# Hide or Surprise?

# Persuasion without Common-Support Priors

Simone Galperti\*

UCSD

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#### Abstract

Often persuaders have a richer understanding of the world than their audience. This paper models such situations by letting persuader and audience have prior beliefs with different supports. This asymmetry adds unexplored aspects to the persuasion problem: Persuaders can hide their superior knowledge or surprise their audience with unexpected information; After surprises Bayes' rule cannot describe how the audience responds. The paper examines persuaders' incentives to hide and surprise and their resulting communication strategies. It also shows under which conditions they surprise their audience as well as hide some information, and which information they hide.

KEYWORDS: persuasion, information control, common support, hiding information, surprise, non-Bayesian updating, concavification.

JEL CLASSIFICATION: D82, D83, K41, M30.

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# 1 Introduction

Persuasion plays an important role across economic activities and beyond.<sup>1</sup> In many situations, the persuader has a superior understanding of the subject matter than her audience. This asymmetry can be better information: a more precise assessment of the likelihood of the possible cases. But more often it is better knowledge: a richer, broader, or more accurate theory of what is possible. Examples include experts advising policymakers, physicians guiding patients, parents teaching young children, and sellers of new products advertising to potential consumers. Due to her superior knowledge, the persuader may view as possible things her audience ignores, deems impossible, or is unaware of.

This asymmetry between persuader and audience adds unexplored aspects to the persuasion problem, which this paper aims to investigate. First of all, the persuader (Sender, she) understands that part of the information she can convey is completely unexpected for her audience (Receiver, he). This raises new questions: How will Receiver react to unexpected information? Given this, should Sender hide such information (if possible) or disclose it, thus surprising Receiver? What are the pros and cons of hiding information and surprising Receiver? Can Sender combine hiding and surprising to her best advantage? How? The paper answers these questions for general persuasion games à la Kamenica and Gentzkow (2011), in which the asymmetry between Sender's and Receiver's 'theory of the world' is modeled by letting their subjective priors over an arbitrary state space have different supports.

To illustrate the analysis and ease the comparison with the standard case of commonsupport priors, it is helpful to revisit the court example in Kamenica and Gentzkow (2011).<sup>2</sup> A lawyer (Sender) is assigned by her company to a new client who claims that a third party caused him some damage. The lawyer has to convince the judge (Receiver) to order that the defendant refund her client. However, the actual value of the damage (the state of the world) remains to be determined. Based on existing evidence, she initially believes that it can amount to 0, 1, 2, or 3 (thousands/millions of) dollars with probabilities 0.35, 0.4, 0.15, and 0.1. She also knows that the judge thinks, based on a wrong interpretation of the evidence, that the damage is either \$0 or \$2 with prior probabilities 0.7 and 0.3.<sup>3</sup> The judge's goal is to match refund to actual damage: in

<sup>&</sup>lt;sup>1</sup>See, e.g., McCloskey and Klamer (1995).

<sup>&</sup>lt;sup>2</sup>This example is solved in Section 5.1.

<sup>&</sup>lt;sup>3</sup>Note that conditional on the event  $\{\$0,\$2\}$  the lawyer and the judge have the same belief. To be clear, the case in which both *priors* are (0.7,0,0.3,0) corresponds to Kamenica and Gentzkow's (2011) example.

each state, he gets payoff 1 from doing so and zero otherwise. By contrast, the lawyer maximizes the expected refund.<sup>4</sup> To do so, she decides to conduct an investigation to produce further evidence on the damage, which she will have to fully disclose as per due-process law. Formally, an investigation is a family  $\pi$  of distributions—one for each state—specifying the probabilities of observing different signals (or evidence). In particular, if the investigation turns out to support, say, states \$1 or \$3, she cannot hide them by simply remaining silent.

The priors' different supports immediately raise a conceptual issue: how does the judge respond to unexpected evidence? If an investigation reveals the true damage, he may observe signals to which ex ante he assigns zero probability. To describe his response to such signals we cannot directly invoke Bayes' rule, the backbone of the entire literature on persuasion (see below). However, allowing the judge to have any posterior belief seems unsatisfactory, for instance simply because the observed signal may rule out some states. It seems also reasonable that his posteriors satisfy some regularity and predictability.

This paper considers different models of how Receiver responds to unexpected information. The first is an application of Ortoleva's (2012) axiomatic model of "paradigm change." The second assumes that Receiver is endowed with a lexicographic belief system (LBS), whose first element corresponds to his prior (e.g., (0.7,0,0.3,0) for the judge). In both models, after expected signals Receiver updates his initial prior using Bayes' rule, as usual. After unexpected signals, however, he first looks for a new prior (or theory) that accounts for the observed evidence—following different procedures depending on the model. Once he finds his new prior, Receiver again updates it using Bayes' rule. In our example, let the judge's LBS have two priors with the secondary one equal to the lawyer's prior. To fix ideas, we shall say that expected signals confirm Receiver's prior and unexpected signals disprove it, causing a surprise.

With these assumptions, we can analyze a key dilemma Sender faces in our setting. Should the lawyer confirm the judge's theory that the damage is \$0 or \$2, or disprove it? Always producing evidence he expects will confirm his theory and lead at most to a \$2 refund. It is intuitive that if the damage is \$3, the lawyer would like the judge to know. But what about the \$1 damage?

The solution to this dilemma relies on a general characterization of the joint distributions over posterior beliefs that Sender can achieve with her signal devices (one of the

<sup>&</sup>lt;sup>4</sup>In the general model, Sender's utility function can also depend on the state.

<sup>&</sup>lt;sup>5</sup>This second class of models is related to the lexicographic sequences of hypotheses in Kreps and Wilson (1982) (see Section 2).

<sup>&</sup>lt;sup>6</sup>This case also corresponds to the simplest version of Ortoleva's (2012) model (see Section 4.1).

paper's main results). This is key because Receiver's posteriors determine his actions and Sender's posteriors her resulting payoffs. As usual, the marginal distribution over Sender's posteriors satisfies the property that her expected posterior equals her prior. Here, however, the relationship between her posteriors and Receiver's has specific properties that are absent in standard settings with common-support priors. In such settings, posteriors are in a one-to-one relationship and each varies continuously as a function of the other; moreover, Receiver's posterior can be any belief over the common state space.<sup>7</sup>

These properties no longer hold without common-support priors. First, each Sender's posterior corresponds to a unique Receiver's posterior, but not vice versa. If an investigation yields signal x when the damage is either \$1 or \$2 and y when it is either \$2 or \$3, the judge will always conclude that it is \$2, but the lawyer learns about state \$1 from x and state \$3 from y. More generally, the paper shows that after expected signals Receiver's posterior depends only on Sender's posterior conditional on the states he initially deemed possible (the event  $\{\$0,\$2\}$  in the example). As a result, Sender can have any posterior over the remaining states and learn about them without affecting Receiver's behavior. Since Receiver continues to view these states as impossible, we conclude that his reluctance to abandon his theory in light of inconclusive evidence can allow Sender to 'hide' states inconsistent with that theory by pooling them with signals on states consistent with it.

Second, Receiver's posterior varies discontinuously in Sender's as we shift from expected to unexpected signals. This discontinuity captures in a nice and formal way the idea of surprise: A small change in information moves slightly the posterior of Sender, who expected it, but drastically that of Receiver, who did not.<sup>8</sup> Third, the Receiver's posteriors that Sender can induce are only a relatively small subset of all beliefs over the states she deems possible. The judge will never assign positive probability both to elements in {\$0,\$2} and to elements in {\$1,\$3}: if the lawyer confirms his theory, he continues to think only \$0 and \$2 are possible; to disprove it, she must show that the damage is neither \$0 nor \$2.

A crucial step in characterizing how Sender communicates is identifying the opportunity cost of surprising. This cost is determined by Sender's highest payoff when she hides a state inconsistent with Receiver's theory. Her optimal hiding strategy has two parts. First, she hides each state using a different signal, which does not entirely rule out all states consistent with Receiver's theory. Second, she manipulates his posterior—which continues to assign zero probability to the hidden state—so that she achieves the

<sup>&</sup>lt;sup>7</sup>See Kamenica and Gentzkow (2011) and Alonso and Câmara (2013).

<sup>&</sup>lt;sup>8</sup>This is different from the notion of surprise in Ely et al. (2013).

best payoff for that state among all actions Receiver would choose across all posteriors consistent with his theory. Concretely, the lawyer hides the \$1 and \$3 damages with an investigation that maps each to a different signal, which can also arise with a tiny probability in states \$0 and \$2. And she allocates this probability between \$0 and \$2 so that the judge orders a \$2 refund. This sophisticated strategy defines the value of hiding states (\$2 in the example) and hence the opportunity cost of not doing so.

Based on these observations, the paper provides a simple necessary and sufficient condition for Sender to surprise Receiver with positive probability. There must exist *some* posterior of Sender at which, given Receiver's response to the surprise, her expected payoff exceeds her expected opportunity cost of surprising. Our example satisfies this condition for the posterior assigning probability 1 to state \$3, since the opportunity cost here is \$2.

To characterize how Sender communicates, the paper extends and modifies the "concavification method" now common in the persuasion literature (see below). Sender communicates as if she could divide her problem into two fictitious steps as follows. She first learns only whether the true state is consistent with Receiver's theory—the event  $\{\$0,\$2\}$  or  $\{\$1,\$3\}$  in the example. If it is, her interim belief has the same support as Receiver's prior, and so she adopts the optimal communication strategy for settings with common-support priors (Kamenica and Gentzkow (2011); Alonso and Câmara (2013)). If not, she combines hiding states and surprising. The optimal combination can be derived from the value, taken at Sender's interim belief (i.e., given event  $\{\$1,\$3\}$ ), by the concavification of the maximum between two functions: for each posterior, one measures her expected payoff from surprising, the other its opportunity cost. This procedure also delivers the ex-ante probability of surprises, how and which states Sender hides, and conditions for her to never hide any state.

In our example, the lawyer designs an investigation with four signals: The first arises if the defendant is innocent, revealing it; The second arises in states \$0 and \$2 and induces the judge to assign them equal probability; The third arises in states \$1 and \$2 and induces her to assign probability arbitrarily close to one to \$1, while the judge assigns probability one to \$2; The fourth arises in states \$1 and \$3 and induces the judge to assign them equal probability. So, even though the lawyer can always hide the most likely \$1 damage with the better \$2 one, she reveals it with positive probability! In so doing, she makes the judge less sure when he orders a \$3 refund, but she increases its overall probability. Overall, the resulting expected refund is \$1.8. For comparison, the

<sup>&</sup>lt;sup>9</sup>These first two signals are reminiscent of the optimal investigation in Kamenica and Gentzkow's (2011) setting.

expected refund can be at most \$1.6 if the investigation always hides states \$1 and \$3, and \$1.5 if it never hides states. So, by optimally combining hiding and surprising, the lawyer increases her expected payoff by 12.5% and 20% respectively.

Section 5.2 applies the main results to the classic model in which Sender's and Receiver's payoffs are quadratic loss functions and their ideal actions differ for each state. Since Crawford and Sobel (1982), this model has been used to study communication in many contexts. Section 5.3 shows that Sender need not benefit from her superior knowledge: She might be strictly better off if initially Receiver shared her theory of the world, rather than viewing some states as impossible. This result is perhaps unsurprising, but the paper sheds light on its origin: As noted, Sender can induce Receiver to entertain only a much smaller set of posteriors when he initially deems some states impossible. In this case, Sender would like to "persuade" Receiver to switch theory before communication occurs, but no signal device (as defined here) allows her to do so. Finally, Section 6.3 shows that, in general, it is not correct to view the case of different-support priors as a limit of the case of common-support priors in which the probability that Receiver initially assigns to some states goes to zero: In the limit, Sender's communication and payoff as well as Receiver's behavior can differ substantively from their respective limits.

# 2 Related Literature

This paper contributes to the literature on games of persuasion and information control. It is the first to study games in which Sender and Receiver disagree on their subjective theory of the possible states of the world. A common question addressed in the literature has been if and when Sender benefits from revealing (expected) information. By contrast, one of this paper's main questions is if and when Sender benefits from surprising Receiver with unexpected information, rather than hiding it. This paper differs from the literature in another important aspect. With common-support priors, Bayes' rule always dictates how Receiver responds to information and represents the constraint on what Sender can achieve. As noted, this is no longer true without common-support priors. While moving beyond Bayesian rationality, this paper does not introduce bounded rationality or behavioral features in how Receiver updates his beliefs.<sup>11</sup>

The closest papers in the persuasion literature are the following. In Brocas and Car-

<sup>&</sup>lt;sup>10</sup>Examples include organizational design (Dessein (2002); Alonso et al. (2008)), political economy (Grossman and Helpman (2002)), legal-dispute resolution (Goltsman et al. (2009)), lobbying (Kamenica and Gentzkow (2011)), and financial advising (Morgan and Stocken (2003)).

<sup>&</sup>lt;sup>11</sup>For a study of persuasion with these features see, e.g., Mullainathan et al. (2008).

rillo (2007), Sender has access to a fixed device producing i.i.d. signals about a binary state. She chooses sequentially whether to produce another signal, or stop and let Receiver act. The paper examines Sender's optimal stopping rule and how much she benefits from controlling information. Kamenica and Gentzkow (2011) study persuasion in more general settings with common priors and no restrictions on signal devices. They show that the problem of choosing a device can be conveniently reformulated as choosing a distribution over posteriors subject to only one constraint imposed by Bayesian rationality: the expected posterior must equal the prior. They also find conditions for Sender to benefit from informative communication, characterize its properties, and examine the effects of changing the alignment of Sender's and Receiver's preferences. Within Kamenica and Gentzkow's (2011) settings, Alonso and Câmara (2013) examine how disagreement in priors (with common support) affects Sender's communication, showing that it increases the scope for benefiting from revealing information. They also show that disagreement does not expand the set of feasible distributions over posteriors. As noted, this set is qualitatively very different in the case of different-support priors.<sup>12</sup>

As in the previous literature (Kamenica and Gentzkow (2011), Alonso and Câmara (2013), Ely (2014)), this paper's analysis relies on the 'concavification method' proposed in the seminal work of Aumann and Maschler (1995) on repeated games with incomplete information. However, it modifies this method in several ways to examine Sender's payoff separately when she confirms and disproves Receiver's theory. This method also turns out to be helpful to characterize Sender's behavior despite the possibility that an optimal device may formally not exist.

This paper models responses to unexpected information in ways that are related to several papers in the literature. On the one hand, it borrows Ortoleva's (2012) axiomatic model of "change of paradigm," which provides specific predictions on how Receiver "chooses" a new prior and updates it after surprises. A key difference is that in Ortoleva (2012) the information structure is exogenous; here it is not, which will require some care in applying his model. On the other hand, the use of lexicographic belief systems (LBS's) is inspired by Kreps and Wilson (1982) and Karni and Vierø (2013). In their work on sequential equilibria, Kreps and Wilson (1982) imagine that each player has a system of "hypotheses" on how the game is played, the primary one corresponding to equilibrium play. A player always applies his primary hypothesis on the equilibrium

<sup>&</sup>lt;sup>12</sup>Other papers study problems of information design, both static and dynamic (see, e.g., Forges and Koessler (2008), Rayo and Segal (2010), Horner and Skrzypacz (2010), Ely et al. (2013), Ely (2014)). All these papers assume common priors and address different questions from the present paper.

<sup>&</sup>lt;sup>13</sup>LBS's appear also in Blume et al. (1991a, 1991b), but in these papers they work in a fundamentally different way: the agent always takes into account, though lexicographically, all layers of his LBS.

path. But if an off-path information set is reached—a zero-probability event under the primary hypothesis—the player attempts to apply other hypotheses until one predicts what happened. The player always updates his current hypothesis using Bayes' rule. In Kreps and Wilson (1982), players have a common prior over Nature's moves and the hypotheses are on each others' strategies; here the common-prior assumption is absent, and we can interpret Receiver's hypotheses as being on player Nature.

Karni and Vierø (2013) model how an agent's beliefs evolve when he either discovers that states he considered impossible are actually possible, or becomes aware of new states. They suggest and axiomatize a notion of belief consistency similar to that used in the present paper. A key difference is that here not only Receiver may have to expand his subjective set of possible states and hence form new priors; he also has to update such priors after observing Sender's information.

Finally, this paper is related to the literature on strategic information transmission following Crawford and Sobel (1982), but differs from it in several aspects shared with all previously mentioned papers on persuasion. In that literature Sender learns the state before choosing how to communicate; here he commits to a signal device before the state occurs. This property changes deeply the incentive problems Sender faces and eliminates multiplicity of equilibria. Also, that literature has focussed on settings with common-support priors.

## 3 Model

The primitives of the model are as in Kamenica and Gentzkow (2011), except of course for the assumptions on priors.

There are two agents, called Sender (she) and Receiver (he).  $\Omega$  is a finite set of mutually exclusive states of the world. Sender and Receiver have a common language to describe each state  $\omega \in \Omega$  and agree on the meaning of this description. Each  $\omega$  is an exhaustive description of reality. If  $\omega$  occurs, we say that  $\omega$  is 'true;' otherwise, we say that  $\omega$  is 'false.'

Neither Sender nor Receiver know the true state. Sender has a subjective prior belief  $\sigma$  with support  $\mathcal{S} = \Omega$  and Receiver has a subjective prior belief  $\rho_0$  with support  $\mathcal{R} \subsetneq \Omega$ .<sup>14</sup> Hereafter, let  $\overline{\mathcal{R}} = \mathcal{S} \setminus \mathcal{R}$  and denote the support of  $\sigma$  by supp  $\sigma$ ; this notation applies

<sup>&</sup>lt;sup>14</sup>In Kamenica and Gentzkow (2011),  $S = \mathcal{R}$  and  $\sigma = \rho_0$ . In Alonso and Câmara (2013),  $S = \mathcal{R}$  but  $\sigma$  may differ from  $\rho_0$ . In the present model,  $\mathcal{R}$  may differ from S without  $\mathcal{R} \subsetneq S$ . The case of  $\mathcal{R} \subsetneq S$ , however, comprises all key aspects of the different-support assumption as explained in Section 6.4.

for any other probability distribution. We shall call  $\sigma$  Sender's theory of the world, and  $\rho_0$  Receiver's theory.

Following the literature on persuasion, Sender and Receiver interact as follows. Sender commits to a signal device to provide Receiver with information on the true state, with the goal of steering his behavior. After observing a signal realization, Receiver chooses an action from the compact set A, with |A| > 1, which affects both agents payoffs. As in Kamenica and Gentzkow (2011), a signal device  $\pi$  consists of a family of conditional distributions  $\{\pi(\cdot|\omega)\}_{\omega\in\Omega}$  over a finite set  $X_{\pi}$  of signal realizations: for each  $\omega\in\Omega$ ,  $\pi(\cdot|\omega)\in\Delta(X_{\pi})$  where  $X_{\pi}=\cup_{\omega\in\Omega}\operatorname{supp}\pi(\cdot|\omega)$ . The set of all signal devices is  $\Pi$ . As in Alonso and Câmara (2013), signal devices are "commonly understood:" Sender and Receiver agree on how  $\pi$  generates signals for each  $\omega$ . Hereafter, the term signal will refer to a realization x of a device  $\pi$ ; the term message will refer to any pair  $(x,\pi)$  with  $x \in X_{\pi}$ .

Regarding payoffs, Sender design her device so as to maximize her subjective expected utility with cardinal utility function  $u_S: A \times \Omega \to \mathbb{R}$ . For Receiver, let his utility function be  $u_R: A \times \Omega \to \mathbb{R}$ . Both  $u_S$  and  $u_R$  are continuous in a for every  $\omega$ . Receiver's behavior after every signal will be specified below.<sup>16</sup>

The paper's goal is to examine how Sender communicates with Receiver in this environment.

#### Interpretation of Priors' Different Supports

Many reasons can explain why Sender and Receiver have different priors.<sup>17</sup> A natural—certainly not unique—one is that Sender is an expert with a more complete and accurate understanding of the world. In this case, it is natural to think that her theory is the "correct" one. This is not necessary, however: We can view the entire analysis as from the *subjective* ex-ante perspective of Sender. That is, the paper examines how Sender communicates if she thinks that Receiver has a different theory  $\rho_0$  from hers and responds to signals as described below.

The mathematical property that the supports of Sender's and Receiver's priors differ

<sup>&</sup>lt;sup>15</sup>For an extensive discussion and justification of the commitment assumption, see Kamenica and Gentzkow (2011). Modifying this assumption is beyond the scope of the present paper.

<sup>&</sup>lt;sup>16</sup>The functions  $u_S$  and  $u_R$  should be interpreted as reduced forms in an Anscombe and Aumann's (1963) setting, where subjective beliefs are well defined. Letting  $\mathcal{G} = \{g_a\}_{a \in A}$  be a set of Anscombe and Aumann's acts where A is a set of "labels" and  $\hat{u}_i$  be i's cardinal utility for i = R, S, we have  $u_i(a, \omega) = \hat{u}_i(g_a(\omega))$ .

<sup>&</sup>lt;sup>17</sup>See also the discussion in Morris (1995).

can be interpreted in several ways. For the sake of clarity, most of the paper will focus on one: Receiver is aware of all states in S and views them as well-specified hypotheses, but simply thinks that some states are impossible based on his theory. For instance, the statement "the Earth goes around the Sun" was formally correct and scientists could understand its meaning even before the Copernican Revolution; yet they deemed such a statement impossible according to their theories. In the more recent debate on climate change, deniers and believers in manmade global warming understand each other's theories; yet each side views the other's theory as impossible. Receiver may assign zero probability to some states for different reasons. Some may be exogenous and subjective, like religion in the previous examples. Sender's and Receiver's theories may be logically inconsistent, so that they cannot have the same set of possible predictions. In some cases, Receiver may frame the situation at hand in some narrow way that leads him to ignore some states. Finally, Receiver may be boundedly rational or face high cognitive costs; as a result, he may be able to reason only in terms of simple theories, which again ignore some states.

Before observing any signal, Receiver is confident that his theory is correct in the following sense. If he doubted it and assigned positive probability to another theory which incorporates states outside  $\mathcal{R}$  as possible, then a correct probabilistic description of his actual theory should include these states in its support. Hence, if  $\mathcal{R} \neq \mathcal{S}$ , it means that ex ante Receiver is fully confident that the states outside  $\mathcal{R}$  are impossible—even though he may know that Sender has a different opinion. Nonetheless, of course Receiver will have to abandon his theory if a signal unambiguously proves that the true state is outside  $\mathcal{R}$ .

Another possible interpretation of the model is that Receiver's prior assigns zero probability to some states because he is unaware of them. Though perhaps natural, this interpretation is more delicate and requires careful explanation in the present environment. We shall defer its discussion until Section 6.1.

# 4 Feasible Posteriors

To understand how Sender communicates, we first need to answer this question: which Receiver's beliefs over  $\Omega$  can Sender induce using signal devices in  $\Pi$ ?

This question does not have an immediate answer in our setting. If supp  $\rho_0 = \Omega$ ,

 $<sup>^{18}</sup>$ Ahn and Ergin (2010), for instance, develop a model of framing in which an agent can 'overlook' some events.

we would usually invoke Bayesian rationality and, using Bayes' rule, obtain a unique posterior of Receiver for each message  $(x, \pi)$ . Here, however, Receiver may endogenously observe unexpected messages, evidence to which he assigns zero probability ex ante. In this case Bayes' rule does not apply. Thus one possibility is to say that after surprises Receiver can have any belief; but this approach would be unsatisfactory. We would like his posteriors to feature some regularity and predictability and to take into account properties of  $\pi$  even after unexpected signals. For instance, a posterior should assign zero probability to states that are inconsistent with message  $(x, \pi)$ .

To this end, we shall first consider Ortoleva's (2012) axiomatically founded model of "change of paradigm." Section 4.2 answers the above question within this model. Section 6.2 considers other models of how Receiver responds to signals and shows that the paper's main conclusions continue to hold.

To fix some terminology, hereafter we say that  $(x, \pi)$  confirms  $\rho_0$  if, given  $\rho_0$ , Receiver assigns positive probability to observing x from  $\pi$ ; otherwise,  $(x, \pi)$  disproves  $\rho_0$ . Let  $C_{\pi}$  be the set of signals such that  $(x, \pi)$  confirms  $\rho_0$  and  $D_{\pi}$  the set of signals such that  $(x, \pi)$  disproves  $\rho_0$ :<sup>19</sup>

$$C_{\pi} = \{x \in X_{\pi} : \pi(x|\omega) > 0 \text{ for some } \omega \in \mathcal{R}\}, \text{ and } D_{\pi} = X_{\pi} \setminus C_{\pi}.$$

Note that  $C_{\pi}$  is always nonempty, but  $D_{\pi}$  may be empty.

# 4.1 A Model of Receiver's Response to Information

Ortoleva's (2012) model of paradigm change provides an as if description of how Receiver responds to Sender's messages, based on observable properties (axioms) of such responses. In the model, Receiver has a prior over priors  $\mu \in \Delta(\Delta(\Omega))$  with finite support. Initially, he adopts the prior with the highest likelihood under  $\mu$  as his theory. This is the theory that he thinks is correct, i.e.,  $\rho_0$ . After expected messages, he updates  $\rho_0$  using Bayes' rule. After unexpected messages, by contrast, he first reassesses the likelihood that each theory could be correct and hence resulted in the observed evidence; that is, he updates  $\mu$  obtaining  $\hat{\mu}$ . He then adopts prior  $\hat{\rho}$  with the highest likelihood under  $\hat{\mu}$  and again updates it using Bayes' rule.<sup>20</sup> Formally, let  $\mu$  be such that for every  $\omega \in \Omega$  there exists

<sup>&</sup>lt;sup>19</sup>Both sets depend on  $\rho_0$ , but this dependence is left implicit to simplify notation. Also, recall that  $X_{\pi}$  includes only signals in the support of  $\pi(\cdot|\omega)$  for some  $\omega$ .

<sup>&</sup>lt;sup>20</sup>As it will become clear in Section 4.2, the analysis of the paper would be unchanged if Receiver adopted a prior  $\rho$  not based on the maximum-likelihood criterion, but using any criterion that depends only on the priors over priors  $\mu$  and  $\hat{\mu}$ . For example, Receiver may adopt  $\rho_0$  (resp.  $\hat{\rho}$ ) so as to minimize the expectation under  $\mu$  (resp.  $\hat{\mu}$ ) of some loss function that depends on his choice and the "true"  $\rho$ .

