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Optimal Law Enforcement with Sophisticated and Naive Offenders

Abstract

Research in criminology has shown that the perceived risk of apprehension often differs substantially from the true level. To account for this insight, we extend the standard economic model of law enforcement (Becker, 1968) by considering two types of offenders, sophisticates and naïves. The former are always fully informed about the enforcement effort, the latter become informed only when the effort is revealed by the authority; otherwise, naïves rely on their perceptions. We characterize the optimal enforcement effort and the decision whether to hide or reveal it. The welfare-maximizing authority chooses either a relatively high effort which is then revealed, or it chooses a relatively low effort which remains hidden. The latter policy becomes more favorable, the larger the share of naïves in the population and the higher their level of perceived effort. We then analyze three empirically important extensions, thereby allowing for lower efficacy of the enforcement effort due to avoidance activities, endogenous fines, and heterogeneity with respect to naïves' perceptions.

JEL-Codes: K420, D730.

Keywords: optimal law enforcement, deterrence, behavioral law & economics, naïveté, shrouding.

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1 Introduction

In the economic literature on law enforcement in the tradition of Becker (1968), an illegal act will be committed if and only if the offender's benefit exceeds the expected punishment (i.e., the probability of apprehension times the respective sanction). Thereby, it is assumed that potential offenders are either fully informed about the actual enforcement effort or form unbiased beliefs in case of uncertainty (see the survey by Polinsky and Shavell, 2007). This assumption is challenged by numerous studies in criminology which find that the correlation between the perceived and the actual probability of apprehension is often weak (e.g., Pogarsky and Piquero, 2003; Lochner, 2007) and that even the perceived manpower of police patrolling on streets often differs largely from its true level (e.g., Kleck and Barnes, 2014). As perceptions are influenced by many factors, they may well be biased in either direction.¹

To account for this these insights, we extend the standard economic model of law enforcement by allowing for biased perceptions of the authority's effort. Building on a recent literature in behavioral industrial organization (Gabaix and Laibson, 2006; Heidhues, Kőszegi, and Murooka, 2012; Armstrong and Vickers, 2012; Heidhues, Kőszegi, and Murooka, 2017), we distinguish between two types of offenders, *sophisticates* and *naïves*. Sophisticated offenders correctly perceive the authority's enforcement effort and the resulting probability of apprehension, and hence behave as in the standard economic model of law enforcement. By contrast, naïve offenders perceive the actual enforcement effort only when it is explicitly revealed by the authority. Otherwise, they have a given perception of the effort, which is independent of the actual level. As a result, naïves may over- or underestimate the probability of apprehension (Kleck et al., 2005), and the authority has the possibility to inform potential offenders about its actual effort (Apel, 2013; Chalfin and McCrary, 2017).² We then analyze how the share of naïve offenders and their perceptions affect the optimal enforcement effort and the decision whether or not to reveal it.

In a first step, we characterize the authority's optimal enforcement effort when it is not revealed. As the effort then affects only the deterrence of sophisticates, the optimal level

¹See Apel and Nagin (2011), Nagin (2013), and Chalfin and McCrary (2017) for comprehensive surveys of this literature, and Levitt and Miles (2007) for a survey of the empirical economic literature on criminal punishment.

²For instance, the public transport authority of the German city of Frankfurt recently announced that the number of ticket inspectors has increased by 52%; and the information was prominently reported in newspapers and reliable internet sources, see e.g., http://www.fr.de/rhein-main/verkehr/rmv-und-vgf-schaerfere-kontrollen-gegen-schwarzfahrer-a-567360, lastly retrieved on May, 28, 2018.

decreases in the share of naïves. Social welfare is U-shaped in the share of naïves as long as their perceived effort is not too large. Intuitively, when the share of naïves is small, the optimal effort exceeds the one perceived by naïves, so that both overall deterrence and social welfare decrease with the share of naïves in the population. Conversely, when the share of naïves is large, the optimal effort is below the perceived effort of naïves, and both deterrence and social welfare increases as the share of naïves increases.

In a second step, we study whether the authority prefers to reveal or hide its enforcement effort and how this is related to its optimal effort choice. With hiding, effort matters only for sophisticates but not for naïves, so that the effort optimally chosen by the authority is lower compared to the regime with revealed effort. This saves on enforcement costs, but it also leads to a low deterrence of sophisticates. In addition, it induces different gain thresholds for sophisticated and naïve offenders, thereby leading to a distortion in the sense that a given number of acts is not committed by the offenders with the highest gains. This distortion does not arise in the regime with revealed effort. These effects driving the regime comparison shift in favor of the regime with hidden effort when the share of naïves in the population is large, and when they perceive the enforcement effort to be high.

Our finding that the optimal policy depends on the percentage and perceptions of naïves is relevant as there is empirical evidence that apprehension probabilities are systematically overestimated for offenses such as robbery and burglary, while they are underestimated for more severe crimes including homicide (Kleck et al., 2005). According to our model, revealing the actual probability of apprehension may rather be optimal in the latter type of situation than for the former one.

We then extend our basic model in three empirically relevant directions. The first extension is motivated by the possibility that revealing the effort facilitates evasion activities. For example, New York City Council recently introduced a policy aimed at increasing the transparency and accountability over the NYPD's use of powerful new surveillance tools.³ The aim of the policy is reduce the risk of information abuse, but officials fear that it may also reduce the efficacy of their effort. More generally, revelation of enforcement effort could foster criminals' activities to avoid sanctions. While there is a literature studying explicitly such strategies (see e.g., Langlais, 2008; Nussim and Tabbach, 2009; Sanchirico, 2006), we focus on how the reduced efficacy under hiding affects the regime

³See e.g., https://www.huffingtonpost.com/entry/new-york-city-is-making-its-citizens-safer-by-overseeing-police-technology_us_58e23f04e4b0ba359596583b, lastly retrieved on May, 28, 2018.

comparison.⁴ When incorporating the model feature that revealed effort reduces its efficacy, this increases the parameter range for which revealing the effort is optimal compared to the baseline model. In particular, a non-monotonicity emerges in the sense that hiding the effort can be optimal for both low shares and high shares of naïves, but not in the intermediate range.

As a second extension, we consider the case where in addition to its enforcement effort and whether to hide or reveal it, the authority can also determine the fine level. When the effort is revealed, the standard argument of Becker (1968) applies, i.e., imposing the maximal fine is optimal. However, when the effort remains hidden, the optimal fine may be strictly lower. The reason is that under the maximum fine, too many naïve offenders with high benefits from the act would be deterred as they overestimate the probability of apprehension. Hence, our analysis adds a further and novel reason why imposing the maximum fine may not be optimal.⁵

In the third extension, we allow for heterogeneity of perceptions by naïve offenders. This is motivated by a number of empirical studies which show that the accuracy of perceptions varies considerably among subjects (Lochner, 2007; Apel, 2013). We first demonstrate that the basic model with homogeneous perceptions is not restrictive in the sense that, for any given enforcement policy and any model with heterogeneous perceptions, there is one with homogeneous perceptions leading to the same welfare level. However, heterogeneity with respect to perceptions reinforces the inefficiency that some offenders with high private benefits may be deterred, while others with low benefits commit the act. As a result, this extension provides an additional argument in favor of revealing the enforcement effort.

Our approach is motivated by empirical findings from criminology, but we still assume that individuals maximize their expected utility, based on their perceptions (accurate or not) of the authority's enforcement effort. In this respect, our paper is related Sah (1991), who considers a rich model with offender perceptions in which crime rates may change over time, but treats the authority's policy as exogenous. Our paper is complementary

⁴Another reason for reduced efficacy of revealing the effort are displacement effects. There is empirical evidence that revealing a high regional concentration of police reduces crime e.g. for motor vehicle thefts (Di Tella and Schargrodsky, 2004; Draca, Machin, and Witt, 2011), but part of the benefits is due to spatial displacements of crime (Donohue, Ho, and Leahy, 2013).

