $\mathcal{E}$  in the example. Let  $(\varepsilon_n)$  be a non-increasing sequence satisfying  $\lim_{n \rightarrow \infty} \varepsilon_n = 0$ . For each  $n \in \mathbb{N}$ , then let  $\sigma_n = \bar{\sigma}(\varepsilon_n)$ , with the function  $\bar{\sigma}(\cdot)$  as defined in Step 3. The results in Steps 1-3 above imply that, given  $(\varepsilon_n, \sigma_n)$ , there exist strictly decreasing subsequences  $(\tilde{\varepsilon}_n)$  and  $(\tilde{\sigma}_n)$  satisfying  $\lim_{n \rightarrow \infty} \tilde{\varepsilon}_n = \lim_{n \rightarrow \infty} \tilde{\sigma}_n = 0$  such that, for any  $n \in \mathbb{N}$ , the conclusions in Step 3 hold for  $\varepsilon = \tilde{\varepsilon}_n$  and  $\bar{\sigma}(\varepsilon_n) = \tilde{\sigma}_n$ . Then let  $\bar{\sigma} = \tilde{\sigma}_0 > 0$  and  $\mathcal{E} : (0, \bar{\sigma}] \rightarrow \mathbb{R}_+$  be the function defined by  $\mathcal{E}(\sigma) = \varepsilon_n$  for all  $\sigma \in (\sigma_{n+1}, \sigma_n]$ . The result in the example then follows from Steps 1-3, by letting  $\mathcal{E}(\cdot)$  be the function so constructed. Q.E.D.

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<sup>11</sup>Observe that the thresholds  $\sigma(\varepsilon)$  and  $\sigma^\#(\varepsilon)$  identified in Steps 1 and 2 above are invariant to the initial policy  $\Gamma$ . The same arguments used to arrive at a policy  $\Gamma^\#$  with mesh  $M(\Gamma^\#) < M(\Gamma)$  can then be iterated till one arrives at a policy  $\Gamma'$  with mesh  $M(\Gamma') \leq \varepsilon$ .