 $\rho \in \operatorname{\mathbf{supp}} \mu$  with  $\rho(\omega) > 0$ . Then, given  $(x, \pi)$ , let the updated prior over priors be

$$\hat{\mu}(\rho|x,\pi) = \frac{\left[\sum_{\omega \in \Omega} \pi(x|\omega)\rho(\omega)\right] \mu(\rho)}{\sum_{\tilde{\rho} \in \mathbf{supp}\,\mu} \left[\sum_{\omega \in \Omega} \pi(x|\omega)\tilde{\rho}(\omega)\right] \mu(\tilde{\rho})} \quad \text{for every } \rho \in \Delta(\Omega).$$
 (1)

Also, let

$$M = \underset{\rho}{\operatorname{arg\,max}} \ \mu(\rho) \quad \text{and} \quad M(x,\pi) = \underset{\rho}{\operatorname{arg\,max}} \ \hat{\mu}(\rho|x,\pi).$$

In general, M and  $M(x,\pi)$  need not be singletons. Indeterminacies in Receiver's 'choice' of a prior, however, would make his behavior ill defined. Following Ortoleva (2012), we endow Receiver with a strict linear order  $\succ$  over  $\Delta(\Omega)$ , which he uses to 'choose' a prior when the maximum-likelihood criterion is inconclusive. For simplicity, assume that M is singleton.

Assumption 1 (A1: Hypothesis-Testing Model (Ortoleva (2012))). Receiver has a prior over priors  $\mu \in \Delta(\Delta(\Omega))$  with finite support and such that for every  $\omega \in \Omega$  there exists  $\rho \in \text{supp }\mu$  with  $\rho(\omega) > 0$  and  $M = \{\rho_0\}$ . Also, (c) if  $x \in C_{\pi}$ , Receiver updates  $\rho_0$  using Bayes' rule; (d) if  $x \in D_{\pi}$ , he updates  $\rho$  using Bayes' rule where  $\rho$  is  $\succ$ -maximal in  $M(x,\pi)$ .

A1 corresponds to Ortoleva's (2012) Hypothesis-Testing Representation with  $\varepsilon = 0$ , which is consistent with our definition of  $C_{\pi}$ . As a simple example, suppose that  $\operatorname{supp} \mu = (\rho_0, \rho_1)$  with  $\mu(\rho_0) > \frac{1}{2}$  and  $\operatorname{supp} \rho_1 = \Omega$ . Then, for  $x \in C_{\pi}$  Receiver uses  $\rho_0$ , but for  $x \in D_{\pi}$  he switches to  $\rho_1$ . In both cases, he updates his prior using Bayes' rule.

A few remarks on part (c) of A1 are in order. If message  $(x, \pi)$  confirms  $\rho_0$ , it gives Receiver no objective reason to doubt his theory; therefore he does not abandon it. A1(c) holds if and only if Receiver's behavior is dynamically consistent (Theorem 1 in Ortoleva (2012)). It is also in the spirit of the notion of Perfect Bayesian Equilibrium and its refinements, which require that Bayes' rule hold whenever possible.<sup>22</sup> It is consistent with the interpretation that, at the outset, Receiver thinks that his theory is correct. It is also consistent with the property that signal devices, per se, contain no information; so Receiver should respond only to the evidence given by signal realizations. Finally, it is consistent with a phenomenon known in the psychology literature as "confirmatory bias:" When presented with inconclusive evidence, people tend to interpret it in favor of their initial hypothesis (see, e.g., Rabin and Schrag (1999) and references therein).<sup>23</sup>

<sup>&</sup>lt;sup>21</sup>Since in our setting Receiver's  $\mu$  is predetermined while the evidence he observes is endogenous, it is logically impossible to construct  $\mu$  so that the maximum-likelihood criterion always gives a unique answer as in Ortoleva (2012).

<sup>&</sup>lt;sup>22</sup>We can always think of a third player, Nature, who chooses  $\omega$  at the beginning of the game and has a constant payoff across final outcomes.

<sup>&</sup>lt;sup>23</sup>Ortoleva's (2012) model allows for the possibility that Receiver views  $(x,\pi)$  as disproving  $\rho_0$  if

#### 4.2 Characterization of Distributions over Posteriors

Several reasons justify examining the distributions over posteriors that signal devices can induce in our setting. Each message leads to a posterior belief of Sender and of Receiver. Receiver's posterior determines his optimal actions. Therefore, Sender ultimately cares about the posteriors that she can feasibly induce Receiver to have. More generally, since the outcome of a signal device depends on the true state, Sender cares about the feasible distributions over posteriors. The literature has characterized in detail such distributions in settings with common-support priors.<sup>24</sup> A similar exercise here will greatly help our understanding of the key differences of settings without common-support priors.

Any  $\pi$  induces a distribution over Sender's posteriors as follows. By Bayes' rule, given  $(x, \pi)$  the probability she assigns to  $\omega$  is given by

$$q(\omega|x,\pi) = \frac{\pi(x|\omega)\sigma(\omega)}{\sum_{\omega' \in S} \pi(x|\omega')\sigma(\omega')}.$$
 (2)

Applying (2) across x's delivers a distribution  $\tau \in \Delta(\Delta(S))$  with finite support, where the probability that  $\tau$  assigns to any  $q \in \Delta(S)$  is

$$\tau(q) = \sum_{\{x: q(\cdot|x,\pi)=q\}} \sum_{\omega \in \mathcal{S}} \pi(x|\omega)\sigma(\omega).$$

As usual, we have  $\sum_{q \in \text{supp } \tau} q\tau(q) = \mathbb{E}_{\tau}[q] = \sigma$  for any  $\tau$  induced by some  $\pi \in \Pi$ . Conversely, for any  $\tau'$  with finite support that satisfies  $\mathbb{E}_{\tau'}[q] = \sigma$ , there exists  $\pi \in \Pi$  that induces  $\tau'$  through Bayes' rule (see, e.g., Kamenica and Gentzkow (2011)).

**Definition 1.** Fix Sender's prior  $\sigma$ . A distribution  $\tau \in \Delta(\Delta(\mathcal{S}))$  is *feasible* if and only if **supp**  $\tau$  is finite and  $\mathbb{E}_{\tau}[q] = \sigma$ . The set of feasible distributions is  $\mathcal{F}_{\sigma}$ .

Definition 1 is equivalent to Kamenica and Gentzkow's (2011) notion of Bayes plausibility.

It is worth mentioning at this point the following result from the literature.

**Proposition 1** (Alonso and Câmara (2013)). Suppose  $\operatorname{supp} \rho_0 = \operatorname{supp} \sigma = \Omega$ . Then, there exists a continuous, one-to-one, onto function from Sender's to Receiver's posterior which depends only on  $\rho_0$  and  $\sigma$  and has continuous inverse.

This result does not hold in the present setting. Indeed, we will show that here we continue to have a function from Sender's to Receiver's posterior which does not depend at all on signal devices, but this function loses all the properties in Proposition 1.

 $<sup>\</sup>sum_{\omega} \pi(x|\omega) \rho_0(\omega) < \varepsilon$  for some  $\varepsilon > 0$ . This violates the usual dynamic-consistency condition which characterizes Bayesian updating. Studying such violations in persuasion games is clearly orthogonal to and beyond this paper's scope—they may arise even if  $\rho_0 = \sigma$ .

<sup>&</sup>lt;sup>24</sup>See Kamenica and Gentzkow (2011) and Alonso and Câmara (2013).

#### Supports of Sender's and Receiver's Posteriors

Imagine that we know Sender's posterior, but not the message  $(x, \pi)$  that led to it. Can we infer whether  $(x, \pi)$  confirmed or disproved Receiver's prior? The answer is yes. The support of  $q(\cdot|x,\pi)$  in (2) reveals whether x belongs to  $C_{\pi}$  or  $D_{\pi}$ . Indeed, supp  $q(\cdot|x,\pi)$  coincides with the set of states that are consistent with  $(x,\pi)$ , defined by

$$\Omega_{\pi}(x) = \{ \omega \in \Omega : \pi(x|\omega) > 0 \}. \tag{3}$$

Therefore,

$$x \in C_{\pi}$$
 if and only if  $\operatorname{supp} q(\cdot | x, \pi) \cap \mathcal{R} \neq \emptyset$ .

Of course, then  $x \in D_{\pi}$  if and only if  $\operatorname{supp} q(\cdot|x,\pi) \subset \overline{\mathcal{R}}$ . Thus the support of Sender's posterior q induced by any message—hence in the support of some  $\tau \in \mathcal{F}_{\sigma}$ —allows us to tell whether  $\rho_0$  has been confirmed or disproved, without even knowing the message that led to q.

This observation suggests to partition the set of Sender's posteriors as follows:

$$\Delta^d = \Delta(\overline{\mathcal{R}})$$
 and  $\Delta^c = \Delta(\mathcal{S}) \setminus \Delta^d$ .

Abusing terminology, we say that q confirms  $\rho_0$  if  $q \in \Delta^c$ , otherwise q disproves  $\rho_0$ .<sup>25</sup> For any  $\tau \in \mathcal{F}_{\sigma}$ , then define

$$D_{\tau} = \{ q \in \operatorname{\mathbf{supp}} \tau : q \in \Delta^d \} \quad \text{and} \quad C_{\tau} = \operatorname{\mathbf{supp}} \tau \setminus D_{\tau}.$$

This leads to the following preliminary properties of Receiver's feasible posteriors. <sup>26</sup>

**Proposition 2.** Given any  $(x,\pi)$ , let  $p(\cdot|x,\pi)$  be Receiver's posterior. Under A1, if  $q(\cdot|x,\pi) \in \Delta^c$ , then  $\operatorname{supp} p(\cdot|x,\pi) \subset \mathcal{R}$ ; if  $q(\cdot|x,\pi) \in \Delta^d$ , then  $\operatorname{supp} p(\cdot|x,\pi) \subset \overline{\mathcal{R}}$ .

Sender cannot make Receiver assign positive probability to both elements in  $\mathcal{R}$  and in  $\overline{\mathcal{R}}$ . If she confirms  $\rho_0$ , he will assign positive probability only to states in  $\mathcal{R}$ ; to disprove  $\rho_0$ , she must show that *all* states in  $\mathcal{R}$  are false. This result highlights a constraint on the posteriors of Receiver that Sender can induce in settings with different-support priors, which does not arise with common-support priors (Proposition 1). The set of Receiver's achievable posteriors is here much smaller than the entire set of possible beliefs  $\Delta(\Omega)$ .

Proposition 2 also restricts the probability Sender assigns ex ante to confirming  $\rho_0$ .

Corollary 1. For  $\tau \in \mathcal{F}_{\sigma}$ , the probability  $\tau^c = \tau(C_{\tau})$  of confirming  $\rho_0$  is at least  $\sigma(\mathcal{R})$ .

<sup>&</sup>lt;sup>25</sup>Since  $\overline{\mathcal{R}} \subseteq \mathcal{S}$ ,  $\Delta^d$  is a strictly lower dimensional subset of  $\Delta(\mathcal{S})$  and  $\Delta^d \cap int(\Delta(\mathcal{S})) = \emptyset$ .  $\Delta^d$  is also a face of  $\Delta(\mathcal{S})$  which lies on its (relative) boundary.

<sup>&</sup>lt;sup>26</sup>All proofs are in Appendix B.

However Sender discloses information, there is always a strictly positive probability that Receiver will not abandon his theory. From Sender's viewpoint, the largest probability of disproving  $\rho_0$  is  $\sigma(\overline{\mathcal{R}}) < 1$  (for instance, when  $\pi$  is fully revealing). By contrast, Sender can always design a  $\pi$  that confirms  $\rho_0$  with probability 1—by ensuring that, for every  $\omega \in \overline{\mathcal{R}}$ ,  $\operatorname{supp} \pi(\cdot|\omega) \subset \operatorname{supp} \pi(\cdot|\omega')$  for some  $\omega' \in \mathcal{R}$ . This feature highlights an asymmetry between confirming and disproving an 'incorrect' theory by producing evidence on its predictions. For instance, if Receiver's theory relied on some flawed logic, then pointing out the flaw might be enough to disprove the theory for sure.

#### Relationship Between Sender's and Receiver's Posteriors

Imagine again that we know Sender's posterior, but not the message that led to it. Can we always recover Receiver's actual posterior? Despite the complexity of Receiver's updating under A1, the answer is yes.

Given any  $(x, \pi)$ , Sender's posterior  $q = q(\cdot|x, \pi)$  tells us which prior Receiver updates. Of course, he updates  $\rho_0$  when q confirms it. If q disproves  $\rho_0$ , even without knowing  $(x, \pi)$ , we can recover how Receiver 'changed paradigm' and selected a new prior. Indeed, how he updates his prior over priors  $\mu$  depends on the total probability each  $\rho' \in \operatorname{supp} \mu$  assigns to x under  $\pi$  (see (1)). These probabilities can be uniquely inferred from  $q(\cdot|x, \pi)$ : using (2), we get

$$\pi(x|\omega) = \frac{q(\omega|x,\pi)}{\sigma(\omega)} \left[ \sum_{\tilde{\omega} \in \Omega} \pi(x|\tilde{\omega}) \sigma(\tilde{\omega}) \right].$$

Therefore, for every  $\rho' \in \operatorname{supp} \mu$ , expression (1) becomes

$$\hat{\mu}(\rho'|x,\pi) = \frac{\left[\sum_{\omega \in \Omega} q(\omega|x,\pi) \frac{\rho'(\omega)}{\sigma(\omega)}\right] \mu(\rho')}{\sum_{\tilde{\rho} \in \mathbf{supp} \, \mu} \left[\sum_{\omega \in \Omega} q(\omega|x,\pi) \frac{\tilde{\rho}(\omega)}{\sigma(\omega)}\right] \mu(\tilde{\rho})}.$$
(4)

Given  $q(\cdot|x,\pi)$ , this expression depends neither on x nor on  $\pi$ . Hence, let  $\hat{\mu}(\cdot;q)$  be the updated prior over priors given Sender's posterior q and

$$M(q) = \underset{\rho'}{\operatorname{arg max}} \hat{\mu}(\rho'; q).$$

By A1, Receiver picks the unique prior in M(q) which is  $\succ$ -maximal. Thus, define the following function of Sender's posterior:

$$\rho(q) = \begin{cases} \rho_0 & \text{if } q \in \Delta^c \\ \rho' \in M(q) \text{ s.t. } \rho' \succ \tilde{\rho} \text{ for all } \tilde{\rho} \in M(q), \ \tilde{\rho} \neq \rho' & \text{if } q \in \Delta^d \end{cases}$$
 (5)

Once we know which prior Receiver updates, we can again recover his posterior from

Sender's. The argument generalizes that in Alonso and Câmara (2013).<sup>27</sup> Recall that by assumption for every  $\omega \in \Omega$  there is a  $\rho' \in \operatorname{supp} \mu$  with  $\rho'(\omega) > 0$ . So for every  $q \in \Delta(\mathcal{S})$  the prior  $\rho(q)$  in (5) must assign positive probability to observing the  $(x, \pi)$  that led to q, i.e.,  $\sum_{\omega \in \Omega} \pi(x|\omega)\rho(\omega;q) > 0$ .

**Lemma 1.** Given any message  $(x, \pi)$ , let  $q(\cdot|x, \pi)$  be Sender's posterior,  $\rho \in \Delta(\Omega)$  satisfy  $\sum_{\omega \in \Omega} \pi(x|\omega)\rho(\omega) > 0$ , and  $p(\cdot|x, \pi)$  be Receiver's posterior after updating  $\rho$ . Then, for all  $\omega \in \Omega$ ,

$$p(\omega|x,\pi) = \frac{q(\omega|x,\pi)\frac{\rho(\omega)}{\sigma(\omega)}}{\sum_{\omega'\in\Omega}q(\omega'|x,\pi)\frac{\rho(\omega')}{\sigma(\omega')}}.$$
 (6)

If  $\rho(\omega) = 0$ , expression (6) is well-defined and  $p(\omega|x,\pi) = 0$ . Moreover, if  $(x,\pi)$  and  $(y,\pi)$  satisfy  $q(\cdot|x,\pi) = q(\cdot|y,\pi)$ , then  $p(\cdot|x,\pi) = p(\cdot|y,\pi)$ .

These observations lead to the paper's first main result: a full characterization of the set of feasible joint distributions over posteriors under A1. Although Receiver's response to information is more complicated here than in models with common-support priors, we can describe his posterior as a function of only Sender's posterior without reference to signal devices.

**Proposition 3.** Fix any  $\tau \in \mathcal{F}_{\sigma}$ . Under A1, for every  $q \in \operatorname{supp} \tau$  and  $\omega \in \Omega$ , Receiver's posterior satisfies

$$p(\omega;q) = \frac{q(\omega)\frac{\rho(\omega;q)}{\sigma(\omega)}}{\sum_{\omega' \in \Omega} q(\omega')\frac{\rho(\omega';q)}{\sigma(\omega')}},$$
(7)

where  $\rho(q)$  is given in (5).

Proposition 3 implies several key properties for the rest of the analysis.

First, when q confirms  $\rho_0$ , Receiver's posterior does not depend on the probability q assigns to states in  $\overline{\mathcal{R}}$ . Every  $q \in \Delta^c$  assigns positive probability to the event  $\mathcal{R}$  and  $\rho_0(\omega) = 0$  for  $\omega \in \overline{\mathcal{R}}$ . So

$$p(\omega;q) = \frac{\frac{q(\omega)}{q(\mathcal{R})} \frac{\rho_0(\omega)}{\sigma(\omega)/\sigma(\mathcal{R})}}{\sum_{\omega' \in \mathcal{R}} \frac{q(\omega')}{q(\mathcal{R})} \frac{\rho_0(\omega')}{\sigma(\omega')/\sigma(\mathcal{R})}} = \frac{q(\omega|\mathcal{R}) \frac{\rho_0(\omega)}{\sigma(\omega|\mathcal{R})}}{\sum_{\omega' \in \mathcal{R}} q(\omega'|\mathcal{R}) \frac{\rho_0(\omega')}{\sigma(\omega'|\mathcal{R})}}.$$
 (8)

Given  $q \in \Delta^c$ , Receiver has the same posterior he would have in a world in which priors were  $\rho_0$  and  $\sigma(\cdot|\mathcal{R})$  and Sender had posterior  $q(\cdot|\mathcal{R})$ .

Second, conditional on confirming  $\rho_0$  and inducing posterior p, Sender can have any posterior over  $\overline{\mathcal{R}}$ . Hence we lose the one-to-one relationship between Sender's and

<sup>&</sup>lt;sup>27</sup>Though immediate, the steps of this generalization are reproduced in Appendix B to make the paper self contained.

Receiver's posteriors which characterizes models with common-support priors (Proposition 1). To see this, take any  $q' \in \Delta^c$  and  $q'' \in \Delta(\overline{\mathcal{R}})$ . For every  $\omega \in \mathcal{S}$ , let  $\hat{q}(\omega) = \alpha q'(\omega|\mathcal{R}) + (1-\alpha)q''(\omega)$  with  $\alpha \in (0,1)$ . Clearly,  $\hat{q} \in \Delta(\mathcal{S})$ . It is easy to see that  $\hat{q}(\cdot|\mathcal{R}) = q'(\cdot|\mathcal{R})$  and  $\hat{q}(\cdot|\overline{\mathcal{R}}) = q''$ .

Third, since there are no restrictions on  $\pi$ , Sender can induce every posterior  $q(\cdot|\mathcal{R})$  and hence, through (8), every Receiver's posterior  $p \in \Delta(\mathcal{R})$ . This follows from Proposition 1 applied to 'priors'  $\rho_0$  and  $\sigma(\cdot|\mathcal{R})$ .

Fourth, although p(q) varies continuously in q over  $\Delta^c$ , it always changes discontinuously as q moves from  $\Delta^c$  to  $\Delta^d$ .

Corollary 2. For any 
$$q \in \Delta^c$$
 and  $q' \in \Delta^d$ , we have  $||p(q) - p(q')|| > ||p(q')|| > 0$ .

To see this differently, recall that by Proposition 2, as q moves between  $\Delta^c$  to  $\Delta^d$ , p jumps between two disjoint faces of  $\Delta(\mathcal{S})$ , i.e.,  $\Delta(\mathcal{R})$  and  $\Delta(\overline{\mathcal{R}})$  respectively. This discontinuity exactly happens when a message disproves Receiver's theory and hence defines formally the idea of surprise. This is clear if we consider any  $q \in \Delta^d$  and suppose that  $\sigma \in \Delta^c$  is arbitrarily close to it. In this case, a message that induces q does not surprise Sender—her belief moves slightly—but it does surprise Receiver—his belief jumps. The discontinuity is then natural: since after unexpected evidence Receiver must abandon  $\rho_0$ , a discrete difference in posteriors with and without surprise is necessarily implied by the very nature of being surprised.

Similar properties to the ones above can hold when Sender disproves  $\rho_0$ . For instance, let  $\sup \mu = (\rho_0, \rho_1)$  with  $\mu(\rho_0) > \frac{1}{2}$ . Then, for every  $q \in \Delta^d$  Receiver updates  $\rho_1$  and expression (7) can be written as

$$p(\omega;q) = \frac{q(\omega)\frac{\rho_1(\omega|\overline{R})}{\sigma(\omega|\overline{R})}}{\sum_{\omega'\in\overline{R}} q(\omega')\frac{\rho_1(\omega'|\overline{R})}{\sigma(\omega'|\overline{R})}}.$$
 (9)

So Receiver has the same posterior he would have in a world in which priors were  $\rho_1(\cdot|\overline{\mathcal{R}})$  and  $\sigma(\cdot|\overline{\mathcal{R}})$  and Sender had posterior q. Moreover, Sender can again induce every  $q \in \Delta^d$  and hence, through (9), every  $p \in \Delta(\overline{\mathcal{R}})$ . This last property, however, may not hold for more general  $\mu$ 's. The continuity of Receiver's posterior in Sender's over  $\Delta^d$  also depends on  $\mu$ . Again, if  $\operatorname{supp} \mu = (\rho_0, \rho_1)$  with  $\mu(\rho_0) > \frac{1}{2}$ , then  $p(\cdot)$  is continuous over  $\Delta^d$  (see (9)). Appendix A provides an example with discontinuities over  $\Delta^d$ . Intuitively, for a given prior Receiver's updating through Bayes' rule varies continuously as q varies in  $\Delta^d$ , but his choice of a prior may vary discontinuously when the maximum-likelihood criterion is inconclusive. This kind of discontinuity, however, is different from the one highlighted in Corollary 2.

# 5 Optimal Communication

Building on the previous results, this section examines the signal devices that Sender chooses.

We first need to specify Receiver's behavior. Given posterior p, he chooses an action that maximizes the resulting expected utility:<sup>28</sup>

$$a(p) \in \mathcal{A}(p) = \underset{a \in A}{\operatorname{arg \, max}} \ \mathbb{E}_p[u_R(a,\omega)].$$

In general, given p Receiver can be indifferent among multiple actions.

**Assumption 2.** If A(p) contains multiple actions, Sender can recommend Receiver to choose any  $a \in A(p)$  and Receiver follows the recommendation.

Assumption 2 corresponds to Kamenica and Gentzkow's (2011) 'Sender-preferred' subgame-perfect equilibrium and also appears in Alonso and Câmara (2013). In their settings with common-support priors, it ensures that an optimal signal device always exists.

Despite this standard assumption, due to the natural discontinuity in Receiver's posteriors after surprises, here optimal signal devices may not exist. To see why, given posteriors q and p let Sender's expected payoff be

$$v(q, p) = \max_{a \in \mathcal{A}(p)} \mathbb{E}_q[u_S(a, \omega)].$$

Given any mapping from Sender's to Receiver's posterior  $\hat{p}: \Delta(\mathcal{S}) \to \Delta(\Omega)$ , define

$$w(q) = v(q, \hat{p}(q))$$
 for all  $q \in \Delta(\mathcal{S})$ . (10)

As in Kamenica and Gentzkow (2011) and Alonso and Câmara (2013),  $v(\cdot, \cdot)$  is upper semicontinuous.<sup>29</sup> Moreover, since in those papers  $\hat{p}(\cdot)$  is continuous (Proposition 1),  $w(\cdot)$  is also upper semicontinuous. By contrast, here  $w(\cdot)$  need not be upper semicontinuous, because  $p(\cdot)$  in Proposition 3 (and 8 below) is discontinuous.

**Example 1** (Non-existence of Optimal Communication). Let  $S = \{\omega_1, \omega_2\}$  with  $\sigma = (\frac{1}{2}, \frac{1}{2})$ ,  $\mathcal{R} = \{\omega_1\}$ , and  $A = \{a, b, c\}$ . Sender's and Receiver's utility functions are as follows:

<sup>&</sup>lt;sup>28</sup>In Ortoleva (2012), SEU maximization is part of the Hypothesis-Testing Representation of the decision-maker's behavior.

<sup>&</sup>lt;sup>29</sup>This is because  $\mathcal{A}(p)$  is a nonempty, compact-valued, upper-hemicontinuous correspondence by Berge's Maximum Theorem.

$u_S(\cdot,\cdot)$	$\omega_1$	$\omega_2$
a	1	0
b	0	1
c	-1	-1

$u_R(\cdot,\cdot)$	$\omega_1$	$\omega_2$
a	1	0
b	1	0
c	1	1

For Sender, a is optimal if  $q(\omega_1) \geq \frac{1}{2}$  and b is optimal if  $q(\omega_1) \leq \frac{1}{2}$ . For any  $q \neq (0,1)$ , p(q) = (1,0) and  $\mathcal{A}((1,0)) = A$ . Therefore, Sender can make Receiver choose a or b depending on q. For q = (0,1), p(q) = (0,1) and  $\mathcal{A}((0,1)) = \{c\}$ .

Consider device  $\pi \in \Pi$  with the following properties:  $X_{\pi} = \{x_1, x_2\}, \ \pi(x_i | \omega_j) = 1 - \varepsilon$  if i = j, and  $\pi(x_i | \omega_j) = \varepsilon$  if  $i \neq j$ . So  $q(\cdot | x_1, \pi) = (1 - \varepsilon, \varepsilon)$  and  $q(\cdot | x_2, \pi) = (\varepsilon, 1 - \varepsilon)$ , each arising with probability  $\frac{1}{2}$ . For any  $\varepsilon > 0$ , Sender's expected payoff is then

$$\frac{1}{2}w(q(\cdot|x_1,\pi)) + \frac{1}{2}w(q(\cdot|x_2,\pi)) = \frac{1-\varepsilon}{2} + \frac{1-\varepsilon}{2} = 1-\varepsilon.$$
 (11)

However, for  $\varepsilon = 0$ , we have

$$\frac{1}{2}w((1,0)) + \frac{1}{2}w((0,1)) = \frac{1}{2}(1) + \frac{1}{2}(-1) = 0.$$

Clearly, the supremum of Sender's expected payoff over  $\Pi$  is 1, which is also the maximum she can hope for. But no  $\pi$  can achieve 1. This would require that Sender learn the true state to make Receiver choose her preferred action accordingly; but then Receiver must also learn the true state and hence will choose c in  $\omega_2$ .