⁵The previous literature has already identified a number of reasons why maximum fines may not be optimal, for example, costs of fine collection, the requirement that the punishment should reflect the severity of the offense, offenders' risk aversion, offenders' heterogeneity with respect to wealth, or offenders who engage in socially undesirable avoidance activities. See the survey by Polinsky and Shavell (2007) for a detailed discussion of these factors.

as we use a more parsimonious framework with regards to perceptions, but we derive the authority's optimal policy.

Moreover, in Bebchuk and Kaplow (1992) and Garoupa (1999) potential offenders get noisy but unbiased signals about the true probability of apprehension. In line with our third extension, the associated heterogeneity of beliefs leads to distortions concerning the decision to commit the act among offenders with low and high private benefits. Bebchuk and Kaplow (1992) show that heterogeneity in beliefs provides a rationale for choosing a fine below the maximum one. The reason is that higher fines aggravate the problem of different perceptions as those fines are multiplied with the probability of apprehension when calculating the expected fine. Garoupa (1999) assumes in addition that the agency can invest in information transmission and derives the optimal investment level. Comparing our paper to Bebchuk and Kaplow (1992) and Garoupa (1999) reveals that it is crucial whether offenders face some uncertainty, but form rational and unbiased beliefs about the true effort as in standard economic models, or whether beliefs are biased in one direction and do not respond to the agency's actual enforcement effort: When assuming that beliefs are unbiased, then heterogeneity is always "bad news" as it misallocates offenses among subjects with high and low private benefits. Revealing the effort is then always beneficial. In our approach, revealing information may be detrimental even when it comes at no cost, and the optimal policy depends on the share of naïve offenders and the direction in which their beliefs are biased.

Ben-Shahar (1997) also analyzes a setting where heterogeneous perceptions trigger offenses by individuals with low private gains who underestimate the true probability of apprehension. Considering a two-period model and assuming that individuals learn the apprehension risk after having been caught once, he shows that the authority has an incentive to set low fines in the first period. This increases first-period arrests, thereby increasing the percentage of individuals who commit the offense in the second period only when their private gains are large.

In our framework, the authority can credibly reveal its detection effort. Conversely, Baumann and Friehe (2013) consider a cheap-talk game and show that credible information transmission is only possible under certain conditions on the levels of harm, sanctions, and the social costs of fines. As in our model, hiding the information may be optimal for the authority, but offenders always benefit from revelation as they can adjust their decisions accordingly. By contrast, naïve offenders may either benefit or suffer in our model when the authority reveals its effort: On the one hand, and similar to Baumann and Friehe (2013), they can adjust their behavior accurately to the actual probability of apprehension. On the other hand, the probability of apprehension is always higher with revelation.

Our focus on the decision whether to hide or reveal the actual enforcement effort is related to studies of so-called crackdowns (Eeckhout, Persico, and Todd, 2010; Lazear, 2006). These are phases and/or regions of very high enforcement effort, e.g., controls for speeding for one day in one part of a city, which are announced in advance to the potential offenders. The announcement leads to more deterrence for the particular group of potential offenders targeted by the enforcement authority, while it might reduce the deterrence of other offenders. Revealing a focused effort is optimal when many potential offenders otherwise perceive the overall probability of apprehension to be low. In our model, this corresponds to the case of a large share of naïves with a low perception of enforcement effort.

We model perceptions that are above or below actual enforcement effort directly, without focusing on why such biased perceptions may arise. The empirical literature in criminology sheds light on these sources. In particular, perceptions seem largely influenced by own experience (Matsueda et al., 2006; Anwar and Loughran, 2011) and by observations in the neighborhood and the social network (Stafford and Warr, 1993; Paternoster and Piquero, 1995; Apel and Nagin, 2011).

Finally, our distinction between two offender types and whether crucial information should be revealed to overcome naiveté is drawn from a recent literature in behavioral industrial organization (Gabaix and Laibson, 2006; Heidhues, Kőszegi, and Murooka, 2012; Armstrong and Vickers, 2012; Heidhues, Kőszegi, and Murooka, 2017), where consumers differ in their ability to fully grasp all attributes of a sales contract. This literature focuses on firm behavior and profits in a competitive environment, and also explores potential welfare implications. In our setting, the decision is taken by a monopolistic authority that acts as a social welfare maximizer.⁶

The remainder of the paper is organized as follows: Section 2 introduces the basic model, while Section 3 characterizes the optimal enforcement policy. Section 4 considers three model extensions: a lower effectiveness of revealed enforcement effort (Section 4.1), endogenous fines (Section 4.2), and heterogeneity with respect to naïves' perceptions (Section 4.3). We conclude and point to further research in Section 5. All proofs are in the Appendix.

⁶Buehler and Eschenbaum (2017) study a model which encompasses both welfare maximization and profit maximization as special cases.

2 Model

Law enforcement is conducted by an authority which takes two decisions: a level of enforcement effort $e \ge 0$, and whether to *hide* (*H*) or to publicly *reveal* (*R*) it to the potential offenders.⁷

There is a unit mass of (risk-neutral) individuals who differ in their gains $g \in \mathbb{R}$ from committing an offense. Gains are distributed according to a cumulative distribution function G, which is twice continuously differentiable and strictly increasing. Each offense leads to a social harm h > 0.

We distinguish two types of offenders. A fraction (1 - a) (with $0 \le a \le 1$) is sophisticated in the sense that they always take into account the authority's true enforcement effort e, irrespective of whether or not it has been revealed. The remaining fraction $a \in [0, 1]$ of offenders is naïve in the sense that they take into account the true enforcement effort e only when it is publicly revealed by the authority.⁸ When it remains hidden, they perceive it to be $\hat{e} \ge 0$ instead, which we take as exogenously given from the viewpoint of the enforcement authority. As mentioned above, the level of \hat{e} might well depend on the type of offense under consideration and other situation-specific factors (Kleck et al., 2005). In the baseline model all naïve agents have the same perception \hat{e} (see Section 4.3 for the case of heterogeneous perceptions). Moreover, the gain distribution G applies to both sophisticated and naïve offenders.

Irrespective of the type, each offender are detected with probability p(e), which satisfies p(0) = 0 and which is twice continuously differentiable and strictly increasing for all effort levels e satisfying p(e) < 1. In deciding whether or not to commit the offense, sophisticates always take into account the actual probability p(e). By contrast, naïves perceive it to be $p(\hat{e})$ when the enforcement effort is hidden, and p(e) when it is revealed. Each detected offender is subject to a fine f > 0. In the basic model, we treat f as exogenous, while the case where it also becomes a choice variable of the enforcement authority is considered in Section 4.2. The cost of enforcement effort e is given by a function C(e), which is twice continuously differentiable, satisfying C(0) = 0, C'(e) > 0, $C''(e) \ge 0$, as well as the Inada conditions C'(0) = 0 and $\lim_{e \to p^{-1}(1)} C'(e) = \infty$.

⁷In the behavioral industrial organization literature discussed above, the revealing (hiding) of the chosen policy is often referred to as unshrouding (shrouding), see e.g., Gabaix and Laibson (2006); Heidhues, Kőszegi, and Murooka (2017).

⁸In this respect, the standard economic model of law enforcement (see e.g., Becker, 1968; Polinsky and Shavell, 2000, 2007) is nested in our model when either the effort is revealed or when all offenders are sophisticates (a = 0).

The sequence of events is as follows: At stage 1, the enforcement authority decides on its effort and on whether to reveal or to hide it. At stage 2, each individual decides on whether or not to commit the offense.

At stage 2 each offender will commit the offense when her gain exceeds the expected (respectively perceived) punishment, i.e., for $g \ge g_T^j$, where the threshold gain g_T^j in general depends on both the regime T = H, R and the offender types j = s, n where s (n)indicates sophisticates and naïves, respectively. Thereby, the deterrence of sophisticates is independent of whether or not the effort is revealed. By contrast, the deterrence of naïves is determined by the true enforcement effort e under regime R and by the perceived effort (\hat{e}) under regime H. In summary, this leads to $g_H^s = g_R^s = g_R^n = p(e) \cdot f$ and $g_H^n = p(\hat{e}) \cdot f$, which is independent of e.

The enforcement authority chooses its policy (T, e) to maximize the overall expected surplus

$$W_T(e) := (1-a) \cdot \left[\int_{p(e) \cdot f}^{\infty} (\pi g - h) G'(g) dg \right] + a \cdot \left[\int_{g_T^n}^{\infty} (\pi g - h) G'(g) dg \right] - C(e), \quad (1)$$

where the first (second) term gives the surplus generated by sophisticates (naïves) and where $\pi \in (0, 1]$ denotes the weight placed on the offenders' gains.⁹ With respect to candidate maximizers of the surplus function (1), the above Inada conditions rule out corner solutions of *e*. Moreover, we also assume that the surplus function (1) is singlepeaked under both regimes T = H, R such that we get a unique interior optimum with respect to the enforcement effort.¹⁰

3 Optimal Enforcement Policy

We consider first the regime where the enforcement authority hides its effort (T = H), so that the two offender types face (different) threshold values, $g_T^s = p(e)f$ and $g_T^n = p(\hat{e})f$.

¹⁰Single-peakedness is for example ensured when the surplus function (1) is globally concave, i.e., when

$$(1-a)f\left[-\pi p'(e) \cdot G'(p(e)f) + (h-\pi p(e)f)\left[G''(p(e)f)(p'^2f \cdot p'(e)f + G'(p(e)f)p''(e)\right]\right] < C''(e)f + C'(p(e)f)p''(e)$$

⁹Most scholars would agree that benefits from severe crimes should not be considered as part of social welfare (see e.g., Stigler, 1970). However, things might be different for smaller offenses such as, for example, violations of environmental standards leading to a monetary gain in the form of a lower production cost. As a result, the gains are usually included (with weight 1) in the social surplus function (see e.g., Polinsky and Shavell, 2000, 2007). Our slightly more general formulation hence allows to capture different forms of offenses or different preferences of the social planner.

holds for all e > 0. For example, this condition is satisfied as long as the distribution of gains G is not too convex or as long as the effort cost function C(e) is sufficiently convex.

The authority therefore chooses its enforcement effort e to maximize

$$W_H(e) := (1-a) \cdot \left[\int_{p(e)f}^{\infty} (\pi g - h) G'(g) dg \right] + a \cdot \left[\int_{p(\hat{e})f}^{\infty} (\pi g - h) G'(g) dg \right] - C(e).$$
(2)

We denote the (unique) maximizer of surplus function (2) by $e_H^*(a)$ and the resulting maximum surplus by $W_H^*(a, \hat{e}) := W_H(e_H^*(a); a, \hat{e}).^{11}$ The interior solution for $e_H^*(a)$ is implicitly given by the first-order condition

$$(1-a)\left[(h-\pi p(e)f) \cdot G'(p(e)f) \cdot p'(e)f\right] = C'(e),$$
(3)

i.e. when the marginal benefit of deterring sophisticates equals the marginal effort cost.

Consider next the regime where the enforcement authority reveals its effort (T = R), so that it is observed by both offender types. As a result, they both face the same threshold $g_R^s = g_R^n = p(e)f$. The optimal enforcement effort therefore maximizes

$$W_R(e) := \int_{p(e)f}^{\infty} (\pi g - h) G'(g) dg - C(e),$$
(4)

and we denote the (unique) maximizer by e_R^* and the resulting maximum surplus by $W_R^* := W_R(e_R^*)$. Since also naïve offenders learn the actual effort under this regime, both e_R^* and W_R^* are independent of \hat{e} and a. The interior solution e_R^* solves the first-order condition:

$$[(h - \pi p(e)f) \cdot G'(p(e)f) \cdot p'(e)f] = C'(e).$$
(5)

While the exact characterization of $e_H^*(a)$ and e_R^* depend on the details of the cost function C(e), the distribution of gains G and the other model parameters, some general results are nevertheless available. Define $e^{max} := p^{-1}(\frac{h}{\pi f})$ as the enforcement effort under which the indifferent (sophisticated) offender's (weighed) gain just equals the social harm.

Lemma 1. (No Over-Deterrence) For both regimes H and R, the (weighed) gain of the indifferent sophisticated offender resulting under the optimal enforcement effort is below the social harm, i.e. $\pi p(e_H^*(a))f < h$ and $\pi p(e_R^*)f < h$ or, equivalently, $e_H^*(a), e_R^* < e^{max}$.

Lemma 1 corresponds to a well-known result of the literature (see e.g. Polinsky and Shavell, 2007), which focuses on the case where all offenders are sophisticates. Intuitively, because enforcement is costly, some offenses with gains below social harm are not deterred

¹¹Notice that the first integral in the surplus function (2) is independent of \hat{e} (as \hat{e} does not affect the behavior of sophisticates), while the second is independent of e (as the determined of naïves is solely determined by \hat{e}). As a result, $e_H^*(a)$ does not depend on \hat{e} .

in the social optimum. Thereby, the optimal enforcement level decreases in the weight π which the authority puts on the offenders' gains. To avoid uninteresting case distinctions, we assume that the naïves' perceived enforcement effort \hat{e} satisfies the same property.

Assumption 1. The perceived enforcement effort of naïve offenders \hat{e} satisfies $\pi p(\hat{e})f < h$, which is equivalent to $\hat{e} < e^{max}$.

The next result characterizes the properties of the optimal policy under regime H.

Proposition 1. (Optimal Policy under Regime H)

- (i) The optimal (interior) effort level $e_H^*(a)$ is strictly decreasing in the share of naïves a and satisfies $e_H^*(0) = e_R^*$, $e_H^*(a) < e_R^*$ for all $a \in (0, 1]$, and $e_H^*(1) = 0$.
- (ii) For any given share of naïves a > 0, the resulting maximum surplus $W_H^*(a, \hat{e})$ is strictly increasing in the naïves' perceived enforcement effort \hat{e} .
- (iii) When the perceived effort of naïve offenders is sufficiently large (i.e., $\hat{e} > e_H^*(0)$), the social welfare under hiding $W_H^*(a, \hat{e})$ is strictly increasing in the share of naïves a for all $a \in [0, 1]$.
- (iv) Otherwise (i.e., for $0 < \hat{e} < e_H^*(0)$), there exists a threshold for the share of naïves $\hat{a} \in (0,1)$, implicitly defined by $e_H^*(\hat{a}) = \hat{e}$, such that $W_H^*(a, \hat{e})$ is strictly decreasing (increasing) in a for all $a < (>)\hat{a}$.

As for part (i), when there are no naïve agents (a = 0), the two surplus functions (2) and (4) coincide, so that $e_H^*(0) = e_R^*$ (and $W_H^*(0, \hat{e}) = W_R^*$) must hold. Moreover, $e_H^*(a)$ is decreasing in a as the authority's effort matters only for sophisticates. Thus, the optimal effort is always smaller when it is hidden rather than revealed (i.e., $e_H^*(a) < e_R^*$ for all $a \in (0, 1]$). In the polar case where all offenders are naïve (a = 1), the optimal effort is zero as the deterrence for the whole population of offenders no longer depends on it, so that a positive effort level would not lead to more deterrence. Note also that under regime H, the fraction of sophisticates that is deterred from committing the offense decreases in the fraction of naïves: Sophisticates with gains $g > p(e_H^*)f$ benefit from the presence of naïves as they face a lower detection probability than they would if there were no naïves. Part (ii) expresses the fact that a higher enforcement effort as perceived by naïves (\hat{e}) increases their deterrence, while at the same time not affecting the behavior of sophisticates. As a result, the maximum attainable surplus under hiding increases (given Assumption 1). As for part (iii), not even the maximum effort the authority would choose exceeds the perceived effort of naïves $(e_H^*(0) < \hat{e})$. As a result, a larger share of naïves (i.e. more offenders deterred by \hat{e} instead of $e_H^*(a)$) will lead to improved deterrence, and hence the (maximum) social surplus is monotone increasing in a (again under Assumption 1).