Importantly, these technical difficulties do not preclude an instructive analysis of how Sender communicates. Under our assumptions, we can view Receiver as a 'machine' computing posteriors and associated optimal actions. Sender's problem is then essentially a single-agent decision problem. Although this problem may not have an exact solution, we can examine its value function and infer from it properties of signal devices that are virtually optimal. For instance, in example 1 expression (11) says that it is virtually optimal for Sender to adopt a  $\pi$  that allows her to almost perfectly learn  $\omega$  but never disproves  $\rho_0$ . To formalize this approach, denote Sender's expected payoff from any  $\tau$  by

$$V(\tau) = \mathbb{E}_{\tau}[w(q)],$$

where  $w(\cdot)$  in (10) uses the function  $p(\cdot)$  in Proposition 3 (or 8 below). Sender would like to reach the highest  $V(\tau)$  by choosing  $\tau \in \mathcal{F}_{\sigma}$ . This problem's value function is<sup>30</sup>

$$W_{\sigma} = \sup_{\tau \in \mathcal{F}_{\sigma}} V(\tau). \tag{12}$$

<sup>&</sup>lt;sup>30</sup>By characterizing  $W_{\sigma}$ , we will also obtain useful information on  $\varepsilon$ -optimal  $\tau$ 's for any  $\varepsilon > 0$ , since there always exists a feasible  $\tau$  with  $V(\tau) \geq W_{\sigma} - \varepsilon$ .

The key observation here is that the usual 'concavification' procedure<sup>31</sup> is well defined even if the function w in (10) is not upper semicontinuous.

**Definition 2** (Concavification). Given any  $g: E \subset \mathbb{R}^n \to \mathbb{R}$ , its concavification  $\hat{g}$  is the lowest concave function with  $\hat{g}(e) \geq g(e)$  for all  $e \in E$ .

The concavification of g is useful because it satisfies

$$\hat{g}(e) = \sup\{\xi : (e, \xi) \in \operatorname{co}(\operatorname{hyp} g)\} \quad \text{for all } e \in E,$$
(13)

where co(hyp g) is the convex hull of the hypograph<sup>32</sup> of g (Rockafellar (1997), p. 52).<sup>33</sup> Relying on the concavification of w, we obtain the following.

**Lemma 2.**  $W_{\sigma} = \hat{w}(\sigma)$ . Moreover, in (12), it is without loss of generality to restrict attention to distributions  $\tau$  with  $|\mathbf{supp}\,\tau| \leq |\mathcal{S}|$ .

This lemma suggests the following definition.

**Definition 3.** If  $\hat{w}(\sigma) = w(\sigma)$ , it is optimal for Sender not to reveal information ( $\tau = \delta_{\sigma}$  is optimal).<sup>34</sup> If  $\hat{w}(\sigma) > w(\sigma)$ , it is (virtually) optimal for Sender to reveal some information.

In the second case, there exists a feasible  $\tau \neq \delta_{\sigma}$  such that  $V(\tau) > w(\sigma)$ . Definition 3 refers to Sender's viewpoint, as some  $\pi$  may reveal information according to her theory,  $\sigma$ , but no information according to Receiver's theory,  $\rho_0$ .<sup>35</sup>

In this paper the main question is not whether Sender reveals any information at all, but if and when she disproves Receiver's prior and how she then optimally communicates. Clearly, if  $\hat{w}(\sigma) = w(\sigma)$ , it is optimal to never communicate and hence never disprove  $\rho_0$ . So hereafter we focus on the case with  $\hat{w}(\sigma) > w(\sigma)$ .

To address our questions, we need a different concavification argument, which considers the function w over  $\Delta^c$  and  $\Delta^d$  separately. Let  $w^c$  and  $w^d$  be the restrictions of w to  $\Delta^c$  and  $\Delta^d$  and  $\hat{w}^c$  and  $\hat{w}^d$  their concavifications. For  $i = c, d, \hat{w}^i(q) \leq \hat{w}(q)$  for all

 $<sup>^{31}</sup>$ See Section 2 for references.

<sup>&</sup>lt;sup>32</sup>Given a function  $g: \mathbb{R}^n \to \mathbb{R}$ , its hypograph is defined as hyp  $g = \{(e, r) \in \mathbb{R}^{n+1} : r \leq g(e)\}$ .

<sup>&</sup>lt;sup>33</sup>In example 1,  $\hat{w}(q) = 1$  for all  $q \neq (0, 1)$  and  $\hat{w}((0, 1)) = 0$ .

<sup>&</sup>lt;sup>34</sup>Hereafter,  $\delta_z$  stands for the distribution that assigns probability 1 to z.

<sup>&</sup>lt;sup>35</sup>In example 1,  $w(\sigma) = \frac{1}{2}$ ,  $W_{\sigma} = 1$ , and Sender can approximate  $W_{\sigma}$  using a feasible  $\tau$  with  $\operatorname{supp} \tau = \{(\varepsilon, 1 - \varepsilon), (1, 0)\}$  for  $\varepsilon$  arbitrarily small. With this  $\tau$ , however, Receiver's posteriors are  $p((1, 0)) = p((\varepsilon, 1 - \varepsilon)) = (1, 0) = \rho_0$ . So, from Receiver's viewpoint, no information is revealed.

 $<sup>^{36}</sup>$ As shown below, conditional on  $\mathcal{R}$  Sender communicates as in the standard settings of Kamenica and Gentzkow (2011) and Alonso and Câmara (2013). So the sufficient conditions for Sender to optimally reveal some information in those settings can be adapted to ensure that  $\hat{w}(\sigma) > w(\sigma)$ .

 $q \in \Delta^i$  since  $\operatorname{co}(\operatorname{hyp} w^i) \subset \operatorname{co}(\operatorname{hyp} w)$ ; moreover,  $\lim_{q' \to q} \hat{w}^i(q') \geq \hat{w}^i(q)$  for all  $q \in \Delta^i$ .<sup>37</sup> As shown later,  $\hat{w}^d$  is continuous if  $w^d$  is continuous, which holds when  $p(\cdot)$  is continuous over  $\Delta^d$ .

**Lemma 3.** If  $\hat{w}^c(\sigma) < \hat{w}(\sigma)$ , there exists  $\tau \in \mathcal{F}_{\sigma}$  such that  $V(\tau) > \hat{w}^c(\sigma)$ . Moreover, if  $V(\tau) > \hat{w}^c(\sigma)$ , then the probability  $\tau^d = \tau(D_{\tau})$  of disproving  $\rho_0$  is strictly positive.

This result suggests the following terminology.

**Definition 4.** If  $\hat{w}^c(\sigma) = \hat{w}(\sigma)$ , it is (virtually) optimal for Sender to always confirm  $\rho_0$ . If  $\hat{w}^c(\sigma) < \hat{w}(\sigma)$ , it is (virtually) optimal for Sender to disprove  $\rho_0$ .

The next result characterizes how Sender communicates with Receiver when she always confirms  $\rho_0$ . This is also a key step in understanding her incentives to disprove  $\rho_0$ . Recall that

$$\hat{w}^{c}(\sigma) = \sup_{\{\tau \in \mathcal{F}_{\sigma} : \mathbf{supp} \ \tau \subset \Delta^{c}\}} V(\tau) = \sup_{\mathcal{F}_{\sigma}^{c}} \sum_{m} \tau_{m} w(q_{m}), \tag{14}$$

where

$$\mathcal{F}_{\sigma}^{c} = \left\{ \{ (q_n, \tau_n) \}_{n=1}^{N} : \sum_{m=1}^{N} \tau_m q_m = \sigma, \sum_{m=1}^{N} \tau_m = 1, \tau_m \ge 0, q_m \in \Delta^c, \forall m \right\}.$$

Note that even if  $q \in \Delta^c$  assigns positive probability to some  $\omega \in \overline{\mathcal{R}}$  and hence Sender learns information about it, Receiver continues to think that  $\omega$  is impossible  $(p(\omega;q)=0)$ . For this reason, if Sender chooses a  $\tau$  that induces such a q, we say that she *hides*  $\omega$  in posterior q. For every  $\omega \in \overline{\mathcal{R}}$ , let

$$u_S^*(\omega) = \max_{p \in \Delta(\mathcal{R})} \{ \max_{a \in \mathcal{A}(p)} u_S(a, \omega) \}.$$

This quantity is well defined because  $\Delta(\mathcal{R})$  is compact and the term in brackets is upper semicontinuous in p; moreover, it can be directly computed from primitives.

**Proposition 4.** To compute  $\hat{w}^c(\sigma)$  in (14), it is without loss of generality to consider distributions  $\tau$  with the following properties:

(1) each posterior hides at most one state: for every  $q \in \operatorname{supp} \tau$ , there exists at most

<sup>&</sup>lt;sup>37</sup>Being concave,  $\hat{w}^c$  is continuous at every  $q \in int\Delta^c$  by Theorem 10.1 in Rockafellar (1997). By Theorem 10.3 in Rockafellar (1997), there exists only one way to extend  $\hat{w}^c$  from  $int\Delta^c$  to a continuous finite concave function on  $\Delta(\mathcal{S})$ . In fact, this extension equals  $-\mathrm{cl}(-\hat{w}^c)$  on  $\Delta(\mathcal{S})$  where  $\mathrm{cl}(-\hat{w}^c)$  is the closure of the convex function  $-\hat{w}^c$  (see the proof of Theorem 10.3 and p. 52 of Rockafellar (1997)). Therefore, for any  $q \in \Delta^c \setminus int\Delta^c$ , we have  $\lim_{q' \to q} \hat{w}^c(q') \geq \hat{w}^c(q)$  since  $\mathrm{cl}(-\hat{w}^c) \leq -\hat{w}^c$ . A similar argument applies for  $\hat{w}^d$ .

one  $\omega \in \overline{\mathbb{R}}$  such that  $q(\omega) \in (0,1)$  and

$$q = (1 - q(\omega))q(\cdot|\mathcal{R}) + q(\omega)\delta_{\omega};$$

(2) each state is hidden in one posterior: for every  $\omega \in \overline{\mathcal{R}}$ , there exists a unique  $q \in \operatorname{supp} \tau$  such that  $q(\omega) > 0$ ; moreover,  $u_S(a(p(q)), \omega) = u_S^*(\omega)$  where a(p(q)) is Receiver's induced action.

When Sender always confirms  $\rho_0$ —either because she has to or because it is optimal—she communicates as follows. First, each message leads to a posterior q such that she rules out all states that are impossible under  $\rho_0$  except at most one, say  $\omega \in \overline{\mathcal{R}}$ . Second, she hides each  $\omega \in \overline{\mathcal{R}}$  in one posterior which, in  $\omega$ , also gives her the best payoff among all actions Receiver would choose for posteriors that are consistent with his theory (i.e., in  $\Delta(\mathcal{R})$ ). We will refer to this strategy as *optimal hiding of*  $\omega$  and to  $u_S^*(\omega)$  as its payoff.

The intuition is this. Recall that the priors' different supports allow Sender to become more informed on states in  $\overline{\mathcal{R}}$  without necessarily making Receiver more informed as well. Property (1) arises because Sender can replace a signal x inducing a posterior that hides, say, two states in  $\overline{\mathcal{R}}$  with two signals, each arising in only one of the two states and thus inducing a distinct posterior that hides only one state. Moreover, she can reallocate the probability that x resulted from each  $\omega \in \mathcal{R}$  across the new signals so that, conditional on  $\mathcal{R}$ , the new posteriors are identical to the original one—and hence lead to the same posterior and optimal actions of Receiver. Property (2) arises because Sender can ensure that each  $\omega \in \overline{\mathcal{R}}$  leads to only one of the signals that her device can generate and, among these, the one with the highest payoff in  $\omega$ . Moreover, she can design for  $\omega$  a signal that arises almost exclusively in this state and hence the resulting Receiver's action almost never matters for all other states. Therefore, she can freely manipulate Receiver's posterior to cater to  $\omega$  entirely and thus get payoff  $u_s^*(\omega)$ .

Proposition 4 allows us to obtain a simple expression for  $\hat{w}^c(\sigma)$ .

Corollary 3. Sender's expected payoff from always confirming  $\rho_0$  is

$$\hat{w}^{c}(\sigma) = \sigma(\mathcal{R})\hat{w}^{c}(\sigma(\cdot|\mathcal{R})) + \sum_{\omega \in \overline{\mathcal{R}}} \sigma(\omega)u_{S}^{*}(\omega).$$

The value of always confirming  $\rho_0$  is the value Sender would achieve if she could divide her problem into two steps as follows. First, she learns whether the state is consistent with Receiver's theory, i.e., the event  $\mathcal{R}$  or  $\overline{\mathcal{R}}$ . Given  $\mathcal{R}$ , her belief has the same support as Receiver's prior and she chooses the optimal signal device as in settings with commonsupport priors, the case studied in Kamenica and Gentzkow (2011) and Alonso and Câmara (2013). Given  $\overline{\mathcal{R}}$ , she chooses a fully informative device, where her payoff in each state  $\omega$  is replaced by that from optimally hiding  $\omega$ .

We can now define the opportunity cost of surprising Receiver. This is the payoff Sender expects, at any  $q \in \Delta^d$ , from optimally hiding all states for which she is providing supporting evidence (i.e., all  $\omega \in \operatorname{\mathbf{supp}} q$ ).

**Definition 5.** The opportunity cost of surprising Receiver is defined by the function  $h: \Delta^d \to \mathbb{R}$  such that

$$h(q) = \mathbb{E}_q[u_S^*(\omega)].$$

This leads to the following simple necessary and sufficient condition for Sender to optimally disprove  $\rho_0$  (i.e., for  $\hat{w}(\sigma) > \hat{w}^c(\sigma)$ ). There must exist *some* posterior q disproving  $\rho_0$  at which, given Receiver's response, Sender's expected payoff strictly exceeds the opportunity cost of surprising.

**Proposition 5.** It is optimal for Sender to disprove  $\rho_0$  if and only if  $w^d(q) > h(q)$  for some  $q \in \Delta^d$ .

Note that this condition can be directly checked without using any concavification.

The next result characterizes Sender's payoff and communication strategy when she disproves  $\rho_0$ . Some preliminary remarks are in order. If in definition 2 E is compact and g is upper semicontinuous (u.s.c.), then  $\hat{g}$  is continuous and hence  $\hat{g}(e)$  in (13) is achieved for every  $e \in E$ .<sup>38</sup> The set  $\Delta^d$  is compact, but  $w^d$  need not be u.s.c.. To deal with this, we slightly modify  $w^d$  by considering the lowest u.s.c. function pointwise larger than  $w^d$ , denoted  $w^d_*$ . On the interior of  $\Delta^d$ ,  $\hat{w}^d$  coincides with  $\hat{w}^d_*$  (see Lemma 4 in Appendix B). Now define  $m: \Delta^d \to \mathbb{R}$  by

$$m = \max\{w_*^d, h\}.$$

For every q disproving  $\rho_0$ , m(q) captures whether with belief q Sender expects to do better by actually surprising Receiver or by optimally hiding states. Note that since h and  $w_*^d$  are u.s.c., so is m.<sup>39</sup>

**Proposition 6.** If Sender optimally disproves  $\rho_0$ , then

$$\hat{w}(\sigma) = \tau^c \hat{w}^c(q^c) + (1 - \tau^c) \hat{w}_*^d(q^d) = \sigma(\mathcal{R}) \hat{w}^c(\sigma(\cdot|\mathcal{R})) + \sigma(\overline{\mathcal{R}}) \hat{m}(\sigma(\cdot|\overline{\mathcal{R}}))$$
(15)

where  $q^c \in \Delta^c$ ,  $q^d \in \Delta^d$ ,  $\tau^c \in [\sigma(\mathcal{R}), 1)$ ,  $\tau^c q^c + (1 - \tau^c)q^d = \sigma$ , and

$$\hat{m}(\sigma(\cdot|\overline{\mathcal{R}})) = \frac{\tau^c q^c(\overline{\mathcal{R}})}{\sigma(\overline{\mathcal{R}})} h(q^c(\cdot|\overline{\mathcal{R}})) + \frac{1 - \tau^c}{\sigma(\overline{\mathcal{R}})} \hat{w}_*^d(q^d).$$

<sup>&</sup>lt;sup>38</sup>This property follows from Part (i) of Lemma 4 in Appendix B.

<sup>&</sup>lt;sup>39</sup>To see that m is u.s.c., note that hyp  $m = \text{hyp } h \cup \text{hyp } w_*^d$  where both h and  $w_*^d$  are u.s.c.. By Theorem 7.1 in Rockafellar (1997), a function is u.s.c. if and only if its hypograph is closed.

Any (conditional) distribution on  $\Delta^d$  achieving  $\hat{w}^d_*(q^d)$  assigns positive probability only to q's such that  $w^d_*(q) \geq h(q)$ . Finally,  $\tau^c = \sigma(\mathcal{R})$  if and only if  $\hat{m}(\sigma(\cdot|\overline{\mathcal{R}})) = \hat{w}^d(\sigma(\cdot|\overline{\mathcal{R}}))$ .

Proposition 6 can be interpreted as follows. The first part of (15) highlights the dichotomy in Sender's problem between confirming and disproving  $\rho_0$ . To solve her problem, she can find the best among all convex combinations averaging to  $\sigma$  and involving only two posteriors: one,  $q^c$ , confirms  $\rho_0$  and one,  $q^d$ , disproves it. To  $q^c$ , she assigns the expected payoff of the fictitious scenario in which  $q^c$  is her *prior* and she always confirms  $\rho_0$  (Corollary 3). To  $q^d$ , she assigns the expected payoff of the fictitious scenario in which  $q^d$  is again her prior, but Receiver responds to messages in the possibly non-Bayesian way captured by the function  $p(\cdot)$  in Proposition 3 after surprises. The weight  $\tau^c$  she assigns to  $q^c$  pins down the ex-ante probabilities of confirming and disproving  $\rho_0$ .

The second part of (15) highlights the difference in Sender's communication between states inside and outside Receiver's theory and makes her overall payoff easier to compute. Conditional on  $\mathcal{R}$ , Sender gets the expected payoff from optimally communicating with Receiver in the fictitious (standard) world with priors  $\sigma(\cdot|\mathcal{R})$  and  $\rho_0$ . Conditional on  $\overline{\mathcal{R}}$ , Sender gets the expected payoff from optimally combining disproving  $\rho_0$  and hiding states given 'prior'  $\sigma(\cdot|\overline{\mathcal{R}})$ . This is clear from the expression of  $\hat{m}(\sigma(\cdot|\overline{\mathcal{R}}))$ , which can be computed by finding  $\hat{w}^d_*$  and the best convex combination  $\gamma h(q_1) + (1 - \gamma)\hat{w}^d_*(q_2)$  with  $q_1, q_2 \in \Delta(\overline{\mathcal{R}})$  and  $\gamma q_1 + (1 - \gamma)q_2 = \sigma(\cdot|\overline{\mathcal{R}})$ . Given the optimal  $\gamma$ , we get the probability of disproving  $\rho_0$ ,  $\tau^d = (1 - \gamma)\sigma(\overline{\mathcal{R}})$ , the probability of hiding states,  $\tau^c q^c(\overline{\mathcal{R}}) = \gamma \sigma(\overline{\mathcal{R}})$ , and  $q^c = q_2 + \frac{1}{\tau^c}(\sigma - q_2)$  since  $q_2 = q^d$ . Proposition 6 also says—as should be expected—that when Sender ends up with q disproving  $\rho_0$ , she never expects that she could do strictly better by hiding the states for which she is providing supporting evidence.

Finally, Proposition 6 gives a necessary and sufficient condition for Sender to optimally disprove Receiver's theory whenever the state is inconsistent with it. The key is that, at the specific 'prior'  $\sigma(\cdot|\overline{\mathcal{R}})$ , Sender has one way to maximize her expected payoff which never involves optimally hiding states. Note, however, that  $\tau^c$  in Proposition 6 need not be unique. Of course, a sufficient (but not necessary) condition for  $\tau^c = \sigma(\mathcal{R})$  is  $w^d \geq h$  with strict inequality for some  $q \in \Delta^d$ .

The next result gives a simpler sufficient condition for Sender *not* to disprove  $\rho_0$  whenever the state is inconsistent with it. Let  $T(\sigma(\cdot|\overline{\mathcal{R}}))$  be the set of distributions  $\tau \in \mathcal{F}_{\sigma(\cdot|\overline{\mathcal{R}})}$  such that  $\mathbb{E}_{\tau}[w_*^d(q)] = \hat{w}_*^d(\sigma(\cdot|\overline{\mathcal{R}}))$ .

Corollary 4. Suppose  $\hat{w}(\sigma) > \hat{w}^c(\sigma)$ . If there exists  $\hat{\tau} \in T(\sigma(\cdot|\overline{\mathcal{R}}))$  and  $q \in \operatorname{supp} \hat{\tau}$  such that  $w_*^d(q) < h(q)$ , then  $\tau^d < \sigma(\overline{\mathcal{R}})$ .

In words, Sender will not disprove  $\rho_0$  whenever  $\omega \in \overline{\mathcal{R}}$  if there is an optimal way of doing so that involves a posterior at which the opportunity cost of surprising exceeds the resulting payoff. All elements in this condition can be directly computed from primitives of the model. Note that we can have  $\hat{w}_*^d(\sigma(\cdot|\overline{\mathcal{R}})) > h(\sigma(\cdot|\overline{\mathcal{R}}))$  and yet  $\tau^d < \sigma(\overline{\mathcal{R}})$ . Also, conditional on  $\overline{\mathcal{R}}$  Sender and Receiver could have sufficiently aligned preferences, so that  $\hat{w}_*^d(\sigma(\cdot|\overline{\mathcal{R}}))$  is achieved by fully revealing all states in  $\overline{\mathcal{R}}$ . Yet Sender may prefer to optimally hide some  $\omega \in \overline{\mathcal{R}}$ , thus confirming  $\rho_0$  in that sate.

Proposition 6 allows us to derive simple conditions for identifying which states Sender may hide.

**Proposition 7.** Fix any  $\omega \in \overline{\mathcal{R}}$ . If  $w^d(\delta_\omega) > u_S^*(\omega)$ , then Sender never hides  $\omega$ —i.e.,  $\omega \notin \operatorname{supp} q^c$  in (15). If  $w^d(\delta_\omega) < u_S^*(\omega)$  and  $w^d$  is convex, then Sender always hides  $\omega$ —i.e.,  $\omega \notin \operatorname{supp} q^d$  in (15).

When  $\omega \in \overline{\mathcal{R}}$  satisfies  $w^d(\delta_\omega) = u_S^*(\omega)$ , we can safely say that Sender never hides this  $\omega$  as well. Indeed, for every  $\tau \in \mathcal{F}_{\sigma(\cdot|\overline{\mathcal{R}})}$  which implies that she optimally hides  $\omega$  with positive probability, there exists  $\tau' \in \mathcal{F}_{\sigma(\cdot|\overline{\mathcal{R}})}$  such that she never hides  $\omega$  and  $\mathbb{E}_{\tau'}[m(q)] = \mathbb{E}_{\tau}[m(q)].^{40}$  Hence optimally hiding  $\omega$  is not necessary to achieve  $\hat{m}(\sigma(\cdot|\overline{\mathcal{R}}))$  in (15). When  $w^d$  is not convex, even if  $w^d(\delta_\omega) < u_S^*(\omega)$ , Sender may reveal  $\omega \in \overline{\mathcal{R}}$  with positive probability—i.e.,  $\omega \in \operatorname{\mathbf{supp}} q^d$  in (15). Section 5.1 provides an example.

Proposition 6 also implies that a (virtually) optimal  $\tau$  has to satisfy the following "no regret" property. When Sender ends up disproving  $\rho_0$ , she cannot strictly prefer any action that Receiver takes when she confirms  $\rho_0$ .

Corollary 5 ('No Regret'). Take any  $\tau \in \mathcal{F}_{\sigma}$  with  $\tau^d > 0$ . Suppose that for some  $\hat{q} \in D_{\tau}$  and  $q' \in C_{\tau}$  we have  $w^d(\hat{q}) < \mathbb{E}_{\hat{q}}[u_S(a(p(q')), \omega)]$ . Then,  $V(\tau) < \hat{w}(\sigma)$ .

This is again because of the flexibility Sender enjoys in managing her posteriors, while not changing Receiver's. She can have posterior  $\hat{q}$  conditional on  $\overline{\mathcal{R}}$  and q' conditional on  $\mathcal{R}$ , and at the same time assign very low probability to the states in  $\mathcal{R}$ . In this way, she makes Receiver choose a(p(q')) while having a posterior arbitrarily close to  $\hat{q}$  and hence an expected payoff close to  $\mathbb{E}_{\hat{q}}[u_S(a(p(q')), \omega)]$ .

 $<sup>^{40}</sup>$ This is easy to see from the proof of Proposition 7.

## 5.1 Illustrative Example: Court

We can now solve the court example from the introduction. Recall that here  $S = \{\omega_0, \omega_1, \omega_2, \omega_3\}$ ,  $\mathcal{R} = \{\omega_0, \omega_2\}$ ,  $\sigma = (0.35, 0.4, 0.15, 0.1)$ ,  $\rho_0 = (0.7, 0, 0.3, 0)$ , and  $\rho_1 = \sigma^{4}$ . The judge (Receiver) cares about matching the defendant's refund with the state. Letting  $f(\omega_i) = i$  (where i stands for i thousands/millions of dollars) and  $A = \{0, 1, 2, 3\}$ , we have  $u_R(a, \omega_i) = 1$  if  $a = f(\omega_i)$  and  $u_R(a, \omega_i) = 0$  otherwise. Concerning the lawyer (Sender), we have  $u_S(a, \omega_i) = a$ , the refund amount.

By Proposition 5, Sender must disprove  $\rho_0$  with strictly positive probability. For Sender the best way to hide both  $\omega_1$  and  $\omega_3$  is to ensure that Receiver chooses a=2. This occurs at posteriors q assigning at least probability 0.5 to  $\omega_2$  conditional on  $\mathcal{R}$ . Therefore,  $u_S^*(\omega_1) = u_S^*(\omega_3) = 2$  and hence h(q) = 2 for all  $q \in \Delta^d = \Delta(\{\omega_1, \omega_3\})$ . Since  $w^d(\delta_{\omega_3}) = 3$ , Proposition 5 implies the claimed property.

To obtain Sender's payoff  $\hat{w}(\sigma)$  in (15), we first compute  $\hat{w}^c(\sigma(\cdot|\mathcal{R}))$  with the help of Figure 1. As in Kamenica and Gentzkow (2011), when both parties have prior  $\sigma(\cdot|\mathcal{R}) = (0.7, 0.3)$ , it is optimal for Sender to generate two posteriors: At the first, both parties assign probability 1 to  $\omega_0$  (i.e., q = (1,0,0,0)) and Receiver chooses a = 0; At the second, both assign equal probabilities to  $\omega_0$  and  $\omega_2$  (i.e., q' = (0.5,0,0.5,0)) and Receiver chooses a = 2. The full red dots in Figure 1 represent the corresponding expected payoffs for Sender. Hence,  $\hat{w}^c(\sigma(\cdot|\mathcal{R})) = \frac{0.3}{0.5}2 = 1.2$ .