As for part (iv), for smaller levels of \hat{e} , $e_H^*(a) < \hat{e}$ only holds when the share of naïves is sufficiently large $(a > \hat{a}(\hat{e}))$. In this case, the effort affects only few sophisticates, so that choosing a high level would be too costly for the authority. In this range, both deterrence and social welfare are again increasing as the share of naïves increases further. By contrast, for smaller shares of naïves $(a < \hat{a}(\hat{e}))$, the optimal effort is relatively large and exceeds the perceived one $(e_H^*(a) > \hat{e})$. A higher share of naïves then reduces the maximum attainable surplus, due to their lower deterrence. As a consequence, welfare $W_H^*(a, \hat{e})$ is U-shaped in a.

Note also that for all $a \neq \hat{a}(\hat{e})$, the thresholds of the indifferent offenders for either type do not coincide (i.e., $g_H^s = p(e_H^*(a))f \neq p(\hat{e})f = g_H^n$). This induces an inefficiency, since any given number of acts is not committed by the offenders with the highest benefits.

In a next step, we characterize the authority's optimal regime choice by comparing the resulting maximum surplus under the optimal enforcement levels $e_H^*(a)$ and e_R^* , respectively:

Proposition 2. (Optimal Regime Choice)

- (i) When the perceived effort of naïve offenders is sufficiently large (i.e., $\hat{e} > e_H^*(0)$), then it is always optimal to hide the effort, i.e., $W_H^*(a, \hat{e}) > W_R^* \ \forall a \in [0, 1]$.
- (ii) Otherwise (i.e, for $\hat{e} < e_H^*(0)$), either regime can be optimal. For $W_H^*(1, \hat{e}) > W_R^*$, there exists a threshold $\tilde{a}(\hat{e}) \in (0, 1)$ implicitly defined by $W_H^*(\tilde{a}(\hat{e}), \hat{e}) = W_R^*$ such that it is optimal to hide (reveal) the effort when the share of naïves is sufficiently large (small), i.e., $W_H^*(a, \hat{e}) > (<) W_R^* \,\forall a > (<) \tilde{a}(\hat{e})$.
- (iii) If $\hat{e} < e_H^*(0)$ and $W_H^*(1, \hat{e}) < W_R^*$ hold, then revealing the effort is always optimal, i.e., $W_H^*(a, \hat{e}) < W_R^* \ \forall a \in [0, 1].$

The proposition is illustrated in Figure 1. Under regime H, the optimal (costly) enforcement effort e_H is only effective for sophisticated offenders, while the naïves are deterred by their perceived effort \hat{e} . By contrast, under regime R, only the actual effort e_R^* matters, while \hat{e} is no longer relevant. As a consequence, hiding the effort is optimal when the share of naïves (a) is large and when the effort perceived by them (\hat{e}) is high.

For $\hat{e} > e_H^*(0)$, the determinance of naïves induced by \hat{e} is so large that it is always optimal to hide the effort. This is the case depicted in panel (i) of Figure 1. Conversely, for low levels of \hat{e} satisfying $\hat{e} < e_H^*(0)$, we know from Proposition 1 that the maximum surplus under hiding $(W_H^*(a, \hat{e}))$ is U-shaped in a. And as social welfare when revealing the effort is independent of the share of naïves (a), revealing the effort is optimal as long as this share is sufficiently small.

Whether or not the regime with hiding eventually becomes optimal for large levels of a depends on whether the value of a (apart from a = 0) where $W_H^*(a, \hat{e}) = W_R^*$ holds lies inside or outside the feasible range $a \in [0, 1]$ (see panels (i) and (iii) of Figure 1). A necessary and sufficient condition for the former case is $W_H^*(1, \hat{e}) > W_R^*$ as stated in the proposition.



Figure 1: Illustration of Proposition 2: Welfare comparison when hiding (W_H^*) respectively revealing (W_R^*) the enforcement effort. The horizontal axis represents the share of naïves *a*. Each panel corresponds to one part of Proposition 2.

A simple implication of this finding is that revealing the effort is helpful when it is underestimated. The empirical literature suggests that this is typically the case for severe crimes including homicide (Kleck et al., 2005). In contrast, for offenses such as robbery and burglary, potential offenders typically underestimate the probability of apprehension. This can be a case for hiding the effort. Importantly, actual effort is endogenous in our model and chosen much lower when hidden than when revealed.

4 Extensions

4.1 Extension A: Revealed Enforcement Effort Reduces its Effectiveness

So far, revealing the enforcement effort only changed the perception (and thereby the deterrence level) of naïve offenders, but not the effectiveness of the actual effort in terms of the detection of offenses. However, it is also argued that revealing the effort might compromise police investigations as this allows offenders to adapt their behavior in order to avoid detection.

We now account for the possibility that the revelation of the effort reduces its effectiveness. In particular, when revealing its effort, the authority detects each offender only with probability $\tilde{p}_R(e)$, where $\tilde{p}_R(e) < p(e) \forall e > 0$. The detection function $\tilde{p}_R(e)$ is assumed to satisfy the same properties as the function p(e). The reduction in effectiveness affects both offender types so that, as in the basic model, the distinction between sophisticates and naïves vanishes when the enforcement effort is revealed.

The enforcement authority's maximization problem is then given by

$$\max_{e} \ W_{R}(e) := \int_{\widetilde{p}_{R}(e)f}^{\infty} (\pi g - h) G'(g) dg - C(e).$$
(6)

We denote the (unique) maximizer by \tilde{e}_R^* and the resulting maximum surplus by $\widetilde{W}_R^* := W_R(\tilde{e}_R^*)$. As in the basic model, \tilde{e}_R^* and \widetilde{W}_R^* are independent of a and \hat{e} . If interior, \tilde{e}_R^* satisfies the first order condition

$$-\left[\left(\pi \tilde{p}_R(e) - h\right) \cdot G'(\tilde{p}_R(e)f) \cdot \tilde{p}'_R(e)f\right] = C'(\tilde{p}_R(e)),\tag{7}$$

which leads to the following Lemma:

Lemma 2. When revealing the effort reduces its effectiveness, both the optimal effort and the resulting maximum surplus are smaller compared to the basic model, i.e. $\tilde{e}_R^* < e_R^*$ and $\widetilde{W}_R^* < W_R^*$ hold.

From Lemma 2, it follows immediately that hiding the effort is strictly superior to revelation when the population of offenders consists of sophisticates only (a = 0). Proposition 3 characterizes the optimal regime choice in more detail:

Proposition 3. (Regime Comparison with Reduced Effectiveness)

(i) For $\hat{e} > e_H^*(0)$, it is optimal to hide the effort, i.e. $W_H^*(a, \hat{e}) > \widetilde{W}_R^* \ \forall a \in [0, 1]$.

- (ii) For $\hat{e} < e_H^*(0)$, it is optimal to hide the effort if the attainable surplus under revealed effort is sufficiently low, i.e. if $\widetilde{W}_R^* < W_H^*(\hat{a}, \hat{e})$, where \hat{a} is the share of naïves at which $e_H^*(\hat{a}) = \hat{e}$.
- (iii) For $\hat{e} < e_H^*(0)$ and $\widetilde{W}_R^* > W_H^*(\hat{a}, \hat{e})$, either regime can be optimal. In particular, for $W_H^*(1, \hat{e}) > \widetilde{W}_R^*$, there exist two thresholds a_1 and a_2 implicitly defined by $W_H^*(a_1, \hat{e}) = W_H^*(a_2, \hat{e}) = \widetilde{W}_R^*$ with $0 < a_1 < a_2 < 1$ such that it is optimal to reveal the effort if the share of naïves is neither too large nor to small, i.e. $\widetilde{W}_R^* > W_H^*(a, \hat{e})$ $\forall a \in (a_1, a_2)$. Otherwise, it is optimal to hide the effort.
- (iv) By contrast, if the two conditions from part (iii) hold, but if $W_H^*(1,\hat{e}) < \widetilde{W}_R^*$ (implying $a_2 > 1$), it is optimal to hide (reveal) the effort if the share of naïves is sufficiently small (large), i.e. $W_H^*(a,\hat{e}) > (<)\widetilde{W}_R^* \forall a < (>)a_1$.

Proposition 3 is illustrated in Figure 2: As in the basic model (see Figure 1 above), hiding the effort is optimal for sufficiently large share of naïves, and/or when their perceived enforcement effort (\hat{e}) is sufficiently large. However, there are also qualitative changes: First, due to $\widetilde{W}_R^* < W_R^*$, the parameter range where hiding the effort is optimal increases. In particular, as shown in part (ii), even when W_H^* is U-shaped, hiding can still be optimal for all shares of naïves a when the negative impact of revelation of the effort on its effectiveness is sufficiently large.

As for parts (iii) and (iv), when hiding is not globally optimal, the regime comparison becomes non-monotonic in the share of naïves (a). In particular, there now also exists an interval of small values of $[0, a_1]$ where hiding is optimal. This qualitative difference to the baseline model (compare with Figure 1) is due to the fact that revealing the effort would indeed improve deterrence for the few naïves, but the deterrence for all sophisticates would decrease due to the lower probability of apprehension ($\tilde{p}(e) < p(e)$). In the interval $[0, a_1]$ this second effect is larger so that hiding is optimal. Furthermore, as in the baseline model, for intermediate values of $a \in [a_1, a_2]$ revealing is optimal, while hiding becomes again optimal for a sufficiently large when $a_2 < 1$ holds.

4.2 Extension B: Endogenous Choice of Fine

We have so far treated the fine f as exogenously given. This is appropriate in settings where the enforcement authority chooses its enforcement effort e, while the fines have been chosen by other parties such as legislators. In other cases, it is the enforcement



Figure 2: Illustration of Proposition 3: Welfare comparison when either hiding the enforcement effort (W_H^*) or revealing it with reduced effectiveness (\widetilde{W}_R^*) (the case of revealing enforcement effort with unchanged effectiveness of the baseline model (W_R^*) is kept as a benchmark). The horizontal axis represents the share of naïves *a*. Each panel corresponds to one part of Proposition 3.

authority which simultaneously decides on both fine and effort. For such settings, the classic insight of Becker (1968) is that any level of deterrence p(e)f > 0 can also be reached with a slightly lower effort and a slightly higher fine. Moreover, such a change leads to higher welfare since increasing the fine is costless, while decreasing the effort saves enforcement costs. As a consequence, it is always optimal to set the largest possible fine.

In this section, we analyze a model extension in which the authority simultaneously decides on both the fine and its enforcement effort. We show that in our setting with sophisticated and naïve offenders, Becker's argument does not always apply, i.e., it might be optimal for the authority to set the fine strictly below its maximum level. (As mentioned in footnote 5 above, the literature has already identified several other scenarios in which the classic reasoning that "fines should be maximal" might not apply.) As for the regime comparison, we find that endogenous fines works in favor of regime H.

Consider an authority that chooses effort $e \ge 0$ and fine $f \in [0, F]$, where the maximal possible fine F might for example by given by law or by the wealth of offenders. The optimization problem of the enforcement authority from the baseline model (see the surplus function (1)) then needs to be adapted as follows:

$$\max_{T,e,f} \quad W_T(e) := (1-a) \cdot \left[\int_{p(e) \cdot f}^{\infty} (\pi g - h) G'(g) dg \right] + a \cdot \left[\int_{g_T^n}^{\infty} (\pi g - h) G'(g) dg \right] - C(e), \quad (8)$$

where the authority can now also affect the respective threshold for the marginal offenders (lower bounds of the integrals) through its choice of f (recall that $g_T^n \in \{p(e)f, p(\hat{e})f\}$). Denoting by e_T^* and f_T^* the respective optimal choices under regime T = R, H, we have the following result:

Proposition 4. (*Endogenous Fine*) When the enforcement authority also chooses the fine $f \in [0, F]$ in addition to its enforcement effort e, then:

- (i) In regime R, the maximal fine is optimal, $f_R^* = F$. All results of the baseline model hold by substituting the exogenous fine \overline{f} with the maximal fine F.¹²
- (ii) In regime H, when the maximal fine is optimal, $f_H^* = F$, then all results of the baseline model hold by substituting the exogenous fine \overline{f} with the maximal fine F.
- (iii) In regime H, when the optimal fine f_R^* is interior (i.e., $f_R^* < F$), then the optimal enforcement effort is below the perceived enforcement effort, i.e., $e_H^* < \hat{e}$. The gain

¹²To emphasize the difference to the baseline model where the fine was exogenous, we now use notation \bar{f} when referring to an exogenous fine.

of the indifferent sophisticated offender is below social harm, while the gain of the indifferent naïve offender is above social harm, i.e., $\pi p(e_H^*) f^* < h < \pi p(\hat{e}) f^*$.

(iv) When regime H leads to higher welfare than regime R in the baseline model for a fixed fine \bar{f} , then this also holds when \bar{f} is the maximal possible fine, i.e. when the fine f is chosen from the interval $[0, F = \bar{f}]$.

Parts (i) and (ii) of the proposition provide an additional justification for considering fixed fines in the baseline model: As the fine optimally chosen is just equal to the maximum amount (which is exogenously given), assuming an exogenous fine in the baseline model can be interpreted as a reduced form.

Part (iii) of the proposition, however, reveals a novel case where the optimal fine is below the maximal one. While increasing small fines is beneficial by deterring more offenders, naïves are over-deterred when the fine becomes too large: Their private benefit from the offense might exceed social harm, but they are nevertheless deterred as the fine has reached the point where $\pi p(\hat{e})f \geq h$. Further increasing the fine then involves a trade-off between deterring more sophisticates, which is still desirable (as $\pi p(e^*)f^* < h$), and deterring more naïves. The optimal interior fine $f^* < F$ satisfies two first order conditions, one of which shows this novel trade-off:

$$(1-a)\left[(h-\pi p(e)f) \cdot G'(p(e)f) \cdot p(e)\right] = a\left[(\pi p(\hat{e})f - h) \cdot G'(p(\hat{e})f) \cdot p(\hat{e})\right], \quad (9)$$

i.e. the marginal benefit of deterring sophisticated agents by increasing fine f (LHS) equals the marginal loss of deterring naïve agents with a high benefit from crime (RHS).¹³

The model with sophisticated and naïve agents thus reveals a new reason for why Becker's classic argument for maximal fines does not always apply. With too high a fine, one would deter inefficiently many offenders who are not aware that effort e^* is actually low. While over-deterrence seems unlikely for severe crimes, it may well be relevant in situations where violating a production standard leads to a large cost reduction, or when not committing an offense leads to high opportunity costs.

Finally, part (iv) of Proposition 4 provides a regime comparison, which favors hiding the enforcement effort. This result follows from the fact that endogenizing the fine gives the enforcer more flexibility. Under regime R, however, the fine is always chosen maximally, so that the authority just replicates the welfare from the baseline model by

¹³The other first order condition reflects the common trade-off between costs and benefits of deterrence: $(1-a) [(h - \pi p(e)f) \cdot G'(p(e)f) \cdot p'(e)f] = C'(e)$, i.e. the marginal benefit of deterring sophisticated agents by increasing effort e equals the marginal effort cost.

choosing $f^* = F = \overline{f}$ and effort optimal as before. Under regime H, this may also be the case, but we have just seen that it may also be optimal to implement a lower fine. Thus, an endogenous fine works in favor of hiding the enforcement effort.

4.3 Extension C: Heterogeneity of Perceptions

In the baseline model, all naïve individuals share the same perception \hat{e} when enforcement effort is hidden. In the following, we first show that the results of the baseline model carry over to heterogeneous perceptions. Then, we demonstrate that heterogeneity sets additional incentives to reveal the effort.