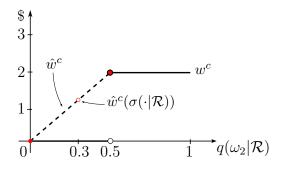


Figure 1: Confirming  $\rho_0$ 

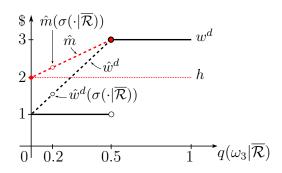


Figure 2: Disproving  $\rho_0$ 

Second, we compute  $\hat{m}(\sigma(\cdot|\overline{\mathcal{R}}))$  in (15). Since for all  $q \in \Delta^d$  Receiver's posterior equals Sender's,  $w^d(q) = 1$  if  $q(\omega_3) < 0.5$  and  $w^d(q) = 3$  otherwise. Therefore,  $m = \max\{2, w^d\}$  equals 2 if  $q(\omega_3) < 0.5$  and 3 otherwise. As illustrated in Figure 2, again when both parties have prior  $\sigma(\cdot|\overline{\mathcal{R}}) = (0.8, 0.2)$  and Sender has payoff function m, she optimally induces two posteriors: At the first, both parties assign equal probability to  $\omega_1$  and  $\omega_3$ 

<sup>41</sup> Note that, with these priors, Sender's and Receiver's posteriors coincide if either  $q \in \Delta(\mathcal{R})$  or  $q \in \Delta(\overline{\mathcal{R}})$ .

(i.e.,  $q^d = (0, 0.5, 0, 0.5)$ ) and Receiver chooses a = 3; At the second, both parties assign probability 1 to  $\omega_1$  (i.e.,  $\hat{q} = (0, 1, 0, 0)$ ) and Sender get's payoff  $u_S^*(\omega_1)$ . So,

$$\hat{m}(\sigma(\cdot|\overline{R})) = \frac{0.2}{0.5}w^d(q^d) + \frac{0.3}{0.5}h(\delta_{\omega_1}) = 2.4.$$

Given this, Sender's overall expected payoff (the expected refund for her client) is

$$W_{\sigma} = 0.5(1.2) + 0.5(2.4) = 1.8.$$

Consider now the probability with which Sender disproves  $\rho_0$ . By Proposition 6, this probability is strictly less than  $0.5 = \sigma(\overline{\mathcal{R}})$ , because

$$\hat{w}^d(\sigma(\cdot|\overline{R})) = \frac{0.2}{0.5}w^d(q^d) + \frac{0.3}{0.5}w^d(\delta_{\omega_1}) = \frac{0.2}{0.5}3 + \frac{0.3}{0.5}1 = 1.8$$

which is lower than  $\hat{m}(\sigma(\cdot|\overline{\mathcal{R}}))$ .<sup>42</sup> In fact,

$$\tau^d = \sigma(\overline{\mathcal{R}}) \frac{0.2}{0.5} = 0.2.$$

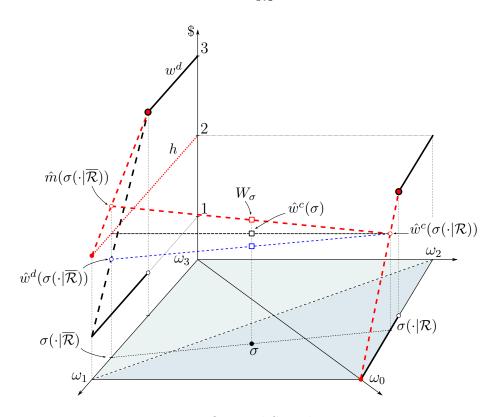


Figure 3: Optimal Signal Device

Figure 3 helps us describe Sender's (virtually) optimal signal device. It involves four signals: three confirm  $\rho_0$  and one disproves it. Signal  $x_0$  arises only in state  $\omega_0$ , revealing

<sup>&</sup>lt;sup>42</sup>In this example, we could have immediately used Corollary 4 to reach the same conclusion since  $w^d(\delta_{\omega_1}) < h(\delta_{\omega_1})$ .

it. Signal  $x_2$  arises in  $\omega_0$  and  $\omega_2$  and leads both parties to assign them equal probability. Signal  $x_1$  arises in  $\omega_1$  and  $\omega_2$  and leads Sender to assign  $\omega_1$  probability arbitrarily close to 1, while Receiver assigns probability 1 to  $\omega_2$ . Finally, signal  $x_3$  arises in  $\omega_1$  and  $\omega_3$  and leads both parties to assign them equal probability. Graphically, the solid red dots in Figure 3 represent Sender's expected payoff for each signal.

For comparison, consider Sender's payoff from the strategies of never and always disproving  $\rho_0$ . In the first case, her maximal expected payoff is given in Corollary 3:

$$\hat{w}^c(\sigma) = 0.5(1.2) + 0.5(2) = 1.6.$$

This corresponds to the black square in Figure 3. In the second case, her maximal expected payoff is

$$0.5\hat{w}^c(\sigma(\cdot|\mathcal{R})) + 0.5\hat{w}^d(\sigma(\cdot|\overline{\mathcal{R}})) = 0.5(1.2) + 0.5(1.8) = 1.5.$$

This corresponds to the blue square in Figure 3. So, by optimally combining hiding and surprising, Sender improves her payoff by 12.5% relative to always hiding and by 20% relative to always surprising.

## 5.2 Application with Quadratic Payoffs

This section applies the previous results to the classic and broadly used<sup>43</sup> setting in which states and actions take values on the real line and Sender's and Receiver's payoffs are quadratic loss functions of the gap between an ideal and the implemented action.

More specifically, let  $\Omega = \{\omega_1, \ldots, \omega_n\} \subset \mathbb{R}_{++}$  where  $\omega_i < \omega_{i+1}$  for all i, and let  $A = \mathbb{R}$ . The payoff functions are  $u_S(a, \omega) = -(a - \omega)^2$  and  $u_R(a, \omega) = -(a - \beta(\omega))^2$  where  $\beta(\omega)$  is Receiver's ideal action in  $\omega$ . Assume that  $\beta(\omega) > \omega$  for all  $\omega$  and that  $\beta(\cdot)$  involves a fixed and a linear component in  $\omega$ , i.e.,  $\beta(\omega) = \kappa\omega + b$  with  $\kappa > 0$  and  $b \geq 0$ . Receiver initially views low and high enough states as impossible:  $\mathcal{R} = \{\omega_i\}_{i=\underline{m}}^{\overline{m}}$  with  $1 < \underline{m} < \overline{m} < n$ . Moreover,  $\sup \mu = (\rho_0, \rho_1)$  with  $\mu(\rho_0) > \frac{1}{2}$ ,  $\rho_0 = \sigma(\cdot | \mathcal{R})$ , and  $\rho_1 = \sigma$ . In this setting, as usual, given posterior p Receiver chooses his expected ideal action:  $a(p) = \mathbb{E}_p[\beta]$ .

First of all, we need to compute Sender's expected payoffs for any posterior and from optimally hiding states. For any  $q \in \Delta(\mathcal{S})$ ,

$$w(q) = \mathbb{E}_{q}[u_{S}(\mathbb{E}_{p(q)}[\beta], \omega)]$$

$$= -\kappa^{2}(\mathbb{E}_{p(q)}[\omega])^{2} + 2\kappa \mathbb{E}_{p(q)}[\omega]\mathbb{E}_{q}[\omega] - 2b(\kappa \mathbb{E}_{p(q)}[\omega] - \mathbb{E}_{q}[\omega]) - b^{2} - \mathbb{E}_{q}[\omega^{2}].$$

$$(16)$$

<sup>&</sup>lt;sup>43</sup>See Footnote 10.

Recall that  $p(q) = q(\cdot | \mathcal{R})$  for  $q \in \Delta^c$  by (8); also, p(q) = q for  $q \in \Delta^d$ . So  $\mathbb{E}_{p(q)}[\omega]$  is always linear in q and  $w^d$  is continuous. To compute  $u_S^*(\omega)$  we have to consider several cases. For  $\omega < \omega_{\underline{m}}$  it is clearly optimal to hide  $\omega$  so that Receiver assigns probability 1 to  $\omega_{\underline{m}}$ , the closest state to  $\omega$  in  $\mathcal{R}$ ; so  $u_S^*(\omega) = -(\beta(\omega_{\underline{m}}) - \omega)^2$ . For  $\omega > \omega_{\overline{m}}$  the optimal hiding strategy depends on  $\beta$ . If  $\omega_{\overline{m}} < \omega < \beta(\omega_{\underline{m}})$  the situation is equivalent to the case  $\omega < \omega_{\underline{m}}$ . If  $\beta(\omega_{\overline{m}}) \leq \omega$ , it is optimal to hide  $\omega$  so that Receiver assigns probability 1 to  $\omega_{\overline{m}}$ ; so  $u_S^*(\omega) = -(\beta(\omega_{\overline{m}}) - \omega)^2$ . If instead  $\beta(\omega_{\overline{m}}) > \omega \geq \beta(\omega_{\underline{m}})$ , there always exists  $p \in \Delta(\mathcal{R})$  such that  $\mathbb{E}_p[\beta] = \omega$ ; hence  $u_S^*(\omega) = 0$ .

Proposition 5 now implies that Sender will disprove Receiver's theory with positive probability  $(\tau^d > 0)$ , independently of  $\kappa$  and b. Indeed, for any  $\omega < \omega_{\underline{m}}$ ,  $\delta_{\omega} \in \Delta^d$  and  $w^d(\delta_{\omega}) = -(\beta(\omega) - \omega)^2$  which is strictly larger than  $h(\delta_{\omega}) = -(\beta(\omega_{\underline{m}}) - \omega)^2$  because  $\omega < \beta(\omega) < \beta(\omega_{\underline{m}})$ . This also implies that Sender never hides states below  $\omega_{\underline{m}}$  by Proposition 7.

Using Proposition 6, we can characterize Sender's communication strategy as a function of  $\kappa$  and b. To compute  $\hat{w}(\sigma)$  in expression (15), consider each component separately. First, if  $\tau \in \mathcal{F}_{\sigma(\cdot|\mathcal{R})}$ , then  $\operatorname{supp} q \subset \mathcal{R}$  for all  $q \in \operatorname{supp} \tau$  and hence  $q = q(\cdot|\mathcal{R})$ . It follows that (16) simplifies to

$$w^{c}(q) = \kappa(2 - \kappa)(\mathbb{E}_{q}[\omega])^{2} + 2b(\kappa - 1)\mathbb{E}_{q}[\omega] - \mathbb{E}_{q}[\omega^{2}] - b^{2}. \tag{17}$$

Note that  $w^c(\cdot)$  is strictly convex if and only if  $\kappa < 2$ . Therefore, if  $\kappa \leq 2$ ,  $\hat{w}^c(\sigma(\cdot|\mathcal{R}))$  is achieved by  $\tau = \{(\delta_\omega, \sigma(\omega|\mathcal{R}))\}_{\omega \in \mathcal{R}}$  (uniquely if  $\kappa < 2$ ), that is, by fully revealing the states; by contrast, if  $\kappa > 2$ , revealing no information is strictly optimal and  $\hat{w}^c(\sigma(\cdot|\mathcal{R})) = w^c(\sigma(\cdot|\mathcal{R}))$ . Second, for any  $q \in \Delta^d$ , an expression similar to (17) holds for  $w^d(\cdot)$ . Hence, if  $\kappa \leq 2$ ,  $\hat{w}^d(\sigma(\cdot|\overline{\mathcal{R}}))$  is again achieved by  $\tau' = \{(\delta_\omega, \sigma(\omega|\overline{\mathcal{R}}))\}_{\omega \in \overline{\mathcal{R}}}$  (uniquely if  $\kappa < 2$ ), whereas  $\hat{w}^d(\sigma(\cdot|\overline{\mathcal{R}})) = w^d(\sigma(\cdot|\overline{\mathcal{R}}))$  if  $\kappa > 2$ .

Based on these observations, consider first the case with  $\kappa \leq 2$ . In short, Sender always fully reveals all states below  $\omega_{\overline{m}}$ . For higher states, she divides them into two groups depending on Receiver's bias: those close enough to  $\omega_{\overline{m}}$  are always hidden, the others are fully revealed. Finally, the stronger the Receiver's bias, the larger the set of hidden states.

Corollary 6. If  $\kappa \leq 2$ , Sender's communication strategy has the following properties.

- For each  $i \leq \overline{m}$ , she fully reveals  $\omega_i$ .
- For all  $i > j > \overline{m}$ , there exist thresholds  $b_i(\kappa)$  and  $b_j(\kappa)$  decreasing in  $\kappa$  (each strictly when positive) and satisfying  $b_i(\kappa) \leq b_j(\kappa)$  (with < if either is positive). If  $b \leq b_{\overline{m}+1}(\kappa)$ , then  $\hat{w}^d(\sigma(\cdot|\overline{\mathcal{R}})) = \hat{m}(\sigma(\cdot|\overline{\mathcal{R}}))$  and hence  $\tau^d = \sigma(\overline{\mathcal{R}})$ . If  $b > b_{\overline{m}+1}(\kappa)$ , there exists  $i^*(b, \kappa)$ ,

non-decreasing in both  $\kappa$  and b, such that Sender hides with probability 1 each  $\omega_i$  with  $\overline{m} < i < i^*(b, \kappa)$  and fully reveals each  $\omega_i$  with  $i \ge i^*(b, \kappa)$ .

So, when  $\kappa \leq 2$ , the fixed component b of Receiver's bias plays no role in how Sender communicates when she is not hiding states. This point generalizes a similar observation in Kamenica and Gentzkow (2011). By contrast, however, here b affects which states Sender hides and how she optimally hides them.

The case with  $\kappa > 2$  is less straightforward. Since  $w^d$  is strictly concave in q, we always have  $\hat{w}^d(q^d) = w^d(q^d)$  in Proposition 6. Therefore, whenever Sender disproves  $\rho_0$ , she reveals no information and makes Receiver take the same action:  $\mathbb{E}_{q^d}[\beta]$ . It turns out that this action always caters to either the low states or the high states, rather than trying to achieve a compromise.

Corollary 7. If  $\kappa > 2$ , then either  $\mathbb{E}_{\sigma}[\beta | \omega < \omega_{\underline{m}}] \leq \mathbb{E}_{q^d}[\beta] < \beta(\omega_{\underline{m}})$  or  $\mathbb{E}_{q^d}[\beta] > \beta(\omega_{\overline{m}})$ . Also, Sender's expected payoff is

$$-\left\{\sum_{\omega\in\mathcal{R}} (\mathbb{E}_{\sigma}[\beta|\mathcal{R}] - \omega)^2 \sigma(\omega) + \tau^d \mathbb{E}_{q^d} (\mathbb{E}_{q^d}[\beta] - \omega)^2 + \tau^c \sum_{\omega>\omega_{\overline{m}}} \{-u_S^*(\omega)\} q^c(\omega)\right\}$$
(18)

where  $\tau^d = 1 - \tau^c > 0$  and  $\tau^c q^c + \tau^d q^d = \sigma$ . If  $\omega_{\underline{m}} \leq \mathbb{E}_{\sigma}[\omega | \overline{\mathcal{R}}] \leq \omega_{\overline{m}}$ , then some  $\omega > \omega_{\overline{m}}$  are hidden with positive probability.

To gain intuition, suppose  $\mathbb{E}_{q^d}[\beta]$  were between  $\beta(\omega_{\underline{m}})$  and  $\beta(\omega_{\overline{m}})$ , the highest and lowest action Receiver would take if his theory is confirmed. Then, for all  $\omega < \omega_{\underline{m}}$ , Sender would strictly prefer to hide them. But if she does so,  $\mathbb{E}_{q^d}[\beta]$  jumps above  $\beta(\omega_{\overline{m}})$ . She can then combine hiding and surprising for states above  $\omega_{\overline{m}}$  to achieve the best  $\mathbb{E}_{q^d}[\beta]$ , thus improving her payoff. A similar improvement may start from hiding states  $\omega > \omega_{\overline{m}}$ , leading to  $\mathbb{E}_{q^d}[\beta] < \beta(\omega_{\underline{m}})$ . Overall, given  $\kappa$  and b, whether it is best for Sender to cater to high or to low states with  $\mathbb{E}_{q^d}[\beta]$  ultimately depends on which states she thinks are more likely.

Sender's expected payoff in (18) contains further information on the actions she persuades Receiver to choose. For states in  $\mathcal{R}$ , he chooses  $\mathbb{E}_{\sigma}[\beta|\mathcal{R}]$  with probability arbitrarily close to 1 (which is exactly 1 if  $\tau^d = \sigma(\overline{\mathcal{R}})$ ). Sender 'uses' the remaining probability to hide states above  $\omega_{\overline{m}}$ , in which case Receiver's action is in  $[\beta(\omega_{\underline{m}}), \beta(\omega_{\overline{m}})]$ , but differs in general from  $\mathbb{E}_{\sigma}[\beta|\mathcal{R}]$ . If  $\mathbb{E}_{\sigma}[\omega|\overline{\mathcal{R}}] \in [\omega_{\underline{m}}, \omega_{\overline{m}}]$ —i.e.,  $\sigma(\cdot|\overline{\mathcal{R}})$  is not too concentrated on high or low states—then Sender will hide some  $\omega > \omega_{\overline{m}}$ . In other words, she makes Receiver's action cater to low or high states by optimally hiding high states. Note that these qualitative properties as well as those in Corollary 7 are independent of  $\kappa$  and b.

## 5.3 Is Having a Richer Theory Always Good for Persuaders?

With the help of an example, this section aims to illustrate two simple points: (1) Sender may be strictly better off if Receiver shared her theory of the world rather than viewing some states as impossible; (2) Sender may reveal some information in the former scenario but not in the latter.<sup>44</sup> The key to these points is Proposition 2: by treating some states as impossible, Receiver's theory can severely limit how Sender can influence his beliefs and hence actions.

Consider a setting with two states:  $S = \{\omega_1, \omega_2\}$ . Sender has prior  $\sigma = (\frac{1}{2}, \frac{1}{2})$ . Receiver has four actions,  $A = \{a, b, c, d\}$ , and payoffs are as follows:

$u_S(\cdot,\cdot)$	$\omega_1$	$\omega_2$
a	0	5
b	3	1
c	1	4
d	2	2

$u_R(\cdot,\cdot)$	$\omega_1$	$\omega_2$
a	2	-3
b	1	0
c	0	1
d	-3	2

The next table summarizes Receiver's best actions depending on his posterior p:

Values of 
$$p(\omega_2)$$
  $\begin{bmatrix} 0, \frac{1}{4} \end{bmatrix}$   $\begin{bmatrix} \frac{1}{4} \end{bmatrix}$   $\begin{bmatrix} \frac{1}{4}, \frac{1}{2} \end{bmatrix}$   $\begin{bmatrix} \frac{1}{2} \end{bmatrix}$   $\begin{bmatrix} \frac{1}{2}, \frac{3}{4} \end{bmatrix}$   $\begin{bmatrix} \frac{3}{4} \end{bmatrix}$   $\begin{bmatrix} \frac{3}{4}, 1 \end{bmatrix}$   $\mathcal{A}(p)$   $\{a\}$   $\{a, b\}$   $\{b\}$   $\{b, c\}$   $\{c\}$   $\{c, d\}$   $\{d\}$ 

If  $\rho_0 = (0,1)$ , Sender either always confirms it—hence, p = (0,1)—or she disproves it in state  $\omega_1$ —hence, p = (1,0). The expected payoff from these strategies are 2 and 1 respectively, so it is optimal for Sender not to communicate at all.

Now suppose that  $\rho_0 = \sigma$  as in Kamenica and Gentzkow (2011). In this case, Sender's and Receiver's posteriors coincide for any message: p(q) = q for all  $q \in \Delta(\mathcal{S})$ . Letting  $q_2 = q(\omega_2)$ , Sender's expected payoff for any q is then

$$w^{KG}(q) = \begin{cases} 5q_2 & \text{if } q_2 \in [0, \frac{1}{4}) \\ 3 - 2q_2 & \text{if } q_2 \in [\frac{1}{4}, \frac{1}{2}) \\ 1 + 3q_2 & \text{if } q_2 \in [\frac{1}{2}, \frac{3}{4}] \\ 2 & \text{if } q_2 \in (\frac{3}{4}, 1] \end{cases}.$$

By Corollary 2 in Kamenica and Gentzkow (2011), her expected payoff from an optimal  $\pi$  is  $W_{\sigma}^{KG} = \hat{w}^{KG}(\sigma)$  and she benefits from revealing information if and only if  $\hat{w}^{KG}(\sigma) > 0$ 

<sup>&</sup>lt;sup>44</sup>When Sender's and Receiver's have common-support but different priors, Alonso and Câmara (2013) show that reveling some information is generically optimal for Sender.

 $w^{KG}(\sigma)$ . It is easy to see that here

$$\hat{w}^{KG}(q) = \begin{cases} 10q_2 & \text{if } q_2 \in [0, \frac{1}{4}) \\ \frac{17}{8} + \frac{3}{2}q_2 & \text{if } q_2 \in [\frac{1}{4}, \frac{3}{4}) \\ 7 - q_2 & \text{if } q_2 \in (\frac{3}{4}, 1] \end{cases}$$

Therefore,  $\hat{w}^{KG}(\frac{1}{2}, \frac{1}{2}) = 2.875$  whereas  $w^{KG}(\frac{1}{2}, \frac{1}{2}) = 2.5$ . Moreover,  $W_{\sigma}^{KG}$  exceeds Sender's expected payoff when Receiver deems  $\omega_1$  impossible  $(W_{\sigma} = 2)$ .

This example raises an intriguing issue for the literature on persuasion. Starting from a situation in which Receiver's theory is  $\rho_0 = (0, 1)$ , Sender would like to first 'persuade' him to adopt her theory  $\sigma$  and only then reveal some information on the states. We have seen, however, that here no signal device in  $\Pi$ —the class commonly studied in the literature—can make Receiver switch from  $\rho_0$  to  $\sigma$ . So, if there is any technique that would make him switch theories (perhaps arguments on their internal logical consistency), it must belong to a different class of persuasion activities than those usually studied in the literature. This class seems to deserve further investigation.

# 6 Extensions and Discussion

# 6.1 Unawareness Interpretation of the Model

The underlying reason why Receiver's prior  $\rho_0$  assigns zero probability to some states of the world (from Sender's viewpoint) may be that he is initially unaware of them. Thus one may interpret this paper's model as *one* way—certainly not the only one—to describe situations in which Sender faces a Receiver who is initially unaware of the states in  $\overline{\mathcal{R}}$ . This interpretation is possible, perhaps even natural, but involves several assumptions and requires further explanation.

The analysis in Sections 4 and 5 applies unchanged to the unawareness case under the following conditions. We can continue to assume that Sender commits to signal devices. It is also reasonable to assume that, though initially unaware of  $\omega$ , Receiver can fully understand its description once it is given to him.<sup>45</sup> Thus, when  $\pi$  is disclosed to Receiver before any signal realizes, he understands that he may have been unaware of some conceivable states. Nonetheless, we can assume that, after observing only  $\pi$ , Receiver does not abandon his prior yet. This may happen for several reasons: (1) Per se  $\pi$ 

<sup>&</sup>lt;sup>45</sup>This description can also exhaustively specify Receiver's payoff in  $\omega$ ; for example, it describes how actions map to monetary prizes, for which he knows his utility function. Therefore we can continue to define  $u_R$  as a function of  $\Omega$ , not just  $\mathcal{R}$ .

contains no information; (2) Receiver may be firmly skeptic and think that Sender simply invented states that are incompatible with his theory, and hence he should not abandon it unless conclusive evidence disproves it; (3) Receiver may always try to interpret any observation in favor of his initial theory, as shown in studies on "confirmatory bias." Finally, we can continue to assume that Receiver responds to unexpected messages as described by Ortoleva's (2012) axioms. These axioms correspond to a model which involves a prior over priors assigning positive probability to states of which Receiver is initially unaware. This model is an as-if representation, however, which does not literally presumes that Receiver has such a prior over priors from the outset—or, for that matter, at any time.

The assumption that Receiver does not abandon  $\rho_0$  after observing a signal device is key for this paper's analysis to apply to the unawareness case. Another possibility, of course, is that as soon as Receiver hears the description of a state  $\omega \notin \mathcal{R}$ , he abandons his initial theory of the world and adopts one that includes  $\omega$ . If we maintain the assumption that, by construction, every  $\pi \in \Pi$  must describe all states in  $\Omega$ , then Sender will design  $\pi$  as if Receiver already assigns positive probability to all states. As a result, any possibility of hiding states and surprising Receiver disappears. The full-support prior that Receiver adopts after observing  $\pi$  may be fixed or vary with  $\pi$  itself. In the first case, the model essentially collapses to one with common-support priors as analyzed by Kamenica and Gentzkow (2011) and Alonso and Câmara (2013). In the second case, for every  $\pi$  the distribution over Receiver's posteriors is always pinned down by Bayes' rule, but can no longer be independent of  $\pi$ . This raises issues that are beyond this paper's scope.

In the unawareness case—but also more generally—a natural concern is that after observing a signal Sender may want to conceal some part of her device. In this way, Sender can literally hide states Receiver is unaware of. Formally, after observing x from  $\pi$ , Sender may want to communicate a message  $(x, \pi')$  where  $\pi'$  is obtained by deleting  $\pi(\cdot|\omega)$  for some  $\omega$ . If  $(x,\pi)$  does not make Receiver abandon  $\rho_0$  and  $\omega \in \overline{\mathcal{R}}$ , such a change does not benefit Sender: Receiver ignores  $\pi(x|\omega)$  anyways. In other cases, however, the change may benefit Sender: she can prevent Receiver from becoming aware of some states, after gaining better information for herself. Concealing parts of  $\pi$  is conceptually different from concealing some signal realization (e.g., by lumping signals together in a coarser one) and can have different consequences. Their analysis is left for future research.

<sup>&</sup>lt;sup>46</sup>See, e.g., Rabin and Schrag (1999) and references therein.

## 6.2 Alternative Models of Receiver's Responses to Information

This section examines a set of assumptions describing other tractable ways to model how Receiver responds to unexpected evidence. These models are less sophisticated than Ortoleva's (2012). They are, however, simpler to the extent that they do not involve a prior over priors—which should be part of what Sender knows about Receiver—and assume a direct procedure by which Receiver picks new theories after surprises. These models rely on *lexicographic belief systems* (LBS's) and are similar in spirit to the 'sequences of hypotheses' in Kreps and Wilson's (1982) work on sequential equilibria. They may also fit better an unawareness interpretation of different-support priors.

The first assumption describes how Receiver responds to messages confirming  $\rho_0$  and has the same rationale as A1(c).

**Assumption 3** (A3). If  $x \in C_{\pi}$ , Receiver updates  $\rho_0$  using Bayes' rule.

To describe how Receiver responds to messages disproving  $\rho_0$ , we consider two possibilities. In both cases, we shall continue to assume that Receiver forms a unique posterior belief.<sup>47</sup> To do so, after a surprising x, he updates some prior other than  $\rho_0$  which is 'triggered' by x. We formalize this idea using lexicographic belief systems (LBS's).