Let there be L groups of naïves with perceptions $\hat{e}_1, ..., \hat{e}_L$ and group sizes $a_1, ..., a_L$. All other model features are as in the baseline model (see Section 2). In particular, gains from crime are distributed according to a cdf G for the sophisticated agents as well as for all groups of naïve individuals. Furthermore, all naïve individuals learn the actual effort e_R in case of revelation.¹⁴ Finally, we extend Assumption 1 to all perceptions \hat{e}_l i.e., $0 < \hat{e}_l < e^{max}$ holds for any group l. As the marginal offender's gain from crime is thus below social harm, deterrence is socially desirable.

Welfare under revealing $W_R(e)$ is unaffected from heterogeneity of perceptions, so that the maximum welfare in this case is still $W_R(e_R^*) = W_R^*$. Welfare under regime H is now given by:

$$W_H(e) := \left(1 - \sum_{l=1}^{L} a_l\right) \cdot \left[\int_{p(e)f}^{\infty} (\pi g - h)G'(g)dg\right] + \sum_{l=1}^{L} a_l \cdot \left[\int_{p(\hat{e}_l)f}^{\infty} (\pi g - h)G'(g)dg\right] - C(e),$$
(10)

which is a straightforward generalization of the surplus function (2).

To analyze the impact of different degrees of heterogeneity, the following definition is useful. Two models are referred to as *welfare equivalent* when they lead to the same welfare when the same policies (T, e) are chosen in the two models.

Proposition 5. (Reduction of General Model to Baseline Model) For every model with heterogeneous perceptions $(\hat{e}_1, ..., \hat{e}_L)$ of the fractions $(a_1, ..., a_L)$ of naïve agents, there is a unique model with a homogeneous perception \tilde{e} of the fraction of naïve agents a := $\sum_{l=1}^{L} a_l$ that is welfare equivalent to it.

¹⁴We assume that revelation of effort is public in the sense that it is impossible to reveal it to just some groups of naïves. While partial revelation could never be optimal in the baseline model, there could now be an incentive to reveal the effort only to naïves who underestimate the effort.

Intuitively, any model with heterogeneous perceptions can be mirrored by the unique (baseline) model with homogeneous perceptions where $\hat{e} = \tilde{e}$ and $a = \sum_{l=1}^{L} a_l$. This implies that all results for this baseline model (Section 3) carry over. In particular, Proposition 1 characterizes the optimal policy under hiding, and Proposition 2 characterizes the optimal regime choice. Moreover, we can generate some additional comparative statics insights by studying how the parameters of the extended model, with $(\hat{e}_1, ..., \hat{e}_L)$ and $(a_1, ..., a_L)$, affect the parameters of the welfare equivalent model, with a and \tilde{e} . This leads to the following observations. First, as the optimal effort e_H^* depends only on the percentage of sophisticated agents, it is strictly decreasing in a_l for each group l. Second, welfare W_H^* is strictly decreasing (increasing) in a_l when the group's perception satisfies $\hat{e}_l < \hat{e}_H^*$ ($\hat{e}_l > \hat{e}_H^*$). Finally, welfare W_H^* is strictly increasing in the perception $\hat{e}_l(< e^{max})$ of any group l (with $a_l > 0$).

The reduction result expressed in Proposition 5 shows that the insights from of our baseline model are robust. However, we have not yet specified how the degree of heterogeneity affects welfare, i.e. which (welfare-equivalent) homogeneous perception \tilde{e} corresponds to the given heterogeneous perception levels $\hat{e}_1, ..., \hat{e}_L$. It turns out that the perception level \tilde{e} is not simply the mean of the heterogeneous perceptions $\sum_{l=1}^{L} \hat{e}_l$, but lower than that, because increasing the dispersion of perceptions has a deteriorating effect on welfare. To illustrate the intuition behind this insight, we analyze a simple set-up, in which the dispersion of the perceptions is varied while the mean level of perceptions is kept constant.

We study L = 2 groups of equal sizes $(a_1 = a_2 > 0)$ and with perceptions $\hat{e}_1 = \hat{e} - \sigma$ and $\hat{e}_2 = \hat{e} + \sigma$. Let σ be small enough such that $0 < \hat{e}_1 < \hat{e}_2 < e^{max}$. The construction is such that the mean perception level is \hat{e} and the distance of each group to the mean is σ . While we already know that a higher mean of perceived effort \hat{e} is welfare enhancing (since this holds true for the perception \hat{e}_l for any group l), we now turn to the impact of the distance σ . We show that increasing the distance σ reduces welfare with hidden effort W_H^* under very mild assumptions.

Proposition 6. (Heterogeneity of Perceptions Reduces Welfare) Let there be L = 2 groups of naïve agents of equal sizes $(a_1 = a_2)$ and with perceptions $\hat{e}_1 = \hat{e} - \sigma$ and $\hat{e}_2 = \hat{e} + \sigma$. Suppose that the gain distribution G and the detection function p(e) are not too convex (they may well be linear, concave or slightly convex), i.e. $\frac{G'(p(\hat{e}+\sigma)f)}{G'(p(\hat{e}-\sigma)f)} \cdot \frac{\partial p(\hat{e}+\sigma)/\partial \sigma}{-\partial p(\hat{e}-\sigma)/\partial \sigma} < \frac{h-\pi p(\hat{e}+\sigma)f}{h-\pi p(\hat{e}+\sigma)f} (> 1)$. Then, welfare W_H^* under regime H is strictly decreasing in the distance σ of the perceptions to the mean. Heterogeneity among naïve offenders reduces welfare under hiding because it induces different thresholds for the indifferent offenders in each group of naïves $(g_H^{n_2} = p(\hat{e}_2)f > p(\hat{e}_1)f = g_H^{n_1})$. As a result, some naïves with \hat{e}_2 and large gains are deterred, while others with \hat{e}_1 and small gains are not. As in the baseline model, for any given number of offenses the total surplus is highest when the offenses are committed by the offenders with the largest gains. Since this condition is violated for any $\sigma > 0$, there is some inefficiency. Moreover, as σ increases, so does the wedge between the two threshold values and the resulting inefficiency, so that overall welfare decreases. As for the regime comparison, as welfare with revealed effort is independent of the perceptions and their distribution, revealing is more likely to be optimal when the heterogeneity of perceptions is large. Our finding also extends the result by Garoupa (1999) on the benefits of revelation to our setting with naïves and sophisticates. While revelation is always optimal in Garoupa (1999) if it incurs no cost, hiding may still be optimal in our setting if perceptions of naïves tend to be large and are not too dispersed.

Note that different perceptions on enforcement effort would not lead to different welfare if gains from offenses were ignored (i.e. if $\pi = 0$) and the density of benefits was constant, as it would then not make a difference from a social welfare perspective who commits an offense. For several applications such as environmental or product liability, however, it can be argued that private gains do matter as they often come in the form of lower avoidance costs. If firms have different perceptions of the authority's enforcement effort, this is likely to lead to inefficiencies in the form of inefficient care levels. Our results suggest that the larger heterogeneity in perceptions, the more likely is it that revealing the true effort to firms is optimal.

5 Conclusion

The economic literature on law enforcement assumes that potential offenders are either fully informed about the agency's enforcement effort (and, hence, the probability of apprehension) or form unbiased beliefs in case of uncertainty. At the same time, criminologists emphasize that the perceived probability of apprehension differs considerably among individuals and is often not systematically related to the true probability. We propose a model that combines both perspectives by distinguishing between sophisticated and naïve offenders, and characterize the optimal enforcement policy. Thereby, in addition to determining its enforcement effort, the enforcement authority can also decide whether to hide or reveal it to the offenders. We show that the welfare-maximizing authority chooses either a policy (R, e_R^*) in which the enforcement effort (e_R^*) is relatively high and is revealed to the offenders; or a policy (H, e_H^*) in which the enforcement (e_H^*) is relatively low and remains hidden. The reason for the low effort under regime H is that it only affects the deterrence of a fraction of the agents, the sophisticates, whereas under regime R it is effective for all agents. The advantage of hiding is that enforcement costs can be saved due to the low effort. However, it also has two disadvantages compared to revealed effort: First, low effort leads to low deterrence of sophisticates. Second, hiding induces different gain thresholds for the indifferent sophisticated and naïve offender, respectively, and hence leads to a distortion in the sense that a given number of acts is not committed by the offenders with the highest gains. The regime comparison is then driven by the relative importance of these different effects. In particular, hiding becomes more attractive when the share of naïves is high.

In extensions, we consider several additional factors that affect the regime comparison just described. First, revealing the effort may reduce its effectiveness as offenders learn how to avoid detection. This makes hiding favorable, not only when the share of naïves is high, but also when it is low. Second, when the fine also becomes part of the enforcement policy, the authority may prefer to set a fine below the maximal level to mitigate the effect that inefficiently many naïve offenders are deterred. Third, the naïve offenders may differ with respect to their perceptions about the enforcement effort, which reinforces the issues that it might no longer the subjects with the highest benefits who commit the act. This works in favor of revelation. Overall, our results show that, when deciding on their effort and communication strategy, authorities should take into account the number of offenders with mis-perceptions and their degree of mis-perception. Those parameters, might well differ across different types of offenses (Kleck et al., 2005). We view our paper as contributing to the overall agenda of integrating the perspectives from law & economics and criminology (see e.g., Chalfin and McCrary, 2017) in the academic debate on law enforcement and deterrence.

Our framework could be extended in several directions. First, we assume that the authority maximizes social welfare, which neglects potential principal-agent issues between society and the law enforcement authority. In particular in the context of private law enforcement, the authority may have an incentive to signal its competency by focusing on the number of detected offenders instead of overall welfare.¹⁵ Second, it would be interesting to relax the assumption that the perceptions of naïve offenders are exogenous

 $^{^{15}}$ As argued by Buechel and Muehlheusser (2016), the number of detections might not be too informative about the underlying enforcement effort.

and static. Instead, they could adapt their beliefs upon receiving noisy signals on the actual enforcement level, for example based on experiences of their own or within their social network as in Sah (1991). Enforcement authorities would then face a dynamic optimization problem that has not yet been solved. Third, the naïveté of offenders may only refer to some but not to all enforcement technologies. As an example, consider Ben Gurion airport where all arriving vehicles must first pass through a preliminary security checkpoint where armed guards search the vehicle and exchange a few words with the driver and occupants to gauge their mood and intentions.¹⁶ As this effort is observable to everyone, our distinction between naïve and sophisticated offenders is likely to be of minor importance. In addition, however, plain clothes officers patrol the area outside the terminal building, assisted by hidden surveillance cameras which operate around the clock, and not all offenders might be aware of this effort. The general question is then how the authority should divide its effort between the directly observable and the not directly observable technology, and how the incentive to reveal information on the latter depends on the relative efficacy and costs of the two technologies. Finally, our model could be extended to include precaution on the side of the victims. Potential victims might invest into safety technologies like alarm equipment, but they may also have inaccurate perceptions on the benefits of those technologies.

¹⁶See e.g., https://edition.cnn.com/travel/article/ben-gurion-worlds-safest-airport-tel-aviv, lastly re-trieved on April, 28, 2018.

Appendix

A Proof of Lemma 1

As for regime H, suppose first that a = 1. Then the maximizer of surplus function (2) is $e_H^*(1) = 0$. Hence, $\pi p(e_H^*(a))f = \pi p(0)f = 0 < h$. Now, let a < 1. Then surplus function (2) is increasing at e = 0 because of the Inada condition C'(0) = 0. Hence, the maximizer satisfies $e_H^*(a) > 0$, i.e. the optimal effort is interior, and satisfies the first order condition Eq. (3). Our assumptions on the cost function C(e) ensure that the RHS of Eq. (3) is always strictly positive for all e > 0. Hence, the condition can only be satisfied when the LHS is also strictly positive. Since $G'(\cdot) > 0$, and p'(e) > 0 and f > 0, it follows that also $h - \pi p(e)f > 0$ must hold at $e = e_H^*(a)$ for the LHS to be strictly positive. This is, however, is just equivalent to the statement in the Lemma. The proof for regime R is completely analogous to the case a < 1 in regime H and hence omitted. \Box

B Proof of Proposition 1

(i) First suppose a < 1. Then surplus function (2) is increasing at e = 0 because of the Inada condition C'(0) = 0. Hence, the maximizer satisfies $e_H^*(a) > 0$, i.e. the optimal effort is interior, and satisfies the first order condition Eq. (3). That the optimal effort $e_H^*(a)$ is strictly decreasing in a can be established as follows: From Eq. (3), applying the implicit function theorem, one gets

$$\frac{\partial e_H^*(a)}{\partial a} = -\frac{(-1)\left[(h - \pi p(e)f) \cdot G'(p(e)f) \cdot p'(e)f\right]}{W_H''(e)} < 0.$$

To verify the sign of this expression, note first that the denominator is just the second derivative of the surplus function (2). At the optimal effort $e^*(a)$, this is strictly negative since this is the condition for a local optimum. Furthermore, the numerator is strictly negative since $G'(\cdot) > 0$, and p'(e) > 0 and f > 0 and by Lemma 1. Moreover, for a = 0, the surplus functions (2) and (4) coincide and so must the optimal enforcement levels. The property $e_H^*(a) < e_R^*$ for a > 0 then follows directly from the above arguments. Finally, for a = 1, the claim $e_H^*(1) = 0$ can be established by contradiction: Suppose, there are only naive offenders in the population (a = 1) and some e > 0 were optimal. Then social welfare would be strictly higher when e is reduced, since it would lead to lower cost (since C(e) is strictly increasing), but to no loss in deterrence. The reason is that as under regime H, the determine of naïves only works through \hat{e} , while the actual enforcement effort e has no impact.

(ii) Using the envelope theorem and taking the derivative of $W_H^*(a, \hat{e})$ w.r.t. \hat{e} yields

$$\frac{\partial W_H^*}{\partial \hat{e}} = -a(\pi p(\hat{e})f - h) \cdot G'(p(\hat{e})f) \cdot p'(\hat{e})f$$

which is strictly positive under Assumption 1.

(iii) and (iv): Using the envelope theorem and taking the derivative of $W_H^*(a, \hat{e})$ w.r.t. a yields

$$\frac{\partial W_H^*}{\partial a} = -\int_{p(e_H^*(a))f}^\infty (\pi g - h)G'(g)dg + \int_{p(\hat{e})f}^\infty (\pi g - h)G'(g)dg,$$

the sign of which is solely determined by comparing the two respective lower bounds of the integrals. By Lemma 1, we have $\pi p(e_H^*(a))f - h < 0$ and by Assumption 1 we have $\pi p(\hat{e})f - h < 0$ such that the first integral is bigger (in absolute terms) than the second one if and only if $p(e_H^*(a))f < p(\hat{e})f$. Hence, $W_H^*(a, \hat{e})$ is increasing in a if and only if $e_H^*(a) < \hat{e}$, and they are identical for $e_H^*(a) = \hat{e}$. Part (iii) supposes that $\hat{e} > e_H^*(0)$. Proposition 1 above shows that $e_H^*(a)$ is decreasing. Hence, we have $\hat{e} > e_H^*(a)$ for all $a \in [0,1]$. Thus, $W_H^*(a, \hat{e})$ is strictly monotone increasing in a. Part (iv) supposes that $0 < \hat{e} < e_H^*(0)$. Proposition 1 above shows that $e_H^*(a)$ is decreasing with $e_H^*(1) = 0$. Together, we have $e_H^*(0) > \hat{e} > e_H^*(1)$, and there must be a threshold $\hat{a}(\hat{e})$ such that $e_H^*(\hat{a}) = \hat{e}$. Hence, $W_H^*(a, \hat{e})$ is strictly decreasing in a for $a < \hat{a}(\hat{e})$ and strictly increasing when the inequality is reversed. \Box

C Proof of Proposition 2

Part (i): Recall first that the two regimes coincide for a = 0, i.e., when there are no naïve offenders $(W_H^*(0, \hat{e}) = W_R^*)$. As shown in Proposition 1, when $\hat{e} > e_H^*(0)$, $W_H^*(a, \hat{e})$ is strictly increasing in a for all $a \in [0, 1]$ and hence it is optimal for the enforcement authority to hide its enforcement effort.