In the first assumption, Receiver has only one alternative theory of the world which contains all states in  $\Omega$ .

**Assumption 4** (A4: Binary LBS). Receiver has an LBS  $(\rho_0, \rho_1)$  with supp  $\rho_1 = \Omega$ . If  $x \in D_{\pi}$ , he updates  $\rho_1$  using Bayes' rule.

To illustrate, suppose  $\Omega = \{\omega_1, \omega_2, \omega_3\}$ ,  $\mathcal{R} = \{\omega_1\}$ , and that x can arise only in  $\omega_2$  under  $\pi$ , i.e.,  $\Omega_{\pi}(x) = \{\omega_2\}$ . Given  $(x, \pi)$ , first of all Receiver switches to viewing  $\omega_1$ ,  $\omega_2$ , and  $\omega_3$  as possible; he then updates his new theory using Bayes' rule. A4 captures a Receiver who is willing to easily abandon his theory of the world and, given evidence contradicting it, admits that all states in  $\Omega$  are actually possible. Together, A3 and A4 are similar to Ortoleva's (2012) model with  $\mu(\rho_0) > \frac{1}{2}$  and  $\mu(\rho_1) = 1 - \mu(\rho_0)$ . His model, however, suggests a richer story for why Receiver first adopts  $\rho_0$  and then turns to  $\rho_1$  after unexpected signals.

The next assumption allows for richer LBS's. Recall that  $\Omega_{\pi}(x)$  is the set of states that are consistent with  $(x, \pi)$  (see (3)).

<sup>&</sup>lt;sup>47</sup>Alternatively, one could imagine that surprising signals may leave Receiver with some ambiguity, captured by a set of posteriors.

**Assumption 5** (A5: Gradual LBS). Receiver has LBS  $(\rho_0, \ldots, \rho_N)$  such that, for each  $\Omega_i \subset \Omega$  with  $\Omega_i \supseteq \mathcal{R}$ , there is exactly one  $\rho_i$  with  $\operatorname{supp} \rho_i = \Omega_i$ .<sup>48</sup> If  $x \in D_{\pi}$ , he updates prior  $\rho_i$  with  $\operatorname{supp} \rho_i = \mathcal{R} \cup \Omega_{\pi}(x)$  using Bayes' rule.

A5 captures a Receiver who is reluctant to abandon his theory. Given evidence contradicting it, he is willing to expand it to include only those states which are either possible under  $\rho_0$  or consistent with  $(x, \pi)$ . Suppose again that  $\Omega = \{\omega_1, \omega_2, \omega_3\}$ ,  $\mathcal{R} = \{\omega_1\}$ , and  $\Omega_{\pi}(x) = \{\omega_2\}$ . Given  $(x, \pi)$ , now Receiver's new theory views only  $\omega_1$  and  $\omega_2$  as possible.

In general, we may require some consistency between layers in Receiver's LBSs. 49

**Assumption 6** (A6: Consistency). Given LBS  $(\rho_0, \ldots, \rho_N)$ , for any  $\rho_i$  and  $\rho_j$  and corresponding supports  $\Omega_i$  and  $\Omega_j$ ,

$$\frac{\rho_i(\omega|\Omega_i\cap\Omega_j)}{\rho_i(\omega'|\Omega_i\cap\Omega_j)} = \frac{\rho_j(\omega|\Omega_i\cap\Omega_j)}{\rho_j(\omega'|\Omega_i\cap\Omega_j)} \quad \text{for all } \omega,\omega'\in\Omega_i\cap\Omega_j.$$

That is, for any  $\rho_i$  and  $\rho_j$ , Receiver assigns the same relative likelihood to all states that are possible under both theories. For example, let  $\Omega = \{\omega_1, \omega_2, \omega_3, \omega_4\}$ ,  $\rho_0 = (\frac{1}{2}, \frac{1}{2}, 0, 0)$ , and  $\rho_1$  have support  $\Omega_1 = \{\omega_1, \omega_2, \omega_3\}$ . Then, conditional on  $\{\omega_1, \omega_2\}$ ,  $\rho_1$  assigns equal probability to the two states. A6 seems reasonable since each state is an exhaustive description of reality. If  $\omega_1$  and  $\omega_2$  can occur only when neither  $\omega_3$  nor  $\omega_4$  occurs, then the assessment of  $\omega_1$  and  $\omega_2$ 's relative likelihood knowing that  $\{\omega_3, \omega_4\}$  is impossible should be the same as when knowing that  $\{\omega_3, \omega_4\}$  was possible but did not occur.

Under A3-A6, we can characterize the joint distributions over posteriors Sender can achieve with signal devices, along the lines of Section 4.2. The analysis in Section 5 then applies unchanged.

The key step is to realize that Sender's posterior after any  $(x, \pi)$  pins down which prior Receiver will update among those in his LBS. Under A4 this is immediate: for every  $x \in D_{\pi}$ ,  $q(\cdot|x,\pi) \in \Delta^d$  and Receiver updates  $\rho_1$ . Consider now A5. For every  $x \in D_{\pi}$ ,  $\operatorname{supp} q(\cdot|x,\pi)$  equals the set  $\Omega_{\pi}(x)$  of states consistent with  $(x,\pi)$ . Given  $(x,\pi)$ , Receiver updates  $\rho_i$  with support  $\mathcal{R} \cup \Omega_{\pi}(x)$ . So, for every  $x \in D_{\pi}$ , let

$$\Omega(q(\cdot|x,\pi)) = \mathcal{R} \cup \operatorname{supp} q(\cdot|x,\pi).$$

This mapping from Sender's posteriors to subsets of  $\Omega$  is well defined:  $\operatorname{supp} q(\cdot|x,\pi) = \operatorname{supp} q(\cdot|y,\pi)$  if  $q(\cdot|x,\pi) = q(\cdot|y,\pi)$  for  $x,y \in D_{\pi}$ . Then, for every  $\tau \in \mathcal{F}_{\sigma}$  and  $\rho_i$  in

<sup>&</sup>lt;sup>48</sup>To clarify, N in Receiver's LBS equals the number of subsets of  $\Omega$  containing  $\mathcal{R}$  as a strict subset.

<sup>&</sup>lt;sup>49</sup>Karni and Vierø (2013) study a decision-theoretic model of growing awareness and provide an axiomatic foundation for this consistency property. This property should not be confused with the notion of consistency in Kreps and Wilson (1982).

Receiver's LBS, let  $D_{\tau}(\rho_i) \subset D_{\tau}$  be the set of Sender's posteriors at which Receiver updates  $\rho_i$ :

$$D_{\tau}(\rho_i) = \{ q \in \operatorname{\mathbf{supp}} \tau : \Omega(q) = \operatorname{\mathbf{supp}} \rho_i \}.$$

Note that  $\{D_{\tau}(\rho_i)\}_{i=1}^N$  forms a partition of  $D_{\tau}$ . Relying on Lemma 1, we may then draw the following conclusion.

**Proposition 8.** Consider any  $\tau \in \mathcal{F}_{\sigma}$  and let Receiver's LBS be  $(\rho_0, \ldots, \rho_N)$ . Then, for every  $q \in \text{supp } \tau$  and  $\omega \in \Omega$ , we have

$$p(\omega;q) = \frac{q(\omega)\frac{\rho(\omega;q)}{\sigma(\omega)}}{\sum_{\omega' \in \Omega} q(\omega')\frac{\rho(\omega';q)}{\sigma(\omega')}},$$
(19)

where under A4 (i.e., N = 1)

$$\rho(q) = \begin{cases} \rho_0 & \text{if } q \in C_\tau \\ \rho_1 & \text{if } q \in D_\tau \end{cases},$$

and under A5

$$\rho(q) = \begin{cases} \rho_0 & \text{if } q \in C_\tau \\ \rho_i & \text{if } q \in D_\tau(\rho_i) \end{cases}.$$

By Proposition 8, if A6 holds, Sender can achieve the same joint distributions over posteriors under A4 and A5. Therefore, considering the simpler case of A4 involves no loss of generality for the purpose of this paper.

Corollary 8. Consider LBS  $(\rho_0^A, \rho_1^A)$  under A4 and LBS  $(\rho_i^B)_{i=0}^N$  under A5. Suppose that  $\rho_0^A = \rho_0^B$ ,  $\rho_1^A = \rho_N^B$  with supp  $\rho_N^B = \Omega$ , and  $(\rho_i^B)_{i=0}^N$  satisfies A6. Then, for any  $\tau \in \mathcal{F}_{\sigma}$ ,

$$p^{A}(\omega;q) = p^{B}(\omega;q)$$
 for all  $\omega \in \Omega$  and  $q \in \operatorname{supp} \tau$ .

Proposition 8 also implies that under A4 p(q) varies continuously in q over the sets  $\Delta^c$  and  $\Delta^d$  separately. By Corollary 8, the same is true under A5 if A6 holds.

Without A6, A4 and A5 can lead to different sets of joint distributions over posteriors and Receiver's posterior may be discontinuous over  $\Delta^d$ . To see this, suppose that  $\mathcal{S} = \{\omega_1, \omega_2, \omega_3, \omega_4\}$ ,  $\mathcal{R} = \{\omega_1\}$ ,  $\sigma = \rho_1^A = \rho_7^B = (\frac{1}{4}, \frac{1}{4}, \frac{1}{4}, \frac{1}{4})$ , and  $\rho_6^B = (\frac{1}{4}, \frac{1}{4}, \frac{1}{2}, 0)$ . Let  $q = (0, \frac{1}{2}, \frac{1}{2}, 0) \in \Delta^d$ . Then,  $\rho(q) = \rho_1^A$  under A4, but  $\rho(q) = \rho_6^B$  under A5. Straightforward calculations yield

 $p^{A}(\omega_{3};q) = \frac{1}{2}$  and  $p^{B}(\omega_{3};q) = \frac{2}{3}$ .

For  $\varepsilon \in (0,1)$ , consider  $q_{\varepsilon} = (0,\frac{1-\varepsilon}{2},\frac{1-\varepsilon}{2},\varepsilon) \in \Delta^d$ . Then, under both A4 and A5, we have

$$p^{A}(\omega_{3};q_{\varepsilon})=p^{B}(\omega_{3};q_{\varepsilon})=\frac{1-\varepsilon}{2},$$

which converges to  $\frac{1}{2}$  as  $\varepsilon \to 0$ .

## 6.3 Different Supports as the Limit of Common Supports?

This section shows that, in general, it is not correct to view the case of different-support priors as a limit of the case of common-support priors in which the probability that Receiver's prior assigns to states in  $\overline{\mathcal{R}}$  converges to zero. The reason of this "lack of continuity" can be understood from the properties of the mapping from Sender's to Receiver's posteriors in the two cases. In the common-support case, no matter how small the probability that Receiver assigns to  $\overline{\mathcal{R}}$  is, Sender can induce Receiver to have any posterior in  $\Delta(\Omega)$  and hence take the corresponding actions. In the different-support case, by contrast, Sender can induce Receiver to have posteriors only in a "small" subset of  $\Delta(\Omega)$ , and hence he may never take some actions that he would instead choose in the other case for some posterior. As a result, in the limit we can have a discontinuous change in Sender's communication strategy, in her expected payoff, and in Receiver's induced behavior.

To see this more formally, consider the following simple example. Let  $\Omega = \{\omega_1, \omega_2\}$ ,  $\sigma = (\frac{1}{2}, \frac{1}{2})$ , and  $\rho_0 = (1 - \varepsilon, \varepsilon)$  for  $\varepsilon > 0$ . Let  $A = \{a, b, c\}$  and the payoff functions be

$u_S(\cdot,\cdot)$	$\omega_1$	$\omega_2$
a	1	1
b	0	0
c	-1	-1

$u_R(\cdot,\cdot)$	$\omega_1$	$\omega_2$
a	1	1
b	2	-1
c	-1	2

Slightly abusing notation, let p and q be the posterior probabilities that Receiver and Sender assign to  $\omega_2$ , so that  $p(q) = \frac{q\varepsilon}{q\varepsilon + (1-q)(1-\varepsilon)}$ . Clearly, Sender can induce Receiver to choose a as long as  $p \in \left[\frac{1}{3}, \frac{2}{3}\right]$ ; otherwise, he chooses b when  $p < \frac{1}{3}$  and c when  $p > \frac{2}{3}$ . Straightforward calculations imply that  $p(q) \in \left[\frac{1}{3}, \frac{2}{3}\right]$  if and only if  $q \in \left[\underline{q}_{\varepsilon}, \overline{q}_{\varepsilon}\right]$ , where  $\underline{q}_{\varepsilon} = \frac{1-\varepsilon}{1+\varepsilon}$  and  $\overline{q}_{\varepsilon} = \frac{2(1-\varepsilon)}{2-\varepsilon}$ . Note that  $0 < \underline{q}_{\varepsilon} < \overline{q}_{\varepsilon} < 1$  for every  $\varepsilon > 0$  and both thresholds are strictly increasing and converge to 1 as  $\varepsilon \to 0$ . This means that, no matter how small  $\varepsilon$  is, Sender can always generate a posterior q such that at the corresponding p Receiver chooses action a. In particular, it is easy to see that when  $\underline{q}_{\varepsilon} > \frac{1}{2}$ , Sender's optimal  $\tau$  induces two posteriors—one with q = 0 and the other with  $q = \underline{q}_{\varepsilon}$ —with  $\tau(\underline{q}_{\varepsilon}) = \frac{1}{2q_{\varepsilon}}$  and  $\tau(0) = 1 - \tau(\underline{q}_{\varepsilon})$ . Thus, for every  $\varepsilon > 0$  such that  $\underline{q}_{\varepsilon} > \frac{1}{2}$ , Sender's expected payoff from  $\tau$  is  $\frac{1}{2q_{\varepsilon}} > 0$ , which converges to  $\frac{1}{2}$  as  $\varepsilon \to 0$ ; moreover, she always induces Receiver to take both a and b with positive probability.

Consider now the case of different-support priors corresponding to the limit of the previous case as  $\varepsilon \to 0$  (i.e.,  $\mathcal{R} = \{\omega_1\}$ ). Now Sender cannot induce Receiver to entertain any posterior with  $p \in \left[\frac{1}{3}, \frac{2}{3}\right]$  and hence to take action a. Moreover, since Sender strictly

prefers action b to c, her best signal device is a completely uninformative one (i.e.,  $\tau = \delta_{\sigma}$ ), which gives her a payoff of 0. Thus neither Sender's optimal communication strategy, nor her payoff, nor Receiver's induced behavior coincide between the case of different-support priors and the corresponding limit of the case with common-support priors.

## 6.4 Alternative Differences in Priors' Supports

The support Receiver's prior need not be a subset of that of Sender's prior  $(\mathcal{R} \subsetneq \mathcal{S})$ . This case, however, comprises all key aspects of the different-support assumption. For  $\mathcal{S}, \mathcal{R} \subset \Omega$ , consider the alternatives (i)  $\mathcal{S} \subsetneq \mathcal{R}$ , (ii)  $\mathcal{S} \cap \mathcal{R} \neq \emptyset$  but  $\mathcal{S} \not\subset \mathcal{R}$  and  $\mathcal{R} \not\subset \mathcal{S}$ , and (iii)  $\mathcal{S} \cap \mathcal{R} = \emptyset$ . Case (i) is trivial and uninteresting for our purposes, as Receiver can never be surprised and Bayes' rule always applies. This case may give rise to other interesting phenomena of persuasion that would be impossible if  $\mathcal{R} = \mathcal{S}$ , but their analysis belongs to a separate paper. For the other cases, suppose first that Sender cannot provide information on states outside  $\mathcal{S}$ . In this setting, we can add a default signal—'no evidence'—which arises whenever  $\omega \notin \mathcal{S}$ . Then, case (ii) is equivalent to  $\mathcal{R} \subsetneq \mathcal{S}$  because Sender cannot affect Receiver's beliefs for states outside  $\mathcal{S}$  and thinks that such states are impossible. Similar considerations apply for case (iii), except that Receiver will always be surprised. If Sender can provide information on states outside  $\mathcal{S}$ , in both case (ii) and (iii) we can always rely on Bayes' rule whenever Receiver is not surprised; when he is, the problem is the same as with  $\mathcal{R} \subsetneq \mathcal{S}$ .

# 7 Conclusion

This paper analyzes persuasion when persuader and audience disagree on their subjective theory of the possible states of the world. It derives the set of joint distributions over posterior beliefs that the persuader can induce for reasonable models of how the audience responds to unexpected information. The properties of this set, which differ significantly from the standard case with agreement on the possible states, highlight that the audience's different theory can seriously limit the room for persuasion.

A key insight of the paper comes from identifying the optimal strategy to 'hide' the states that the audience deems impossible. Its reluctance to abandon its theory in light of inconclusive evidence can allow the persuader to hide those states by pooling them with signals on states it deems possible, while simultaneously manipulating its beliefs on these states to her best advantage.

The paper finally characterizes the persuader's optimal communication strategies, showing when and how she chooses to hide states or surprise her audience. When considered in isolation, some states may seem clearly unfavorable to the persuader and hence we may conclude that she should always hide them. This is not the right way to thinks about the persuader's problem, however. As part of an overall persuasion strategy, revealing unfavorable states can help her increase the probability that the audience takes actions which benefit her.

# A Appendix: Discontinuity of Receiver's Posterior over $\Delta^d$ under Assumption 1

Given  $\Omega = \{\omega_1, \omega_2, \omega_3, \omega_4\}$  and  $\sigma = (\frac{1}{4}, \frac{1}{4}, \frac{1}{4}, \frac{1}{4})$ , let  $\sup \mu = (\rho_0, \rho_1, \rho_2)$  with  $\mu(\rho_0) = \frac{1}{2}$ ,  $\mu(\rho_1) = \mu(\rho_2) = \frac{1}{4}$ ,  $\mathcal{R} = \{\omega_1\}$ ,  $\rho_1 = (\frac{1}{4}, \frac{1}{4}, \frac{1}{2}, 0)$ , and  $\rho_2 = \sigma$ . Consider Sender's posterior  $q_z = (0, \frac{1-z}{2}, \frac{1-z}{2}, z) \in \Delta^d$  for  $z \in (0, 1)$ . Then,

$$\mu(\rho_1; q_z) = \frac{\sum_{\omega \in \Omega} q_z(\omega) \rho_1(\omega)}{\sum_{\omega \in \Omega} q_z(\omega) \rho_1(\omega) + \sum_{\omega \in \Omega} q_z(\omega) \rho_2(\omega)} = \frac{1}{1 + \frac{2}{3(1-z)}},$$

$$\mu(\rho_2; q_z) = 1 - \mu(\rho_1; q_z).$$

Hence,  $\mu(\rho_1; q_z) \ge \mu(\rho_2; q_z)$  if and only if  $z \le \frac{1}{3}$ . For  $z = \frac{1}{3}$  Receiver will choose either  $\rho_1$  or  $\rho_2$  depending on how he ranks them under  $\succ$ .

Using Lemma 1, we can compute Receiver's posteriors starting with  $\rho_1$  and  $\rho_2$  when Sender has posterior  $q_z$ . Focusing on  $\omega_3$ , we have

$$p_1(\omega_3; q_z) = \frac{(1-z)\frac{1}{2}}{(1-z)\frac{1}{2} + (1-z)\frac{1}{4}} = \frac{2}{3},$$

$$p_2(\omega_3; q_z) = \frac{(1-z)\frac{1}{4}}{(1-z)\frac{1}{4} + (1-z)\frac{1}{4} + 2z\frac{1}{4}} = \frac{1-z}{2}.$$

So Receiver's posterior must vary discontinuously in  $q_z$  at  $z = \frac{1}{3}$ .

# B Appendix: Proofs

#### B.1 Proof of Proposition 2

If  $q(\cdot|x,\pi) \in \Delta^c$ , then  $p(\cdot|x,\pi)$  results from updating  $\rho_0 \in \Delta(\mathcal{R})$  using Bayes' rule. Hence,  $\operatorname{supp} p(\cdot|x,\pi) \subset \mathcal{R}$ . If  $q(\cdot|x,\pi) \in \Delta^d$ , then under A1, we have that  $p(\cdot|x,\pi)$  results from updating some prior  $\rho \in \Delta(\Omega)$  using Bayes' rule and hence  $\operatorname{supp} p(\cdot|x,\pi) \subset \Omega_{\pi}(x)$ . Finally, by definition of  $\Delta^d$ , we have  $\Omega_{\pi}(x) = \operatorname{supp} q(\cdot|x,\pi) \subset \overline{\mathcal{R}}$ .

#### B.2 Proof of Lemma 1

This steps generalize the proof of Proposition 1 in Alonso and Câmara (2013). Take any  $\pi$ ,  $x \in X_{\pi}$ , and  $\rho \in \Delta(\Omega)$  such that  $\sum_{\omega' \in \Omega} \pi(x|\omega')\rho(\omega') > 0$ . Applying Bayes' rule, we get

$$p(\omega|x,\pi) = \frac{\pi(x|\omega)\rho(\omega)}{\sum_{\omega' \in \Omega} \pi(x|\omega')\rho(\omega')} \quad \text{for all } \omega \in \Omega.$$

For any  $\omega \in \Omega$ , since  $\sigma(\omega) > 0$ , using (2) we can write

$$\pi(x|\omega)\rho(\omega) = q(\omega|x,\pi)\frac{\rho(\omega)}{\sigma(\omega)} \left[ \sum_{\omega' \in \Omega} \pi(x|\omega')\sigma(\omega') \right].$$

Since the term in brackets is a constant for every x, substituting and simplifying, we obtain that

$$p(\omega|x,\pi) = \frac{q(\omega|x,\pi)\frac{\rho(\omega)}{\sigma(\omega)}}{\sum_{\omega' \in \Omega} q(\omega'|x,\pi)\frac{\rho(\omega')}{\sigma(\omega')}} \quad \text{for all } \omega \in \Omega.$$

#### B.3 Proof of Corollary 2

By Proposition 2, supp  $p(q) \cap \text{supp } p(q') = \emptyset$ . Therefore,

$$\left|\left|p(q)-p(q')\right|\right|^2 = \sum_{\omega \in \mathcal{R}} |p(\omega;q)|^2 + \sum_{\omega \in \overline{\mathcal{R}}} \left|p(\omega;q')\right|^2 > \sum_{\omega \in \overline{\mathcal{R}}} \left|p(\omega;q')\right|^2 > 0.$$

#### B.4 Proof of Lemma 2

For  $q \in \Delta(\mathcal{S})$ , we have  $w(q) > -\infty$  by continuity of  $u_S$  and compactness of A. For all  $q \in \mathbb{R}^{|\mathcal{S}|-1} \setminus \Delta(\mathcal{S})$  define  $w(q) = -\infty$ . By Carathéodory's Theorem (see Rockafellar (1997), Corollary 17.1.5),

$$\hat{w}(\sigma) = \sup_{T_{\sigma}} \sum_{m} \tau_{m} w(q_{m}),$$

where

$$T_{\sigma} = \{\{(q_m, \tau_m)\}_{m=1}^{|\mathcal{S}|} : \sum_{m=1}^{|\mathcal{S}|} \tau_m q_m = \sigma, \sum_{m=1}^{|\mathcal{S}|} \tau_m = 1, \tau_m \ge 0, q_m \in \Delta(\mathcal{S}), \forall m\}.$$

Since  $T_{\sigma} \subset \mathcal{F}_{\sigma}$ , it follows that  $\hat{w}(\sigma) \leq W_{\sigma}$ . By definition of  $W_{\sigma}$ , for every  $\varepsilon > 0$ , there exists  $\tau_{\varepsilon} \in \mathcal{F}_{\sigma}$  such that  $V(\tau_{\varepsilon}) \geq W_{\sigma} - \varepsilon$ . However,  $V(\tau_{\varepsilon}) \in \{\xi : (\sigma, \xi) \in \operatorname{co}(\operatorname{hyp} w)\}$  and hence  $V(\tau_{\varepsilon}) \leq \hat{w}(\sigma)$ . So for every  $\varepsilon > 0$ ,  $\hat{w}(\sigma) \geq W_{\sigma} - \varepsilon$  which implies that  $\hat{w}(\sigma) \geq W_{\sigma}$ .

#### B.5 Proof of Lemma 3

The first part follows from Lemma 2. For the second part, note that by the same argument as in the proof of Lemma 2,

$$\hat{w}^c(\sigma) = \sup_{T_\sigma^c} \sum_m \tau_m w(q_m),$$

where

$$T_{\sigma}^{c} = \{\{(q_{m}, \tau_{m})\}_{m=1}^{N} : N \ge 1, \sum_{m=1}^{N} \tau_{m} q_{m} = \sigma, \sum_{m=1}^{N} \tau_{m} = 1, \tau_{m} \ge 0, q_{m} \in \Delta^{c}, \forall m\}.$$

Suppose  $V(\tau) > \hat{w}^c(\sigma)$  but  $\tau(D_\tau) = 0$ . Since  $\tau \in \mathcal{F}_\sigma$ ,  $|\mathbf{supp}\,\tau| = N$  for some finite N and hence  $D_\tau = \emptyset$ . Therefore,  $\tau \in T^c_\sigma$  and hence  $V(\tau) \leq \hat{w}^c(\sigma)$ . A contradiction.

#### B.6 Proof of Proposition 4

Part (1): Take any  $\tau \in \mathcal{F}_{\sigma}^{c}$  with  $q \in \operatorname{supp} \tau$  such that  $q(\omega) > 0$  and  $q(\omega') > 0$  for some  $\omega, \omega' \in \overline{\mathcal{R}}$  with  $\omega \neq \omega'$ . We will show that there exists another  $\tau_{1} \in \mathcal{F}^{c}$  which satisfies the first part of (1) and such that  $V(\tau_{1}) \geq V(\tau)$ . Since  $\tau \in \mathcal{F}_{\sigma}^{c}$ , there exists  $\pi \in \Pi$  that induces  $\tau$ , i.e., for every  $q \in \operatorname{supp} \tau$  there exists x such that  $q = q(\cdot|x,\pi)$  induced by  $\pi$  through Bayes' rule. For every  $q \in \operatorname{supp} \tau$ , let  $\overline{\mathcal{R}}(q) = \{\omega \in \overline{\mathcal{R}} : q(\omega) > 0\}$ . By assumption,  $|\overline{\mathcal{R}}(\hat{q})| > 1$  for some  $\hat{q} \in \operatorname{supp} \tau$ .

For any such  $\hat{q}$  do the following. Let  $\hat{x}$  be the signal inducing it under  $\pi$ , i.e.,  $\hat{q} = q(\cdot|\hat{x}, \pi)$ . Clearly,  $\pi(\hat{x}|\omega) > 0$  if and only if  $\omega \in \text{supp } q(\cdot|\hat{x}, \pi)$  with includes as a strict subset  $\overline{\mathcal{R}}(q(\cdot|\hat{x}, \pi))$  because  $q(\cdot|\hat{x}, \pi) \in \Delta^c$ . Modify  $\pi$  to  $\pi'$  as follows. For each  $\omega \in \overline{\mathcal{R}}(q(\cdot|\hat{x}, \pi))$ , create a signal  $x_\omega$  with the following properties: (i)  $\pi'(x_\omega|\omega) = \pi(\hat{x}|\omega)$ , (ii)  $\pi'(x_\omega|\omega') = 0$  for all  $\omega' \in \overline{\mathcal{R}}(q(\cdot|\hat{x}, \pi)) \setminus \{\omega\}$ , (iii) for each  $\tilde{\omega} \in \mathcal{R}$ 

$$\pi'(x_{\omega}|\tilde{\omega}) = \pi(\hat{x}|\tilde{\omega}) \frac{\pi(\hat{x}|\omega)}{\sum_{\omega' \in \overline{\mathcal{R}}(g(\cdot|\hat{x},\pi))} \pi(\hat{x}|\omega')}.$$

Note that  $\pi'$  is a well-defined signal device, since  $\pi(\cdot|\omega)$  is a probability distribution over finitely many signals for every  $\omega$ . By construction, for every  $\omega \in \overline{\mathcal{R}}(q(\cdot|\hat{x},\pi))$ ,  $q(\omega'|x_{\omega},\pi')=0$  if  $\omega' \in \overline{\mathcal{R}}(q(\cdot|\hat{x},\pi)) \setminus \{\omega\}$  and for every  $\tilde{\omega} \in \mathcal{R}$ 

$$q(\tilde{\omega}|x_{\omega}, \pi', \mathcal{R}) = \frac{\pi'(x_{\omega}|\tilde{\omega})\sigma(\tilde{\omega}|\mathcal{R})}{\sum_{\omega' \in \mathcal{R}} \pi'(x_{\omega}|\omega')\sigma(\omega'|\mathcal{R})}$$
$$= \frac{\pi(\hat{x}|\tilde{\omega})\sigma(\tilde{\omega}|\mathcal{R})}{\sum_{\omega' \in \mathcal{R}} \pi(\hat{x}|\omega')\sigma(\omega'|\mathcal{R})} = q(\tilde{\omega}|\hat{x}, \pi, \mathcal{R}).$$

This implies that  $\mathcal{A}(p(q(\cdot|x_{\omega},\pi'))) = \mathcal{A}(p(q(\cdot|\hat{x},\pi)))$  for every  $\omega \in \overline{\mathcal{R}}(q(\cdot|\hat{x},\pi))$ . Let the total probability that  $q(\cdot|x_{\omega},\pi')$  arises under  $\pi'$  be  $\beta'(x_{\omega}) = \sum_{\omega' \in \mathcal{S}} \pi'(x_{\omega}|\omega')\sigma(\omega')$  and note that

$$\sum_{\omega \in \overline{\mathcal{R}}(q(\cdot|\hat{x},\pi))} \beta'(x_{\omega}) = \sum_{\omega \in \overline{\mathcal{R}}(q(\cdot|\hat{x},\pi))} \pi(\hat{x}|\omega)\sigma(\omega) + \sum_{\omega \in \overline{\mathcal{R}}(q(\cdot|\hat{x},\pi))} \left[ \sum_{\tilde{\omega} \in \mathcal{R}} \pi'(x_{\omega}|\tilde{\omega})\sigma(\tilde{\omega}) \right]$$

$$= \sum_{\omega \in \overline{\mathcal{R}}(q(\cdot|\hat{x},\pi))} \pi(\hat{x}|\omega)\sigma(\omega) + \sum_{\tilde{\omega} \in \mathcal{R}} \pi(\hat{x}|\tilde{\omega})\sigma(\tilde{\omega}) = \beta(\hat{x}),$$

i.e., the probability that  $q(\cdot|\hat{x},\pi)$  arises under  $\pi$ . Note also that  $q(\cdot|\hat{x},\pi) = \sum_{\omega \in \overline{\mathcal{R}}(q(\cdot|\hat{x},\pi))} q(\cdot|x_{\omega},\pi') \frac{\beta'(x_{\omega})}{\beta(\hat{x})}$ , so  $\hat{q}$  is the conditional expectation of posteriors  $\{q(\cdot|x_{\omega},\pi')\}_{\omega \in \overline{\mathcal{R}}(q(\cdot|\hat{x},\pi))}$ . Indeed, if  $\omega \in \overline{\mathcal{R}}(q(\cdot|\hat{x},\pi))$ , then  $q(\omega|x_{\omega'}) = 0$  for all  $\omega' \neq \omega$  and

$$q(\omega|x_{\omega}, \pi') \frac{\beta'(x_{\omega})}{\beta(\hat{x})} = \frac{\pi(\hat{x}|\omega)\sigma(\omega)}{\beta'(x_{\omega})} \frac{\beta'(x_{\omega})}{\beta(\hat{x})} = q(\omega|\hat{x}, \pi);$$

if  $\omega \in \mathcal{R}$ , then

$$\sum_{\omega' \in \overline{\mathcal{R}}(q(\cdot|\hat{x},\pi))} \frac{\pi'(x_{\omega'}|\omega)\sigma(\omega)}{\beta(\hat{x})} = \sum_{\omega' \in \overline{\mathcal{R}}(q(\cdot|\hat{x},\pi))} \frac{\pi(\hat{x}|\omega)\sigma(\omega) \frac{\pi(\hat{x}|\omega')}{\sum_{\omega'' \in \overline{\mathcal{R}}(q(\cdot|\hat{x},\pi))} \pi(\hat{x}|\omega'')}}{\beta(\hat{x})} = q(\omega|\hat{x},\pi).$$

In summary, the distribution  $\pi'$  replaces the posterior  $\hat{q}$  allocating its probability  $\tau(\hat{q})$  across a collection of posteriors  $q_{\omega} = q(\cdot|x_{\omega}, \pi')$ , each with probability  $\tau'(q_{\omega})$ , such that  $\hat{q} = \sum_{\omega \in \overline{\mathcal{R}}(\hat{q})} q_{\omega} \frac{\tau'(q_{\omega})}{\tau(\hat{q})}$ . Note that  $\tau'(q) = \tau(q)$  for all other  $q \in \text{supp } \tau$ , hence Sender's payoff changes only when posterior  $\hat{q}$  arises. We want to show that this change can only be a (weak) improvement. Given posterior  $\hat{q}$ , let  $a(\hat{q}) \in \mathcal{A}(p(\hat{q}))$  be Receiver's action. Then, Sender's conditional expected payoff from the distribution over  $q_{\omega}$ 's is

$$\sum_{\omega \in \overline{\mathcal{R}}(\hat{q})} \left\{ \max_{a \in \mathcal{A}(p(\hat{q}))} \sum_{\tilde{\omega} \in \mathcal{S}} u_{S}(a, \tilde{\omega}) q_{\omega}(\tilde{\omega}) \right\} \frac{\tau'(q_{\omega})}{\tau(\hat{q})} \geq \sum_{\omega \in \overline{\mathcal{R}}(\hat{q})} \left\{ \sum_{\tilde{\omega} \in \mathcal{S}} u_{S}(a(\hat{q}), \tilde{\omega}) q_{\omega}(\tilde{\omega}) \right\} \frac{\tau'(q_{\omega})}{\tau(\hat{q})} \\
= \sum_{\tilde{\omega} \in \mathcal{S}} u_{S}(a(\hat{q}), \tilde{\omega}) \left\{ \sum_{\omega \in \overline{\mathcal{R}}(\hat{q})} q_{\omega}(\tilde{\omega}) \frac{\tau'(q_{\omega})}{\tau(\hat{q})} \right\} \\
= \sum_{\tilde{\omega} \in \mathcal{S}} u_{S}(a(\hat{q}), \tilde{\omega}) \hat{q}(\tilde{\omega}),$$

namely Sender's expected payoff at posterior  $\hat{q}$ .

This construction can be replicated for all  $\hat{q} \in \operatorname{supp} \tau$  with  $|\overline{\mathcal{R}}(\hat{q})| > 1$ , leading to a new distribution  $\tau_1$  over  $\Delta^c$  such that  $V(\tau_1) \geq V(\tau)$  and, by construction, for every  $q \in \operatorname{supp} \tau_1$  there exists at most one  $\omega \in \overline{\mathcal{R}}$  such that  $q(\omega) \in (0,1)$ . For any such  $q, q(\mathcal{R}) = \sum_{\omega' \in \mathcal{R}} q(\omega') = 1 - q(\omega)$ . Hence,  $q(\omega') = (1 - q(\omega))q(\omega'|\mathcal{R})$  for every  $\omega' \in \mathcal{R}$  and  $q(\omega) = q(\omega)\delta_{\omega}$ , so that  $q = (1 - q(\omega))q(\cdot|\mathcal{R}) + q(\omega)\delta_{\omega}$ .

Part (2): Take any  $\tau \in \mathcal{F}_{\sigma}$  with  $\operatorname{supp} \tau \subset \Delta^{c}$  and satisfying property (1) in the proposition. For every  $\omega \in \overline{\mathcal{R}}$  there must exist at least one  $q \in \operatorname{supp} \tau$  with  $q(\omega) > 0$ : in the device  $\pi$  leading to  $\tau$ ,  $\pi(\cdot|\omega)$  must assign positive probability to some signal  $x \in \operatorname{supp} \pi(\cdot|\omega')$  for some  $\omega' \in \mathcal{R}$ . Let  $Q(\omega) = \{q \in \operatorname{supp} \tau : q(\omega) > 0\}$  and suppose that for some  $\omega^* \in \overline{\mathcal{R}}$  we have  $|Q(\omega^*)| > 1$ . Then let  $T^* = \sum_{q \in Q(\omega^*)} \tau(q)$  and

$$q^* = \sum_{q \in Q(\omega^*)} q\tau(q|Q(\omega^*)) = \sum_{q \in Q(\omega^*)} \left[ (1 - q(\omega^*))q(\cdot|\mathcal{R}) + q(\omega^*)\delta_{\omega^*} \right] \tau(q|Q(\omega^*))$$

$$= \sum_{q \in Q(\omega^*)} q(\cdot|\mathcal{R})(1 - q(\omega^*))\tau(q|Q(\omega^*))$$

$$+\delta_{\omega^*} \sum_{q \in Q(\omega^*)} q(\omega^*)\tau(q|Q(\omega^*)),$$

where the second equality follows from Part (1). So,  $q^*$  arises with probability  $T^*$  and is the convex combination of the posteriors  $\delta_{\omega^*}$  and  $\{q(\cdot|\mathcal{R})\}_{q\in Q(\omega^*)}$ .

Now consider Sender's expected payoff conditional on  $Q(\omega^*)$ , letting a(q) be Receiver's choice at each  $q \in Q(\omega^*)$ :

$$\sum_{q \in Q(\omega^*)} \left\{ \sum_{\omega \in \mathcal{S}} u_S(a(q), \omega) q(\omega) \right\} \tau(q|Q(\omega^*))$$

$$= \sum_{q \in Q(\omega^*)} \left\{ (1 - q(\omega^*)) \sum_{\omega \in \mathcal{S}} u_S(a(q), \omega) q(\omega|\mathcal{R}) + q(\omega^*) u_S(a(q), \omega^*) \right\} \tau(q|Q(\omega^*))$$

$$= \sum_{q \in Q(\omega^*)} \left\{ \sum_{\omega \in \mathcal{S}} u_S(a(q), \omega) q(\omega|\mathcal{R}) \right\} (1 - q(\omega^*)) \tau(q|Q(\omega^*))$$

$$+ \sum_{q \in Q(\omega^*)} u_S(a(q), \omega^*) q(\omega^*) \tau(q|Q(\omega^*))$$

$$= (1 - \xi(\omega^*)) \left\{ \sum_{q \in Q(\omega^*)} \left\{ \sum_{\omega \in \mathcal{S}} u_S(a(q), \omega) q(\omega|\mathcal{R}) \right\} \frac{(1 - q(\omega^*)) \tau(q|Q(\omega^*))}{1 - \xi(\omega^*)} \right\} \right\}$$

$$+ \xi(\omega^*) \left\{ \sum_{q \in Q(\omega^*)} u_S(a(q), \omega^*) \frac{q(\omega^*) \tau(q|Q(\omega^*))}{\xi(\omega^*)} \right\}, \tag{20}$$

where  $\xi(\omega^*) = \sum_{q \in Q(\omega^*)} q(\omega^*) \tau(q|Q(\omega^*))$ . Now recall that  $\mathcal{A}(p(q))$  depends only on  $q(\cdot|\mathcal{R})$ , so any change in q which leaves  $q(\cdot|\mathcal{R})$  unaffected does not change the actions Sender can make Receiver choose. Expression (20) can only increase if, for every  $q \in Q(\omega^*)$ , we replace  $u_S(a(q), \omega^*)$  with  $u_S(a(\tilde{q}), \omega^*) \equiv \max_{q' \in Q(\omega^*)} u_S(a(q'), \omega^*)$ —that is, we shift the entire weight  $\xi(\omega^*)$  to the largest  $u_S(a(q), \omega^*)$ . This change modifies  $\tau$  to  $\tau'$  as follows. For every  $q \notin Q(\omega^*)$ ,  $\tau'(q) = \tau(q)$ . Each  $q \in Q(\omega^*)$  with  $q \neq \tilde{q}$  is replaced by  $q' = q(\cdot|\mathcal{R})$  and  $\tilde{q}$  is replaced by

$$\tilde{q}' = \tilde{q}(\cdot|\mathcal{R}) \frac{(1 - \tilde{q}(\omega^*))\tau(\tilde{q}|Q(\omega^*))}{(1 - \tilde{q}(\omega^*))\tau(\tilde{q}|Q(\omega^*)) + \xi(\omega^*)} + \delta_{\omega^*} \frac{\xi(\omega^*)}{(1 - \tilde{q}(\omega^*))\tau(\tilde{q}|Q(\omega^*)) + \xi(\omega^*)}.$$

Moreover, letting  $Q' = \sup \tau' \setminus \overline{Q(\omega^*)}$  (where  $\overline{Q(\omega^*)}$  is the complement of  $Q(\omega^*)$ ), we have  $\tau'(q'|Q') = (1-q(\omega^*))\tau(q|Q(\omega^*))$  and  $\tau'(\tilde{q}'|Q') = (1-\tilde{q}(\omega^*))\tau(\tilde{q}|Q(\omega^*))+\xi(\omega^*)$ . By construction,  $\sum_{q'\in Q'} q'\tau'(q'|Q') = q^*$  and  $\sum_{q'\in Q'} \tau'(q') = T^*$ , so that  $\tau'\in \mathcal{F}_{\sigma}^c$ . Using this notation, by the previous argument, (20) is less than or equal to

$$\sum_{q \in Q(\omega^*)} \{ \sum_{\omega \in \mathcal{S}} u_S(a(q), \omega) q(\omega | \mathcal{R}) \} (1 - q(\omega^*)) \tau(q | Q(\omega^*)) + \xi(\omega^*) u_S(a(\tilde{q}), \omega^*)$$

$$= \sum_{q' \in Q'} \{ \sum_{\omega \in \mathcal{S}} u_S(a(q'), \omega) q'(\omega) \} \tau(q' | Q') \le \sum_{q' \in Q'} \{ \max_{a \in \mathcal{A}(p(q'))} \sum_{\omega \in \mathcal{S}} u_S(a, \omega) q'(\omega) \} \tau(q' | Q'),$$

where the inequality follows because, for each q' and associated original q in the construction,  $a(q) \in \mathcal{A}(p(q'))$ .

This shows that, for every  $\omega \in \overline{\mathcal{R}}$  such that  $|Q(\omega)| > 1$  under the original  $\tau \in \mathcal{F}_{\sigma}^{c}$ , it is possible to modify  $\tau$  to obtain  $\tau' \in \mathcal{F}_{\sigma}^{c}$  such that  $|Q'(\omega)| = 1$  and  $V(\tau') \geq V(\tau)$ .

To prove the last claim, take any  $\omega \in \overline{\mathbb{R}}$  and associated  $q_{\omega} \in \operatorname{supp} \tau$  such that  $q_{\omega}(\omega) > 0$ . Let  $a(q_{\omega})$  be Receiver's choice at  $q_{\omega}$ . Suppose that there exists  $q' \in \Delta(\mathbb{R})$  such that  $u_S(a(q_{\omega}), \omega) < \max_{a \in \mathcal{A}(p(q'))} u_S(a, \omega) = u_S(a(q'), \omega)$ . By part (1), for any  $\varepsilon > 0$  small enough, we can write

$$q_{\omega} = \lambda_{\varepsilon} q_{\omega}(\cdot | \mathcal{R}) + (1 - \lambda_{\varepsilon}) \left[ \varepsilon q_{\omega}(\cdot | \mathcal{R}) + (1 - \varepsilon) \delta_{\omega} \right],$$

where  $\lambda_{\varepsilon} \in (0,1)$  is chosen so that  $q_{\omega}(\omega) = (1-\lambda_{\varepsilon})(1-\varepsilon)$  and hence  $\lambda_{\varepsilon} \to 1-q_{\omega}(\omega)$  as  $\varepsilon \to 0$ . For any z > 0, define  $q_z = \hat{q} + \frac{1}{z}(\sigma - \hat{q})$  with  $\hat{q} = \sum_{q \neq q_{\omega}} q \frac{\tau(q)}{1-\tau(q_{\omega})}$ . Recall that  $q_{\omega} = q_{\tau(q_{\omega})}$ . Now take posterior q' and consider  $\varepsilon q' + (1-\varepsilon)\delta_{\omega}$ . Note that, for  $\varepsilon > 0$  small enough,  $\varepsilon q' + (1-\varepsilon)\delta_{\omega}$  is arbitrarily close to  $\varepsilon q_{\omega}(\cdot|\mathcal{R}) + (1-\varepsilon)\delta_{\omega}$  and hence

$$q'_{\omega} = \lambda_{\varepsilon} q_{\omega}(\cdot | \mathcal{R}) + (1 - \lambda_{\varepsilon})[\varepsilon q' + (1 - \varepsilon)\delta_{\omega}]$$

is arbitrarily close to  $q_{\omega}$ .

Given  $q_{\omega}$ , there exists  $z_{\varepsilon} \geq \tau(q_{\omega})$  and  $\alpha_{\varepsilon} \in (0,1)$  such that  $q_{z_{\varepsilon}} \in int\Delta^{c}$ ,  $q_{\alpha_{\varepsilon}} \in \Delta^{c}$ , and  $q_{z_{\varepsilon}} = \alpha_{\varepsilon}q_{\alpha_{\varepsilon}} + (1 - \alpha_{\varepsilon})q'_{\omega}$ . Moreover, as  $\varepsilon \to 0$ , we can choose  $z_{\varepsilon}$  and  $\alpha_{\varepsilon}$  so that  $z_{\varepsilon} \downarrow \tau(q_{\omega})$  and  $\alpha_{\varepsilon} \downarrow 0$ . Hence, we can modify  $\tau$  to obtain  $\tau_{\varepsilon}$  with

$$\tau_{\varepsilon}(q) = \begin{cases} \frac{1 - z_{\varepsilon}}{1 - \tau(q_{\omega})} \tau(q) & \text{if } q \in \mathbf{supp } \tau, q \neq q_{\omega} \\ z_{\varepsilon} \alpha_{\varepsilon} & \text{if } q = q_{\alpha_{\varepsilon}} \\ z_{\varepsilon} (1 - \alpha_{\varepsilon}) \lambda_{\varepsilon} & \text{if } q = q_{\omega}(\cdot | \mathcal{R}) \\ z_{\varepsilon} (1 - \alpha_{\varepsilon}) (1 - \lambda_{\varepsilon}) & \text{if } q = \varepsilon q' + (1 - \varepsilon) \delta_{\omega} \end{cases}.$$

It can be easily checked that  $\tau_{\varepsilon} \in \mathcal{F}_{\sigma}$  for every  $\varepsilon > 0$ .

Now consider  $V(\tau)$  and  $V(\tau_{\varepsilon})$ . Letting  $k = \sum_{q \neq q_{\omega}} w(q) \frac{\tau(q)}{1 - \tau(q_{\omega})}$ , we have

$$V(\tau) = \tau(q_{\omega})\{(1 - q_{\omega}(\omega)) \sum_{\omega' \in \mathcal{S}} u_S(a(q_{\omega}), \omega') q_{\omega}(\omega' | \mathcal{R}) + q_{\omega}(\omega) u_S(a(q_{\omega}), \omega)\} + (1 - \tau(q_{\omega}))k,$$

$$V(\tau_{\varepsilon}) = (1 - z_{\varepsilon})k + z_{\varepsilon}\alpha_{\varepsilon} \sum_{\omega' \in \mathcal{S}} u_{S}(a(q_{\alpha_{\varepsilon}}), \omega')q_{\alpha_{\varepsilon}}(\omega')$$

$$+ z_{\varepsilon}(1 - \alpha_{\varepsilon})\lambda_{\varepsilon} \sum_{\omega' \in \mathcal{S}} u_{S}(a(q_{\omega}), \omega')q_{\omega}(\omega'|\mathcal{R})$$

$$+ z_{\varepsilon}(1 - \alpha_{\varepsilon})(1 - \lambda_{\varepsilon})\{\varepsilon \sum_{\omega' \in \mathcal{S}} u_{S}(a_{\omega}, \omega')q'(\omega') + (1 - \varepsilon)u_{S}(a_{\omega}, \omega)\},$$

where  $a_{\omega} = \arg \max_{a \in \mathcal{A}(p(q'))} u_S(a, \omega)$ . Recall that Sender's expected payoff from any action and posterior q is finite. So,

$$\lim_{\varepsilon \to 0} V(\tau_{\varepsilon}) = \tau(q_{\omega})(1 - q_{\omega}(\omega)) \sum_{\omega' \in \mathcal{S}} u_{S}(a(q_{\omega}), \omega') q_{\omega}(\omega' | \mathcal{R})$$
$$+ \tau(q_{\omega}) q_{\omega}(\omega) u_{S}(a_{\omega}, \omega) + (1 - \tau(q_{\omega}))k > V(\tau).$$

Therefore, there exists  $\varepsilon > 0$  small enough such that  $V(\tau_{\varepsilon}) > V(\tau)$ . This shows that we can focus without loss on distributions  $\tau \in \mathcal{F}_{\sigma}^{c}$  such that  $u_{S}(a(q_{\omega}), \omega) \geq \max_{q \in \Delta(\mathcal{R})} \{\max_{a \in \mathcal{A}(p(q))} u_{S}(a, \omega)\}$  for every  $\omega \in \overline{\mathcal{R}}$ . Finally, recall that by varying  $q \in \Delta(\mathcal{R})$  Sender can achieve all Receiver's posteriors  $p \in \Delta(\mathcal{R})$  (see discussion after Proposition 3), hence we can express the right-hand side of the last inequality directly in terms of p's.

#### B.7 Proof of Corollary 3

Hereafter, let  $\hat{w}_*^c(\sigma)$  be the expression of  $\hat{w}^c(\sigma)$  in the statement and  $\mathcal{F}_{\sigma}^{c*} \subset \mathcal{F}_{\sigma}^c$  be the family of all feasible distributions satisfying the properties in Proposition 4.

Claim 1. 
$$V(\tau) \leq \hat{w}_*^c(\sigma)$$
 for any  $\tau \in \mathcal{F}_{\sigma}^{c*}$ .

*Proof.* Given  $\tau$ , for each  $\omega \in \overline{\mathcal{R}}$ , let  $q_{\omega}$  be the unique posterior in  $\operatorname{supp} \tau$  assigning positive probability to  $\omega$  as in part (2) of Proposition 4. Also, let  $\overline{Q} = \{q_{\omega}\}_{{\omega} \in \overline{\mathcal{R}}}$ ,  $Q = \operatorname{supp} \tau \setminus \overline{Q}$ , and  $a(q_{\omega}) = a_{\omega} \in \arg \max_{a \in \mathcal{A}(p(q_{\omega}))} u_S(a, \omega)$  for every  $\omega \in \overline{\mathcal{R}}$ . Then,

$$V(\tau) = \sum_{q \in Q} w(q)\tau(q)$$

$$+ \sum_{q_{\omega} \in \overline{Q}} \left[ \sum_{\tilde{\omega} \in \mathcal{S}} u_{S}(a_{\omega}, \tilde{\omega}) \{ (1 - q_{\omega}(\omega))q_{\omega}(\tilde{\omega}|\mathcal{R}) + q_{\omega}(\omega)u_{S}(a_{\omega}, \omega) \} \right] \tau(q_{\omega})$$

$$= \sum_{q \in Q} w(q)\tau(q) + \sum_{q_{\omega} \in \overline{Q}} \left[ \sum_{\tilde{\omega} \in \mathcal{S}} u_{S}(a_{\omega}, \tilde{\omega}) q_{\omega}(\tilde{\omega}|\mathcal{R}) \right] (1 - q_{\omega}(\omega))\tau(q_{\omega})$$

$$+ \sum_{q_{\omega} \in \overline{Q}} u_{S}(a_{\omega}, \omega) q_{\omega}(\omega)\tau(q_{\omega})$$

$$\leq \sum_{q \in Q} w(q)\tau(q) + \sum_{q_{\omega} \in \overline{Q}} w(q_{\omega}(\cdot|\mathcal{R}))(1 - q_{\omega}(\omega))\tau(q_{\omega})$$

$$+ \sum_{q_{\omega} \in \overline{Q}} u_{S}^{*}(\omega) q_{\omega}(\omega)\tau(q_{\omega}). \tag{21}$$

Since

$$\sigma = \sum_{q \in Q} q\tau(q) + \sum_{q_{\omega} \in \overline{Q}} q_{\omega}(\cdot|\mathcal{R})(1 - q_{\omega}(\omega))\tau(q_{\omega}) + \sum_{q_{\omega} \in \overline{Q}} \delta_{\omega}q_{\omega}(\omega)\tau(q_{\omega}),$$

we have  $\sigma(\omega) = q_{\omega}(\omega)\tau(q_{\omega})$  for every  $\omega \in \overline{\mathcal{R}}$  and  $\sum_{q_{\omega} \in \overline{Q}} q_{\omega}(\omega)\tau(q_{\omega}) = \sigma(\overline{\mathcal{R}})$ . This in turn implies that for each  $\tilde{\omega} \in \mathcal{R}$ ,

$$\frac{1}{\sigma(\mathcal{R})} \left[ \sum_{q \in Q} q(\tilde{\omega}) \tau(q) + \sum_{q_{\omega} \in \overline{Q}} q_{\omega}(\tilde{\omega}|\mathcal{R}) (1 - q_{\omega}(\omega)) \tau(q_{\omega}) \right] = \frac{\sigma(\tilde{\omega})}{\sigma(\mathcal{R})} = \sigma(\tilde{\omega}|\mathcal{R}).$$

The first two terms in (21) are then a convex combination of values  $w^c(q)$  with  $q \in \Delta(\mathcal{R})$  and with average posterior  $\sigma(\cdot|\mathcal{R})$ . Therefore, expression (21) is bounded above by

$$\sigma(\mathcal{R})\hat{w}^c(\sigma(\cdot|\mathcal{R})) + \sum_{\omega \in \overline{\mathcal{R}}} u_S^*(\omega)\sigma(\omega) = \hat{w}_*^c(\sigma).$$

Claim 2. For any  $\varepsilon > 0$  there exists  $\tau_{\varepsilon} \in \mathcal{F}_{\sigma}^{c*}$  such that  $V(\tau_{\varepsilon}) \geq \hat{w}_{*}^{c}(\sigma) - \varepsilon$ . Hence,  $\hat{w}_{*}^{c}(\sigma)$  is the least upper bound of the values of  $V(\tau)$  over  $\mathcal{F}_{\sigma}^{c*}$ .

*Proof.* Starting from any  $\tau \in \mathcal{F}_{\sigma}^{c*}$ , first construct a sequence  $\{\tau_n\}_{n=1}^{\infty} \subset \mathcal{F}_{\sigma}^{c*}$  with  $\tau_0 = \tau$  as follows. Define Q and  $\overline{Q}$  as in the proof of Claim 1. For every  $q \in Q$ , let  $\tau_n(q) = \tau_0(q)$ . For each  $q_{\omega} \in \overline{Q}$  and each  $n \geq 1$ , split  $\tau_0(q_{\omega})$  by replacing it with

$$\tau_n(q') = \begin{cases} \tau_0(q_{\omega}) z_{\omega,n} & \text{for } q' = q_{\omega}(\cdot | \mathcal{R}) \\ \tau_0(q_{\omega}) (1 - z_{\omega,n}) & \text{for } q' = q_{\omega,n} \equiv \frac{1}{K_{\omega}^n} q_{\omega}(\cdot | \mathcal{R}) + (1 - \frac{1}{K_{\omega}^n}) \delta_{\omega} \end{cases},$$

with  $z_{\omega,n} \in (0,1)$  so that  $q_{\omega}(\omega) = (1-z_{\omega,n})(1-\frac{1}{K_{\omega}^n})$  for every n, where  $K_{\omega} > 1$  is chosen large enough so as to satisfy this condition for n=1 and hence for all  $n \geq 1$ . By construction,  $z_{\omega,n} \uparrow (1-q_{\omega}(\omega))$  and  $q_{\omega,n} \to \delta_{\omega}$  as  $n \to \infty$ , and for every n

$$q_{\omega} = z_{\omega,n} q_{\omega}(\cdot | \mathcal{R}) + (1 - z_{\omega,n}) q_{\omega,n}.$$

So, for every n, the conditional distribution defined by  $z_{\omega,n+1}$  is a mean-preserving spread around  $q_{\omega}$  of that defined by  $z_{\omega,n}$ . It follows that, for every  $q_{\omega} \in \overline{Q}$ ,

$$\max_{a \in \mathcal{A}(p(q_{\omega}))} \sum_{\tilde{\omega} \in \mathcal{S}} u_{S}(a, \tilde{\omega}) q_{\omega}(\tilde{\omega}) \leq z_{\omega, 1} \max_{a \in \mathcal{A}(p(q_{\omega}))} \sum_{\tilde{\omega} \in \mathcal{S}} u_{S}(a, \tilde{\omega}) q_{\omega}(\tilde{\omega} | \mathcal{R}) + (1 - z_{\omega, 1}) \max_{a \in \mathcal{A}(p(q_{\omega}))} \sum_{\tilde{\omega} \in \mathcal{S}} u_{S}(a, \tilde{\omega}) q_{\omega, 1}(\tilde{\omega}),$$

and for all  $n \ge 1$ 

$$\begin{split} z_{\omega,n} \max_{a \in \mathcal{A}(p(q_{\omega}))} \sum_{\tilde{\omega} \in \mathcal{S}} u_{S}(a,\tilde{\omega}) q_{\omega,n}(\tilde{\omega}|\mathcal{R}) + (1-z_{\omega,n}) \max_{a \in \mathcal{A}(p(q_{\omega}))} \sum_{\tilde{\omega} \in \mathcal{S}} u_{S}(a,\tilde{\omega}) q_{\omega,n}(\tilde{\omega}) \\ \leq z_{\omega,n+1} \max_{a \in \mathcal{A}(p(q_{\omega}))} \sum_{\tilde{\omega} \in \mathcal{S}} u_{S}(a,\tilde{\omega}) q_{\omega}(\tilde{\omega}|\mathcal{R}) + (1-z_{\omega,n+1}) \max_{a \in \mathcal{A}(p(q_{\omega}))} \sum_{\tilde{\omega} \in \mathcal{S}} u_{S}(a,\tilde{\omega}) q_{\omega,n+1}(\tilde{\omega}). \end{split}$$

For every n, letting  $Q'_n = \sup \tau_n \setminus Q$ , by construction we also have  $\sum_{q_\omega \in \overline{Q}} q_\omega \tau_0(q_\omega) = \sum_{q' \in Q'_n} q' \tau_n(q')$  and  $\sum_{q_\omega \in \overline{Q}} \tau_0(q_\omega) = \sum_{q' \in Q'_n} \tau_n(q')$ —hence each  $\tau_n \in \mathcal{F}^{c*}_{\sigma}$ . We conclude that  $V(\tau_n) \leq V(\tau_{n+1})$  for every n.

Now, for each n, let  $Z_n = \sum_{q \in Q} \tau_n(q) + \sum_{\omega \in \overline{\mathcal{R}}} \tau_0(q_\omega) z_{\omega,n}$  and express  $V(\tau_n)$  as

$$Z_n \underbrace{\sum_{q \in Q} w(q) \frac{\tau_0(q)}{Z_n} + \sum_{\omega \in \overline{\mathcal{R}}} w(q_{\omega}(\cdot|\mathcal{R})) \frac{\tau_0(q_{\omega}) z_{\omega,n}}{Z_n}}_{B_n} + \underbrace{\sum_{\omega \in \overline{\mathcal{R}}} w(q_{\omega,n}) \tau_0(q_{\omega}) (1 - z_{\omega,n})}_{B'_n}.$$

Since  $z_{\omega,n} \uparrow (1 - q_{\omega}(\omega))$  for every  $\omega \in \overline{\mathcal{R}}$  as  $n \to \infty$  and we must have

$$\sigma(\omega) = \tau_0(q_\omega)(1 - z_{\omega,n})(1 - \frac{1}{K_\omega^n}) = \tau(q_\omega)q_\omega(\omega)$$

for all n, it follows that  $\lim_{n\to\infty} B'_n = \sum_{\omega\in\overline{\mathcal{R}}} u_S^*(\omega)\sigma(\omega)$ . Note also that  $\lim_{n\to\infty} Z_n = 1 - \lim_{n\to\infty} \sum_{\omega\in\overline{\mathcal{R}}} \tau_0(q_\omega)(1-z_{\omega,n}) = \sigma(\mathcal{R})$ .

Regarding the term  $B_n$ , first observe that

$$\frac{1}{Z_n} \left[ \sum_{q \in Q} q \tau_n(q) + \sum_{\omega \in \overline{\mathcal{R}}} q_{\omega}(\cdot | \mathcal{R}) \tau_0(q_{\omega}) z_{\omega,n} \right] = \hat{q}_n \in \Delta(\mathcal{R}),$$

and hence  $\lim_{n\to\infty} \hat{q}_n = \sigma(\cdot|\mathcal{R})$ . This implies that, for every n,  $B_n$  is a convex combination of values  $w^c(q)$  with  $q \in \Delta(\mathcal{R})$  and with average posterior  $\hat{q}_n$ . Recall that, restricted to  $\Delta(\mathcal{R})$ , the function  $\hat{w}^c(q)$  is continuous in q. Letting  $\chi = \hat{w}^c(\sigma(\cdot|\mathcal{R}))$  and  $\zeta = \sum_{\omega \in \overline{\mathcal{R}}} u_S^*(\omega)\sigma(\omega)$  for simplicity, we can write

$$\hat{w}_*^c(\sigma) - V(\tau_n) \le |\chi| |Z_n - \sigma(\mathcal{R})| + |Z_n| |B_n - \chi| + |B'_n - \zeta|.$$

Given any  $\varepsilon > 0$ , there exists  $N_1$  such that  $|\chi| |Z_n - \sigma(\mathcal{R})| + |B'_n - \zeta| \leq \frac{\varepsilon}{2}$  for all  $n \geq N_1$ . Also, there exists  $N_2$  such that  $|\chi - \hat{w}^c(\hat{q}_n)| \leq \frac{\varepsilon}{2}$  for all  $n \geq N_2$ . So, fix  $n^* \geq \max\{N_1, N_2\}$  and consider the distribution  $\hat{\tau} \in \Delta(\Delta(\mathcal{R}))$  that achieves  $\hat{w}^c(\hat{q}_{n^*})$ . Define the distribution  $\tau_{\varepsilon}$  as follows:

$$\tau_{\varepsilon}(q) = \begin{cases} Z_{n^*} \hat{\tau}(q) & \text{if } q \in \mathbf{supp} \, \hat{\tau} \\ \tau_{n^*}(q_{\omega,n^*}) & \text{if } q = q_{\omega,n^*} \\ 0 & \text{otherwise.} \end{cases}$$

By construction,  $V(\tau_{\varepsilon}) \geq \hat{w}_{*}^{c}(\sigma) - \varepsilon$  and

$$\sum_{q \in \mathbf{supp}} q \tau_{\varepsilon}(q) = Z_{n^*} \sum_{q \in \mathbf{supp} \, \hat{\tau}} q \hat{\tau}(q) + \sum_{\omega \in \overline{\mathcal{R}}} q_{\omega,n^*} \tau_{n^*}(q_{\omega,n^*})$$

$$= \sum_{q \in Q} q \tau_0(q) + \sum_{\omega \in \overline{\mathcal{R}}} q_{\omega}(\cdot | \mathcal{R}) \tau_0(q_{\omega}) z_{\omega,n^*}$$

$$+ \sum_{\omega \in \overline{\mathcal{R}}} q_{\omega,n^*} \tau_0(q_{\omega}) (1 - z_{\omega,n^*}) = \sigma.$$

# B.8 Proof of Proposition 5

The result follows from the next two claims.

Claim 3. If  $w^d(q) \leq h(q)$  for all  $q \in \Delta^d$ , then  $\hat{w}(\sigma) \leq \hat{w}^c(\sigma)$ .

*Proof.* For any  $\tau \in \mathcal{F}_{\sigma}$ , let  $q^c = \sum_{q \in C_{\tau}} q\tau(q|C_{\tau})$  and  $q^d = \sum_{q \in D_{\tau}} q\tau(q|D_{\tau})$  so that  $q^c \in \Delta^c$ ,  $q^d \in \Delta^d$ ,  $\sigma = \tau^c q^c + (1 - \tau^c)q^d$ . Then,

$$V(\tau) \leq \tau^{c} \hat{w}^{c}(q^{c}) + (1 - \tau^{c}) \sum_{q \in D_{\tau}} w^{d}(q) \tau(q|D_{\tau})$$

$$\leq \tau^{c} \hat{w}^{c}(q^{c}) + (1 - \tau^{c}) \sum_{q \in D_{\tau}} h(q) \tau(q|D_{\tau})$$

$$= \tau^{c} \hat{w}^{c}(q^{c}) + (1 - \tau^{c}) \sum_{q \in D_{\tau}} \left[ \sum_{\omega \in \overline{\mathcal{R}}} u_{S}^{*}(\omega) q(\omega) \right] \tau(q|D_{\tau})$$

$$= \tau^{c} \hat{w}^{c}(q^{c}) + \sum_{\omega \in \overline{\mathcal{R}}} u_{S}^{*}(\omega) \beta(\omega), \tag{22}$$

 $\beta(\omega) = \tau^d \sum_{q \in D_\tau} q(\omega) \tau(q|D_\tau)$ . Note that  $\operatorname{supp} q^c \supset \mathcal{R}$  and  $q^c \in \operatorname{int}\Delta(\operatorname{supp} q^c)$ . Therefore, we can view  $q^c$  as Sender's prior in the fictitious environment with  $\tilde{\Omega} = \operatorname{supp} q^c$  and  $\tilde{\rho} = \rho_0$ . By

Proposition 4, we then have

$$\hat{w}^c(q^c) = q^c(\mathcal{R})\hat{w}^c(q^c(\cdot|\mathcal{R})) + \sum_{\omega \in \overline{\mathcal{R}}} u_S^*(\omega)q^c(\omega).$$

Moreover, we have

$$\sigma = \tau^{c} q^{c}(\mathcal{R}) q^{c}(\cdot | \mathcal{R}) + \sum_{\omega \in \overline{\mathcal{R}}} \delta_{\omega} \{ \beta(\omega) + \tau^{c} q^{c}(\omega) \},$$

which implies that  $\sigma(\omega) = \beta(\omega) + \tau^c q^c(\omega)$  for all  $\omega \in \overline{\mathcal{R}}$  and hence  $q^c(\cdot|\mathcal{R}) = \sigma(\cdot|\mathcal{R})$ . Therefore, (22) is equal to

$$\sigma(\mathcal{R})\hat{w}(\sigma(\cdot|\mathcal{R})) + \sum_{\omega \in \overline{\mathcal{R}}} u_S^*(\omega)\sigma(\omega) = \hat{w}^c(\sigma).$$

Using the definition of  $W_{\sigma}$  and Lemma 2, we conclude that  $\hat{w}(\sigma) \leq \hat{w}^{c}(\sigma)$ .

Claim 4. If  $w^d(q) > h(q)$  for some  $q \in \Delta^d$ , then there exists  $\tau$  such that  $V(\tau) > \hat{w}^c(\sigma)$  and hence  $\hat{w}(\sigma) > \hat{w}^c(\sigma)$ .

*Proof.* Let  $q^*$  be any element of  $\Delta(\overline{\mathcal{R}})$  with  $w^d(q^*) > h(q^*)$ . Since  $\sigma \in int\Delta(\mathcal{S})$ , there exists  $\lambda \in (0,1)$  and  $q^c \in int\Delta(\mathcal{S})$  such that  $\sigma = \lambda q^c + (1-\lambda)q^*$ . By the same argument in the proof of Claim 3

$$\begin{split} (1-\lambda)w^d(q^*) + \lambda \hat{w}^c(q^c) &> (1-\lambda)\sum_{\omega \in \overline{\mathcal{R}}} u_S^*(\omega)q^*(\omega) + \lambda \hat{w}^c(q^c) \\ &= \sum_{\omega \in \overline{\mathcal{R}}} u_S^*(\omega)\{(1-\lambda)q^*(\omega) + \lambda q^c(\omega)\} + \lambda q^c(\mathcal{R})\hat{w}^c(q^c(\cdot|\mathcal{R})) \\ &= \hat{w}^c(\sigma), \end{split}$$

where the last equality follows again from observing that  $\sigma(\omega) = (1 - \lambda)q^*(\omega) + \lambda q^c(\omega)$  for all  $\omega \in \overline{\mathcal{R}}$  and hence  $q^c(\cdot|\mathcal{R}) = \sigma(\cdot|\mathcal{R})$ . Therefore, there exists  $\tau \in \mathcal{F}_{\sigma}$  such that  $V(\tau) > \hat{w}^c(\sigma)$ .

#### B.9 Lemma 4

**Lemma 4.** The function  $\hat{w}_*^d$  satisfies the following properties:

- (i) for every  $q \in \Delta^d$ , there exists  $\tau \in \Delta(\Delta^d)$  such that  $\hat{w}^d_* = \sum_{q'} w^d_*(q')\tau(q')$  with  $q = \sum_{q'} q'\tau(q')$  and  $|\mathbf{supp}\,\tau| \leq |\overline{\mathcal{R}}|$ ;
- (ii)  $\hat{w}^d \leq \hat{w}_*^d$  with equality over  $int\Delta^d$ ;
- (iii)  $\hat{w}_*^d = \operatorname{cl}\hat{w}_*^d$  and hence it is continuous.

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Proof. Part (i): By Corollary 17.1.5 in Rockafellar (1997),

$$\hat{w}_*^d(q) = \sup_{T(q)} \sum_{m=1}^{|\overline{\mathcal{R}}|} w_*^d(q_m) \tau_m$$

where

$$T(q) = \left\{ \{(q_m, \tau_m)\}_{m=1}^{|\overline{\mathcal{R}}|} : \sum_{m=1}^{|\overline{\mathcal{R}}|} q_m \tau_m = q, \sum_{m=1}^{|\overline{\mathcal{R}}|} \tau_m = 1, \tau_m \ge 0, q_m \in \Delta(\overline{\mathcal{R}}), \forall m \right\}.$$

Since  $w_*^d$  is upper semicontinuous and T(q) is compact, by standard arguments  $\hat{w}_*^d(q)$  is achieved for every  $q \in \Delta^d$ .

Part (ii): Given a function  $f: \Delta^d \to \mathbb{R}$ , let hyp f be the hypograph of f: hyp  $f = \{(q, \xi) : q \in \Delta^d, \xi \in \mathbb{R}, \xi \leq f(q)\}$ . Note that hyp  $w_*^d = \overline{\text{hyp } w^d}$ . Therefore, for all  $q \in \Delta^d$ ,

$$\begin{split} \hat{w}^d(q) &= \sup\{\xi : (q, \xi) \in \operatorname{co}(\operatorname{hyp} w^d)\} \\ &\leq \sup\{\xi : (q, \xi) \in \operatorname{co}(\overline{\operatorname{hyp} w^d})\} = \hat{w}_*^d(q). \end{split}$$

Now consider the closure of  $\hat{w}^d$ ,  $\text{cl}\hat{w}^d$ , which is the unique continuous extension of  $\hat{w}^d$  to  $\Delta^d$  by Theorem 10.3 in Rockafellar (1997), is concave, and satisfies  $\text{cl}\hat{w}^d \geq \hat{w}^d \geq w^d$ . So, for every  $q \in \Delta^d$ ,

$$w_*^d(q) = \lim \sup_{q' \to q} w^d(q') \le \lim \sup_{q' \to q} \operatorname{cl} \hat{w}^d(q') = \operatorname{cl} \hat{w}^d(q).$$

Hence,  $\operatorname{cl} \hat{w}^d$  is a concave function majorizing  $w^d_*$ . Since  $\hat{w}^d_*$  is the smallest of such functions,  $\operatorname{cl} \hat{w}^d \geq \hat{w}^d_*$ . Finally, since  $\operatorname{cl} \hat{w}^d = \hat{w}^d$  over  $\operatorname{int} \Delta^d$ , property (i) follows.

Part (iii): We already know that  $\hat{w}_*^d = \operatorname{cl} \hat{w}_*^d$  over  $\operatorname{int} \Delta^d$ . By definition,  $\operatorname{hyp} \operatorname{cl} \hat{w}_*^d = \overline{\operatorname{hyp} \hat{w}_*^d}$ . If  $\operatorname{hyp} \hat{w}_*^d$  is closed, then  $\operatorname{hyp} \hat{w}_*^d = \operatorname{hyp} \operatorname{cl} \hat{w}_*^d$  and hence we are done. Indeed, by definition  $\hat{w}_*^d \leq \operatorname{cl} \hat{w}_*^d$ . So, suppose there exists  $q \in \partial \Delta^d$  such that  $\hat{w}_*^d(q) < \operatorname{cl} \hat{w}_*^d(q)$ . Then there exists  $\xi \in \mathbb{R}$  such that  $\hat{w}_*^d(q) < \xi \leq \operatorname{cl} \hat{w}_*^d$ , which is a contradiction. So, we need to prove that  $\operatorname{hyp} \hat{w}_*^d$  is closed.

First, for every  $q \in \Delta^d$ , by property  $\hat{w}_*^d(q) = \max\{\xi : (q, \xi) \in \operatorname{co}(\operatorname{hyp} w_*^d)\}$  and therefore  $\operatorname{hyp} \hat{w}_*^d = \operatorname{co}(\operatorname{hyp} w_*^d)$ . Second, define  $\underline{w}_*^d = \inf_{q \in \Delta^d} w_*^d(q)$  so that we can express  $\operatorname{hyp} w_*^d$  as  $G \cup H$  where

$$G = \{(q,\xi): q \in \Delta^d, \underline{w}_*^d - 1 \leq \xi \leq w_*^d(q)\} \quad \text{and} \quad H = \{(q,\xi): q \in \Delta^d, \xi \leq \underline{w}_*^d - 1\}.$$

Now co(hyp  $w_*^d$ ) = (coG)  $\cup$  (coH) = (coG)  $\cup$  H. One inclusion is trivial. Now consider  $(q, \xi) \in$  co(hyp  $w_*^d$ ). Then, by Theorem 2.3 in Rockafellar (1997),  $(q, \xi)$  is a convex combination of points  $(q_n, \xi_n)$  in hyp  $w_*^d$ . Therefore,  $q \in \Delta^d$  as the latter is a convex set and  $\xi = \sum_n \alpha_n \xi_n \le$ 

 $\sum_{n} \alpha_{n} w_{*}^{d}(q_{n})$  as  $\alpha_{n} \geq 0$  for all n. But  $(coG) \cup H$  contains all convex combinations of points in hyp  $w_{*}^{d}$  that satisfy this property. Finally, note that H is closed and G is bounded and closed since  $w_{*}^{d}$  is upper semicontinuous. Therefore, co(G) is also closed by Theorem 17.2 in Rockafellar (1997). We conclude that  $co(hyp w_{*}^{d}) = (coG) \cup H$  is closed, as desired.

#### B.10 Proof of Proposition 6

Claim 5. If  $\hat{w}(\sigma) > \hat{w}^c(\sigma)$ , then

$$\hat{w}(\sigma) = \tau^c \hat{w}^c(q^c) + (1 - \tau^c) \hat{w}_*^d(q^d)$$

with  $q^c \in \Delta^c$ ,  $q^d \in \Delta^d$ ,  $\sigma(\mathcal{R}) \leq \tau^c < 1$ , and  $\sigma = \tau^c q^c + (1 - \tau^c) q^d$ .

*Proof.* Take any  $\tau \in \mathcal{F}_{\sigma}$  with  $\operatorname{supp} \tau \leq |\mathcal{S}|$  and  $\tau^d = \tau(D_{\tau}) > 0$ , which is necessary by Lemma 3. Define<sup>50</sup>  $\tau(q|C_{\tau}) = \frac{\tau(q)}{\tau(C_{\tau})}$  and

$$\tau(q|D_{\tau}) = \begin{cases} 0 & \text{if } \tau(D_{\tau}) = 0 \text{ or } q \notin D_{\tau} \\ \frac{\tau(q)}{\tau(D_{\tau})} & \text{if } \tau(D_{\tau}) > 0 \text{ and } q \in D_{\tau} \end{cases}.$$

We can then write

$$V(\tau) = \tau^c \sum_{q \in C_{\tau}} w(q)\tau(q|C_{\tau}) + \tau^d \sum_{q \in D_{\tau}} w(q)\tau(q|D_{\tau}).$$

If we define  $q^c = \sum_{q \in C_\tau} q\tau(q|C_\tau)$  and  $q^d = \sum_{q \in D_\tau} q\tau(q|D_\tau)$ , we have  $q^c \in \Delta^c$ ,  $q^d \in \Delta^d$ ,  $\sigma = \tau^c q^c + (1 - \tau^c)q^d$ . Since  $V(\tau) \leq \tau^c \hat{w}^c(q^c) + (1 - \tau^c)\hat{w}^d(q^d)$ , we must have

$$\hat{w}(\sigma) = W_{\sigma} = \sup_{\tau \in \mathcal{F}_{\sigma}} V(\tau) \le \sup_{\mathcal{T}} \{ \tau^{c} \hat{w}^{c}(q^{c}) + (1 - \tau^{c}) \hat{w}^{d}(q^{d}) \}, \tag{23}$$

where

$$\mathcal{T} = \{ (\tau^c, q^c, q^d) : \tau^c \in [\sigma(\mathcal{R}), 1], q^c \in \Delta^c, q^d \in \Delta^d, \sigma = \tau^c q^c + (1 - \tau^c) q^d \};$$

moreover, the inequality in (23) must be an equality, otherwise there would be  $\tau \in \mathcal{F}_{\sigma}$  with  $V(\tau) > W_{\sigma}$ . Since  $\tau^c \in [\sigma(\mathcal{R}), 1]$ , we must have  $q^c(\omega) = \frac{1}{\tau^c} \sigma(\omega) \geq \sigma(\omega)$  for all  $\omega \in \mathcal{R}$ . The function  $\hat{w}^c$  is continuous over  $\Delta(\mathcal{R})$  and therefore, by Corollary 3,  $\hat{w}^c(q^c)$  is continuous over  $Q^c = \{q \in \Delta^c : q(\omega) \geq \sigma(\omega), \forall \omega \in \mathcal{R}\}$ , which is a compact subset of  $\Delta^c$ . Construct  $\mathcal{T}'$  by replacing  $\Delta^c$  with  $Q^c$  in  $\mathcal{T}$ . So, since  $\hat{w}^d \leq \text{cl}\hat{w}^d = \text{cl}\hat{w}^d_* = \hat{w}^d_*$  and  $\hat{w}^d_*$  is continuous by Lemma

<sup>&</sup>lt;sup>50</sup>Recall that  $\tau(C_{\tau}) > 0$  always by Corollary 1

4, the right-hand side of (23) equals

$$\sup_{\mathcal{T}'} \{ \tau^c \hat{w}^c(q^c) + (1 - \tau^c) \hat{w}^d(q^d) \} \le \max_{\mathcal{T}'} \{ \tau^c \hat{w}^c(q^c) + (1 - \tau^c) \hat{w}_*^d(q^d) \}.$$

First, note that the maximum on the right-hand side must be attained at  $\tau^c < 1$ , because by assumption  $\hat{w}^c(\sigma) < \hat{w}(\sigma)$ . Second, note that the inequality must be an equality. This is immediate if  $|\overline{\mathcal{R}}| = 1$ , since in this case  $\hat{w}^d = w^d = w^d_* = \hat{w}^d_*$ . So, suppose  $|\overline{\mathcal{R}}| > 1$  and define  $\Delta_n^d = \{q \in \Delta(\overline{\mathcal{R}}) : q(\omega) \ge |\overline{\mathcal{R}}|^{-n}, \forall \omega \in \overline{\mathcal{R}}\}$  for  $n \ge 1$ . Construct  $\mathcal{T}'_n$  by replacing  $\Delta^d$  with  $\Delta_n^d$  in  $\mathcal{T}'$  for every n. Note that, for all n,  $\Delta_n^d \subset int\Delta^d$ ,  $\Delta_n^d \subset \Delta_{n+1}^d$ , and  $\Delta_n^d \to \Delta^d$  as  $n \to \infty$ . Since  $\hat{w}^d = \hat{w}^d_*$  over  $int\Delta^d$  by Lemma 4, for all n we have

$$\max_{\mathcal{T}_n'} \{ \tau^c \hat{w}^c(q^c) + (1 - \tau^c) \hat{w}_*^d(q^d) \} \le \sup_{\mathcal{T}'} \{ \tau^c \hat{w}^c(q^c) + (1 - \tau^c) \hat{w}^d(q^d) \}.$$

Since the left-hand side forms an increasing sequence in n that converges to the maximum over  $\mathcal{T}'$ , the desired equality follows.

Claim 6. If  $\hat{w}(\sigma) > \hat{w}^c(\sigma)$ , then (15) holds. Moreover, if  $\hat{\tau} \in \Delta(\Delta^d)$  is such that  $\hat{w}_*^d(q^d) = \sum_q w_*^d(q)\hat{\tau}(q)$  and  $q^d = \sum_q q\hat{\tau}(q)$ , then  $w_*^d(q) \geq h(q)$  for all  $q \in \operatorname{supp} \hat{\tau}$  with strict inequality for some q.

*Proof.* Again by Lemmas 3 and Claim 5,  $\tau^c \in (0,1)$  and

$$\begin{array}{lcl} \hat{w}(\sigma) & = & \tau^c \hat{w}^c(q^c) + (1 - \tau^c) \hat{w}^d_*(q^d) \\ & = & \tau^c q^c(\mathcal{R}) \hat{w}^c(q^c(\cdot|\mathcal{R})) + \tau^c \sum_{\omega \in \overline{\mathcal{R}}} u^*_S(\omega) q^c(\omega) + (1 - \tau^c) \hat{w}^d_*(q^d) \end{array}$$

with

$$\sigma = \tau^c q^c(\mathcal{R}) q^c(\cdot | \mathcal{R}) + \sum_{\omega \in \overline{\mathcal{R}}} \delta_{\omega} \{ \tau^c q^c(\omega) + (1 - \tau^c) q^d(\omega) \}.$$

Therefore,  $\sigma(\omega) = \tau^c q^c(\omega) + (1 - \tau^c) q^d(\omega)$  for all  $\omega \in \overline{\mathcal{R}}$  and hence  $\tau^c q^c(\mathcal{R}) = \sigma(\mathcal{R})$  and  $q^c(\cdot|\mathcal{R}) = \sigma(\cdot|\mathcal{R})$ . Note also that  $(1 - \tau^c) + \tau^c \sum_{\omega \in \overline{\mathcal{R}}} q^c(\omega) = \sigma(\overline{\mathcal{R}})$  and therefore

$$\sigma = \sigma(\mathcal{R})\sigma(\cdot|\mathcal{R}) + \sigma(\overline{\mathcal{R}}) \left[ \frac{\tau^c q^c(\overline{\mathcal{R}})}{\sigma(\overline{\mathcal{R}})} \sum_{\omega \in \overline{\mathcal{R}}} \delta_{\omega} q^c(\omega|\overline{\mathcal{R}}) + \frac{1 - \tau^c}{\sigma(\overline{\mathcal{R}})} q^d \right].$$

Hence, for every  $\omega \in \overline{\mathcal{R}}$ 

$$\delta_{\omega} \frac{\tau^{c} q^{c}(\omega)}{\sigma(\overline{\mathcal{R}})} + \frac{1 - \tau^{c}}{\sigma(\overline{\mathcal{R}})} q^{d}(\omega) = \sigma(\omega | \overline{\mathcal{R}}).$$

So, we obtain

$$\hat{w}(\sigma) = \sigma(\mathcal{R})\hat{w}^c(\sigma(\cdot|\mathcal{R})) + \sigma(\overline{\mathcal{R}})\underbrace{\left[\gamma h(q^c(\cdot|\overline{\mathcal{R}})) + (1-\gamma)\hat{w}_*^d(q^d)\right]}_{\xi^*},\tag{24}$$

where 
$$\gamma = \frac{\tau^c q^c(\overline{\mathcal{R}})}{\sigma(\overline{\mathcal{R}})}$$
 and  $\gamma q^c(\cdot|\overline{\mathcal{R}}) + (1 - \gamma)q^d = \sigma(\cdot|\overline{\mathcal{R}})$ .

Now consider any  $\hat{\tau} \in \Delta(\Delta^d)$  such that  $\hat{w}_*^d(q^d) = \sum_q w_*^d(q)\hat{\tau}(q)$  and  $q^d = \sum_q q\hat{\tau}(q)$ . Suppose  $w_*^d(q') < h(q')$  for any  $q' \in \operatorname{\mathbf{supp}} \hat{\tau}$ . Then,

$$\sum_{q} w_*^d(q)\hat{\tau}(q) < \sum_{\{q: q \neq q'\}} w_*^d(q)\hat{\tau}(q) + \hat{\tau}(q') \sum_{\omega \in \overline{\mathcal{R}}} u_S^*(\omega)q'(\omega).$$

But this implies the existence of  $\tau \in \mathcal{F}_{\sigma}$  with  $V(\tau) > W_{\sigma}$ , a contradiction. So, for all  $q \in \operatorname{supp} \hat{\tau}$ ,  $w_*^d(q) \ge h(q)$ . Finally, suppose  $w_*^d(q) = h(q)$  for all  $q \in \operatorname{supp} \hat{\tau}$ . Then  $\xi^*$  becomes

$$\gamma h(q^c(\cdot|\overline{\mathcal{R}})) + (1 - \gamma) \sum_q h(q)\hat{\tau}(q) = \sum_{\omega \in \overline{\mathcal{R}}} u_S^*(\omega) \sigma(\omega|\overline{\mathcal{R}}),$$

and hence  $\hat{w}(\sigma) = \hat{w}^c(\sigma)$  by Corollary 3, contradicting  $\hat{w}(\sigma) > \hat{w}^c(\sigma)$ .

Similarly, in (24) we must have  $h(q^c(\cdot|\overline{\mathcal{R}})) \geq w_*^d(q^c(\cdot|\overline{\mathcal{R}}))$  because otherwise it would again be possible to improve upon  $W_{\sigma}$ . Hence,  $\xi^*$  in (24) belongs to the set  $\{\xi : (\sigma(\cdot|\overline{\mathcal{R}}), \xi) \in \text{co}(\max\{h, w_*^d\})\}$  and must equal to its maximum, which exists and equals  $\hat{m}(\sigma(\cdot|\overline{\mathcal{R}}))$ .

Claim 7.  $\tau^c = \sigma(\mathcal{R})$  if and only if  $\hat{m}(\sigma(\cdot|\overline{\mathcal{R}})) = \hat{w}^d(\sigma(\cdot|\overline{\mathcal{R}}))$ .

Proof. If  $1 - \tau^c = \sigma(\overline{\mathcal{R}})$ , then  $\gamma = 0$  in (24). This implies that  $q^c(\overline{\mathcal{R}}) = 0$ ,  $q^d = \sigma(\cdot|\overline{\mathcal{R}})$ , and  $\xi^* = \hat{w}^d_*(q^d)$ ; moreover, since  $\sigma(\cdot|\overline{\mathcal{R}}) \in int\Delta^d$ ,  $\hat{w}^d_*(\sigma(\cdot|\overline{\mathcal{R}})) = \hat{w}^d(\sigma(\cdot|\overline{\mathcal{R}}))$  by Lemma 4. Conversely, suppose  $\hat{m}(\sigma(\cdot|\overline{\mathcal{R}})) = \hat{w}^d(\sigma(\cdot|\overline{\mathcal{R}}))$ . Then,  $\xi^* = \hat{w}^d_*(\sigma(\cdot|\overline{\mathcal{R}}))$  in (24) and hence  $\gamma = 0$ , which implies that  $q^c(\overline{\mathcal{R}}) = 0$  and hence  $\sigma(\overline{\mathcal{R}}) = \tau^c q^c(\overline{\mathcal{R}}) + 1 - \tau^c = 1 - \tau^c$ .

#### B.11 Proof of Proposition 7

To prove the result, we need the following observation.

Claim 8. For every  $\omega \in \overline{\mathbb{R}}$ ,  $w_*^d(\delta_\omega) = w^d(\delta_\omega)$ .

*Proof.* Fix any  $\omega \in \overline{\mathbb{R}}$ . It is enough to show that  $p(\cdot)$  in (7) is continuous at  $\delta_{\omega}$ , where of course  $p(\delta_{\omega}) = \delta_{\omega}$ . Indeed, this implies that  $w^d$  is upper semicontinuous at  $\delta_{\omega}$  and hence coincides with  $w^d_*$  (see Footnote 29). Recall that  $\operatorname{supp} \mu$  is finite an contains a least one  $\rho$  with  $\rho(\omega) > 0$ 

by A1. Consider  $\hat{\mu}(\cdot|q)$  in (4) and note that, for every  $\rho \in \operatorname{supp} \mu$ ,  $\lim_{q \to \delta_{\omega}} \hat{\mu}(\rho;q) = \phi(\rho;\omega)$  where  $\phi(\rho;\omega) = \rho(\omega)\mu(\rho) \left[ \sum_{\rho' \in \operatorname{supp} \mu} \rho'(\omega)\mu(\rho') \right]^{-1}$ . Clearly, if  $\rho$  satisfies  $\rho(\omega) = 0$ , then  $\phi(\rho;\omega) = 0$  and hence there exists  $\varepsilon(\rho) > 0$  such that  $||q - \delta_{\omega}|| < \varepsilon(\rho)$  implies  $\rho \notin M(q)$ . Let  $\varepsilon = \min_{\{\rho \in \operatorname{supp} \mu: \rho(\omega) = 0\}} \varepsilon(\rho) > 0$ . Then, for any  $q \in \Delta^d$  such that  $||q - \delta_{\omega}|| < \varepsilon$ ,  $\rho \in M(q)$  implies that  $\rho(\omega) > 0$ . So, hereafter restrict attention to  $K = \{\rho \in \operatorname{supp} \mu: \rho(\omega) > 0\}$ . For every  $\rho \in K$  and  $q \in \Delta^d$ , let

$$\tilde{p}(\hat{\omega}; q, \rho) = q(\hat{\omega}) \frac{\rho(\hat{\omega})}{\sigma(\hat{\omega})} \left[ \sum_{\omega' \in \Omega} q(\omega') \frac{\rho(\omega')}{\sigma(\omega')} \right]^{-1} \quad \text{for all } \hat{\omega} \in \overline{\mathcal{R}}.$$

Clearly,  $\lim_{q\to\delta_{\omega}}\tilde{p}(\hat{\omega};q,\rho)$  equals 1 if  $\hat{\omega}=\omega$  and 0 if  $\hat{\omega}\neq\omega$ . Now define  $\underline{p}(\omega;q)=\min_{\rho\in K}\tilde{p}(\omega;q,\rho)$  and, for  $\hat{\omega}\neq\omega$ , let  $\overline{p}(\hat{\omega};q)=\max_{\rho\in K}\tilde{p}(\hat{\omega};q,\rho)$ . Then, for every q with  $||q-\delta_{\omega}||<\varepsilon$  we have  $\underline{p}(\omega;q)\leq p(\omega;q)$  and  $\overline{p}(\hat{\omega};q)\geq p(\hat{\omega};q)$  for every  $\hat{\omega}\neq\omega$ . Therefore,  $\lim_{q\to\delta_{\omega}}p(\omega;q)\geq\lim_{q\to\delta_{\omega}}\underline{p}(\omega;q)=1$  and, for every  $\hat{\omega}\neq\omega$ ,  $\lim_{q\to\delta_{\omega}}p(\hat{\omega};q)\leq\lim_{q\to\delta_{\omega}}\overline{p}(\hat{\omega};q)=0$ . This shows that  $p(\cdot)$  is continuous at  $\delta_{\omega}$ .

We can now prove Proposition 7. By Proposition 6, conditional on event  $\overline{\mathcal{R}}$ , Sender's expected payoff equals  $\hat{m}(\sigma(\cdot|\overline{\mathcal{R}})) = (1-\gamma)h(q_1) + \gamma \hat{w}_*^d(q_2)$  with  $q_1, q_2 \in \Delta^d$  and  $(1-\gamma)q_1 + \gamma q_2 = \sigma(\cdot|\overline{\mathcal{R}})$ . Suppose that  $w^d(\delta_{\omega'}) > u_S^*(\omega')$  for some  $\omega' \in \overline{\mathcal{R}}$  and  $\omega'$  is hidden with positive probability. Letting  $\hat{\tau} \in \Delta(\Delta^d)$  be such that  $\mathbb{E}_{\hat{\tau}}[q] = q_2$  and  $\mathbb{E}_{\hat{\tau}}[w_*^d(q)] = \hat{w}_*^d(q_2)$ , we have

$$\sum_{\omega \neq \omega'} u_S^*(\omega)(1 - \gamma)q_1(\omega) + u_S^*(\omega')(1 - \gamma)q_1(\omega') + \sum_q w_*^d(q)\gamma\hat{\tau}(q)$$

$$= \sum_{\omega \neq \omega'} h(\delta_\omega)\tau'(\delta_\omega) + h(\delta_{\omega'})\tau'(\delta_{\omega'}) + \sum_q w_*^d(q)\tau'(q),$$
(25)

where  $\tau'(\delta_{\omega}) = (1 - \gamma)q_1(\omega)$  for the first summation and  $\tau'(q) = \gamma \hat{\tau}(q)$  for the second. So, for every  $\omega \in \overline{\mathcal{R}}$ ,

$$\mathbb{E}_{\tau'}[q(\omega)] = \tau'(\delta_{\omega})\delta_{\omega} + \gamma \mathbb{E}_{\hat{\tau}}[q(\omega)] = (1 - \gamma)q_1(\omega) + \gamma q_2(\omega) = \sigma(\cdot|\overline{\mathcal{R}}).$$

Since by assumption  $\tau'(\delta_{\omega'}) > 0$ , (25) is strictly less than  $\mathbb{E}_{\tau'}[m(q)]$ . But this leads to a contradiction since  $\hat{m}(\sigma(\cdot|\overline{\mathcal{R}})) = \max_{\tau \in \mathcal{F}_{\sigma(\cdot|\overline{\mathcal{R}})}} \mathbb{E}_{\tau}[m(q)]$ .

For the second part, note that if  $w^d$  is convex, so is its closure  $w^d_*$  (see Rockafellar). Let  $\overline{\mathcal{R}}_{<} = \{\omega : w^d(\delta_\omega) < u^*_S(\omega)\}$  and  $\overline{\mathcal{R}}_{\geq} = \{\omega : w^d(\delta_\omega) \geq u^*_S(\omega)\}$ . Suppose that  $q^d$  in (15) satisfies  $\operatorname{supp} q^d \cap \overline{\mathcal{R}}_{<} \neq \varnothing$ . Then, as noted before,  $\hat{m}(\sigma(\cdot|\overline{\mathcal{R}})) = (1 - \gamma)h(q_1) + \gamma \hat{w}^d_*(q^d)$  which can be

written as

$$(1 - \gamma) \sum_{\omega} u_S^*(\omega) q_1(\omega) + \gamma \sum_{\omega} w^d(\delta_{\omega}) q^d(\omega)$$

$$< (1 - \gamma) \sum_{\omega} u_S^*(\omega) q_1(\omega) + \gamma \sum_{\omega \in \overline{\mathcal{R}}_{<}} u_S^*(\omega) q^d(\omega) + \gamma \sum_{\omega \in \overline{\mathcal{R}}_{\geq}} w^d(\delta_{\omega}) q^d(\omega)$$

$$= \sum_{\omega \in \overline{\mathcal{R}}_{<}} u_S^*(\omega) \sigma(\omega | \overline{\mathcal{R}}) + \sum_{\omega \in \overline{\mathcal{R}}} w^d(\delta_{\omega}) \sigma(\omega | \overline{\mathcal{R}}),$$

where the last equality follows because  $(1 - \gamma)q_1 + \gamma q^d$  must equal  $\sigma(\cdot|\overline{\mathcal{R}})$ . However, if  $w_*^d$  is convex, so is m. Hence, the last expression equals  $\hat{m}(\sigma(\cdot|\overline{\mathcal{R}}))$ , which is a contradiction.

#### B.12 Proof of Corollary 5

First, observe that

$$\sum_{\omega \in \Omega} u_S(a(p(q')), \omega) \hat{q}(\omega) \le \max_{a \in \mathcal{A}(p(q'))} \sum_{\omega \in \Omega} u_S(a, \omega) \hat{q}(\omega) \le \sum_{\omega \in \Omega} u_S^*(\omega) \hat{q}(\omega) = h(\hat{q}).$$

Hence, in the expression of  $V(\tau)$ 

$$\sum_{q} w^d(q)\tau(q|D_{\tau}) < \sum_{q \neq \hat{q}} w^d(q)\tau(q|D_{\tau}) + h(\hat{q})\tau(\hat{q}|D_{\tau}),$$

which implies that there exists  $\tau' \in \mathcal{F}_{\sigma}$  such that  $V(\tau) < V(\tau') \leq \hat{w}(\sigma)$ .

#### B.13 Proof of Corollary 6

For  $\kappa \leq 2$ , we already know that Sender fully reveals every  $\omega_i$  with  $i \leq \overline{m}$ . For  $i > \overline{m}$ , since  $w^d$  is convex, by Proposition 7  $\omega_i$  is hidden with probability 1 if  $w^d(\delta_{\omega_i}) < h(\delta_{\omega_i})$  and is never hidden otherwise. So fix  $i > \overline{m}$ . For each value of  $\kappa$  there exists a value  $b_i(\kappa)$  such that  $w^d(\delta_{\omega_i}) \geq h(\delta_{\omega_i})$  if and only if  $b \leq b_i(\kappa)$ : this threshold is given by

$$b_i(\kappa) = \max\{\omega_i - \frac{\kappa}{2}(\omega_{\overline{m}} + \omega_i), 0\} = \max\{(1 - \frac{\kappa}{2})\omega_i - \frac{\kappa}{2}\omega_{\overline{m}}, 0\}.$$

Each  $b_i(\kappa)$  is decreasing in  $\kappa$  (strictly when positive) and  $b_i(\kappa) \leq b_j(\kappa)$  if and only if i < j (with < if either threshold is positive). If  $b \leq b_{\overline{m}+1}(\kappa)$ , we have that  $\hat{w}^d(\sigma(\cdot|\overline{\mathcal{R}})) = \hat{m}(\sigma(\cdot|\overline{\mathcal{R}}))$  and hence  $\tau^d = \sigma(\overline{\mathcal{R}})$  by Proposition 6. On the other hand, if  $b > b_{\overline{m}+1}(\kappa)$ , let  $i^*(b,\kappa) = \min\{i > \overline{m} : b_i(\kappa) \geq b\}$ , which is non-decreasing in both  $\kappa$  and b. Then, it is optimal to hide with probability 1 every  $\omega_i$  with  $\overline{m} < i < i^*(b,\kappa)$  and fully reveal all others states in  $\overline{\mathcal{R}}$ .

# B.14 Proof of Corollary 7

By Proposition 6,  $\hat{m}(\sigma(\cdot|\overline{\mathcal{R}}))$  is given by

$$\max_{\gamma \in [0.1], q_1, q_2 \in \Delta^d} \gamma \sum_{\omega \in \overline{\mathcal{R}}} u_S^*(\omega) q_1(\omega) + (1 - \gamma) \sum_{\omega \in \overline{\mathcal{R}}} \{-(\mathbb{E}_{q_2}[\beta] - \omega)^2\} q_2(\omega)$$

subject to  $\gamma q_1 + (1 - \gamma)q_2 = \sigma(\cdot | \overline{\mathcal{R}})$ . By continuity of h and  $w^d$ , a solution  $(\gamma, q_1, q_2)$  to this problem exists. Recall that  $\omega_i \in \text{supp } q_2$  for all  $i < \underline{m}$ . Suppose that  $(\gamma, q_1, q_2)$  implies  $\beta(\omega_{\underline{m}}) \leq \mathbb{E}_{q_2}[\beta] \leq \beta(\omega_{\overline{m}})$ . We will show that there exists a feasible  $(\gamma', q_1', q_2')$  which strictly dominates  $(\gamma, q_1, q_2)$ . Since  $\beta$  is strictly increasing, we must have  $\omega_i, \omega_j \in \text{supp } q_2$  for some  $i < \underline{m}$  and  $j > \overline{m}$ . Suppose first that  $\omega > \beta(\omega_{\underline{m}})$  for some  $\omega > \omega_{\overline{m}}$ . Then, for any  $\xi \in [\beta(\omega_{\underline{m}}), \beta(\omega_{\overline{m}})]$ ,

$$-\sum_{\omega \in \overline{\mathcal{R}}} (\xi - \omega)^2 q_2(\omega) < -\sum_{\omega \in \overline{\mathcal{R}}} (\beta(\omega_{\underline{m}}) - \omega)^2 \mathbf{1} \{\omega < \beta(\omega_{\underline{m}})\} q_2(\omega) - \sum_{\omega \in \overline{\mathcal{R}}} (\beta(\omega_{\overline{m}}) - \omega)^2 \mathbf{1} \{\omega > \beta(\omega_{\overline{m}})\} q_2(\omega) = \sum_{\omega \in \overline{\mathcal{R}}} u_S^*(\omega).$$

This means that  $\gamma' = 1$  and  $q'_1 = \sigma(\cdot|\overline{\mathcal{R}})$  strictly dominates  $(\gamma, q_1, q_2)$ . Now, suppose that  $\omega \leq \beta(\omega_{\underline{m}})$  for all  $\omega > \omega_{\overline{m}}$ . If  $\beta(\omega_{\underline{m}}) < \mathbb{E}_{q_2}[\beta] \leq \beta(\omega_{\overline{m}})$ , then again  $\gamma' = 1$  and  $q'_1 = \sigma(\cdot|\overline{\mathcal{R}})$  strictly dominates  $(\gamma, q_1, q_2)$ . If  $\mathbb{E}_{q_2}[\beta] = \beta(\omega_{\underline{m}})$ , then  $\hat{m}(\sigma(\cdot|\overline{\mathcal{R}})) = h(\sigma(\cdot|\overline{\mathcal{R}}))$ . But we know that always hiding all states in  $\overline{\mathcal{R}}$  is not optimal:  $\hat{w}(\sigma) > \hat{w}^c(\sigma)$  since  $w^d(\delta_{\omega_1}) > h(\delta_{\omega_1})$ . Therefore,  $(\gamma, q_1, q_2)$  is again strictly dominated.

Finally, if  $\sigma$  is such that  $\omega_{\underline{m}} \leq \mathbb{E}_{\sigma(\cdot|\overline{\mathcal{R}})}[\omega] \leq \omega_{\overline{m}}$ , then  $\tau^d = \sigma(\overline{\mathcal{R}})$  implies that  $q^d = \sigma(\cdot|\overline{\mathcal{R}})$  and hence  $\beta(\omega_{\underline{m}}) \leq \mathbb{E}_{\sigma(\cdot|\overline{\mathcal{R}})}[\beta] \leq \beta(\omega_{\overline{m}})$ , which cannot be optimal as we have just argued. Moreover, hiding can only occur for states above  $\omega_{\overline{m}}$ .

# B.15 Proof of Corollary 8

Given any  $\tau \in \mathcal{F}_{\sigma}$ , the claim follows immediately from (7) for every  $q \in C_{\tau}$ . Now consider any  $q \in D_{\tau}$ . If  $q \in D_{\tau}(\rho_N)$ , then again the claim follows directly from (7). So suppose  $q \in D_{\tau}(\rho_i)$  for some  $i \neq N$ . Then,

$$p^A(\omega;q) = \frac{q(\omega)\frac{\rho_1^A(\omega)}{\sigma(\omega)}}{\sum_{\omega' \in \Omega} q(\omega')\frac{\rho_1^A(\omega')}{\sigma(\omega')}} \quad \text{and} \quad p^B(\omega;q) = \frac{q(\omega)\frac{\rho_i^B(\omega)}{\sigma(\omega)}}{\sum_{\omega' \in \Omega} q(\omega')\frac{\rho_i^B(\omega')}{\sigma(\omega')}}.$$

Note that  $q(\omega) = 0$  implies both  $p^A(\omega; q) = 0$  and  $p^B(\omega; q) = 0$ . So, restrict attention to  $\operatorname{supp} q$  which contains both  $\operatorname{supp} p^A(q)$  and  $\operatorname{supp} p^B(q)$ . By definition,  $\rho_1^A(\omega) > 0$  for all  $\omega \in \Omega$ ; therefore,  $\operatorname{supp} q = \operatorname{supp} p^A(q)$ . Also, by definition,  $\operatorname{supp} q \subset \operatorname{supp} \rho_i^B = \Omega(q)$ ; therefore,

 $\operatorname{supp} q = \operatorname{supp} p^B(q)$  as well. Now, restrict attention to  $\Omega(q) \subsetneq \Omega = \operatorname{supp} \rho_1^A$ . By A6, we have

$$\frac{\rho_1^A(\omega|\Omega(q))}{\rho_1^A(\omega'|\Omega(q))} = \frac{\rho_i^B(\omega)}{\rho_i^B(\omega')}$$

for all  $\omega, \omega' \in \Omega(q)$ , where  $\rho_1^A(\omega|\Omega(q)) = \frac{\rho_1^A(\omega)}{\rho_1^A(\Omega(q))}$ . So, fixing one  $\hat{\omega} \in \Omega(q)$ , we have

$$\rho_1^A(\omega|\Omega(q)) = \frac{\rho_1^A(\hat{\omega}|\Omega(q))}{\rho_i^B(\hat{\omega})} \rho_i^B(\omega).$$

It follows that, for all  $\omega \in \operatorname{supp} q$ ,

$$\begin{split} p^A(\omega;q) &= \frac{\rho_1^A(\Omega(q))q(\omega)\frac{\rho_1^A(\omega)}{\sigma(\omega)}}{\rho_1^A(\Omega(q))\sum_{\omega'\in\Omega_i}q(\omega')\frac{\rho_1^A(\omega')}{\sigma(\omega')}} = \frac{q(\omega)\frac{\rho_1^A(\omega|\Omega(q))}{\sigma(\omega)}}{\sum_{\omega'\in\Omega(q)}q(\omega')\frac{\rho_1^A(\omega'|\Omega(q))}{\sigma(\omega')}} \\ &= \frac{\frac{q(\omega)}{\sigma(\omega)}\frac{\rho_1^A(\hat{\omega}|\Omega(q))}{\rho_i^B(\hat{\omega})}\rho_i^B(\omega)}{\sum_{\omega'\in\Omega(q)}\frac{q(\omega')}{\sigma(\omega')}\frac{\rho_i^A(\hat{\omega}|\Omega(q))}{\rho_i^B(\hat{\omega})}\rho_i^B(\omega')} = \frac{\frac{q(\omega)}{\sigma(\omega)}\rho_i^B(\omega)}{\sum_{\omega'\in\Omega(q)}\frac{q(\omega')}{\sigma(\omega')}\rho_i^B(\omega')} = p^B(\omega;q). \end{split}$$

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