Parts (ii) and (iii): When $0 < \hat{e} < e_H^*(0)$, then as shown in Proposition 1 above, $W_H^*(a, \hat{e})$ is U-shaped and strictly decreasing in the interval $[0, \hat{a})$ and increasing for $a > \hat{a}$. Hence, there must exist a (second) point of intersection between $W_H^*(a, \hat{e})$ and W_R^* at some point $\tilde{a}(\hat{e}) > 0$. The condition $W_H^*(1, \hat{e}) > W_R^*$ is a necessary and sufficient condition for $\tilde{a}(\hat{e})$ to lie in the relevant range (0, 1), which is also illustrated in Figure 1. When it is satisfied, as assumed in part (ii), then $W_H^*(a, \hat{e}) < (>)W_R^*$ for all $a < (>)\tilde{a}(\hat{e})$. When it is not satisfied, as assumed in part (iii), then $W_H^*(a, \hat{e}) < W_R^*$ for all $a \in (0, 1]$. For the special case $\hat{e} = 0$, $W_H^*(a, \hat{e})$ is strictly decreasing in [0, 1), starting at $W_H^*(0, \hat{e}) = W_R^*$. Hence, this case is treated in part (iii). \Box

D Proof of Lemma 2

First, due to the Inada condition C'(0) = 0, optimal effort is strictly positive for both surplus functions (4) and (6). Comparing the first order conditions Eq. (5) and Eq. (7) reveals that, since $\tilde{p}_R(e) < p(e)$ for all e > 0, the marginal benefit is pointwise smaller in Eq. (7) compared to Eq. (5). Hence, the point of intersection with the marginal cost curve C'(e) must occur at a smaller value of e.

Second, we now show that $W_R(e = \tilde{e}_R^*) > \widetilde{W}_R^*$ holds, and hence a fortiori, $W_R^* > \widetilde{W}_R^*$ must hold. Note that we have

$$\widetilde{W}_R^* = \int_{\widetilde{p}_R(\widetilde{e}_R^*)f}^{\infty} (\pi g - h) G'(g) dg - C(\widetilde{e}_R^*).$$
(11)

and evaluating W(e) from the surplus function (4) at $e = \tilde{e}_R^*$ yields

$$W_R(\tilde{e}_R^*) = \int_{p(\tilde{e}_R^*)f}^{\infty} (\pi g - h) G'(g) dg - C(\tilde{e}_R^*).$$
(12)

Taking the difference $W_R(\widetilde{e}_R^*) - \widetilde{W}_R^*$ yields

$$\int_{p(\tilde{e}_R^*)f}^{\infty} (\pi g - h)G'(g)dg - \int_{\widetilde{p}_R(\tilde{e}_R^*)f}^{\infty} (\pi g - h)G'(g)dg.$$
(13)

Note that since $\tilde{p}_R(e) < p(e)$ for all e > 0, the lower bound is strictly larger in the first integral. Hence, we can rewrite the difference as

$$\int_{p(\tilde{e}_R^*)f}^{\infty} (\pi g - h)G'(g)dg - \left(\int_{p(\tilde{e}_R^*)f}^{\infty} (\pi g - h)G'(g)dg + \int_{\widetilde{p}_R(\tilde{e}_R^*)f}^{p(\tilde{e}_R^*)f} (\pi g - h)G'(g)dg\right).$$
(14)

Since the first two terms cancel, this is equal to

$$-\int_{\widetilde{p}_R(\widetilde{e}_R^*)f}^{p(\widetilde{e}_R^*)f} (\pi g - h)G'(g)dg.$$
(15)

A sufficient condition for this expression to be strictly positive is that the value of the integrand at the upper bound $p(\tilde{e}_R^*)f$ is negative, i.e. if $\pi p(\tilde{e}_R^*)f - h < 0$. Note that we have established in Lemma 1 above that $\pi p(e_R^*)f - h < 0$ holds. Since we have shown at the beginning of this proof that $\tilde{e}_R^* < e_R^*$, a fortiori $\pi p(\tilde{e}_R^*)f - h < 0$ must hold as $p(\cdot)$ is an increasing function. This in turn implies that expression (15) is indeed strictly positive.

As a final step, since we have shown that $W_R(\tilde{e}_R^*) > \widetilde{W}_R^*$ holds, this must a fortiori be true for the maximum surplus under revealed enforcement effort in the basic model (W_R^*) , i.e. we have $W_R^* \ge W_R(\tilde{e}_R^*) > \widetilde{W}_R^*$. \Box

E Proof of Proposition 3

Part (i): From Proposition 1, when $\hat{e} > e_H^*(0)$, $W_H^*(a, \hat{e})$ is strictly increasing in a for all $a \in [0, 1]$. Moreover, as shown in Lemma 2, $W_R^* > \widetilde{W}_R^*$ holds, so that we have $W_H^*(a, \hat{e}) \ge W_R^* > \widetilde{W}_R^*$ for all $a \in [0, 1]$ and hence it is always optimal for the enforcement authority to hide its enforcement effort.

Part (ii): When $0 < \hat{e} < e_H^*(0)$, then from Proposition 1, $W_H^*(a, \hat{e})$ is U-shaped in a, and it takes its minimum value at $a = \hat{a}$. When this minimum value still exceeds \widetilde{W}_R^* (i.e. when $W_H^*(\hat{a}, \hat{e}) > \widetilde{W}_R^*$), then hiding the enforcement effort is again globally optimal. For the special case $\hat{e} = 0$, $W_H^*(a, \hat{e})$ is decreasing in a and $e_H^*(1) = \hat{e} = 0$, i.e. $\hat{a} = 1$. Thus, the statement also holds.

Part (iii): When $0 < \hat{e} < e_H^*(0)$ (so that $W_H^*(a, \hat{e})$ is U-shaped in a, but $W_H^*(\hat{a}, \hat{e}) < \widetilde{W}_R^*$), then there must exist a threshold $a_1 > 0$ such that $W_H^*(a, \hat{e}) > \widetilde{W}_R^*$ for all $a \in [0, a_1)$ (recall that $W_H^*(0, \hat{e}) = W_R^* > \widetilde{W}_R^*$). Moreover, if also the condition $W_H^*(1, \hat{e}) > \widetilde{W}_R^*$ is satisfied, then there must exist a second threshold a_2 with $a_1 < a_2 < 1$ such that hiding the effort is also optimal for all $a \in (a_2, 1]$, and revealing it is optimal in the intermediate range (a_1, a_2) .¹⁷

Part (iv): This part refers to the setting of part (iii), but where the condition $W_H^*(1, \hat{e}) > \widetilde{W}_R^*$ does not hold, i.e. we have $W_H^*(1, \hat{e}) < \widetilde{W}_R^*$. Then a range where hiding the enforcement effort is again optimal for sufficiently large a does not exist (i.e. $a_2 > 1$), so that hiding is optimal (only) for $a \in [0, a_1)$ and revealing is optimal for $a > a_1$. This also holds for the special case $\hat{e} = 0$ (given that $W_H^*(1, \hat{e}) < \widetilde{W}_R^*$). \Box

F Proof of Proposition 4

Part (i): In regime R, the authority chooses its enforcement effort e and the fine f to maximize welfare.

$$\max_{e \ge 0, f \in [0,F]} W_R(e,f) = \int_{p(e)f}^{\infty} (\pi g - h) G'(g) dg - C(e).$$
(16)

¹⁷This cannot occur in the special case $\hat{e} = 0$ since for $\hat{a} = 1$, we cannot have $W_H^*(\hat{a}, \hat{e}) < \widetilde{W}_R^*$ and $W_H^*(1, \hat{e}) > \widetilde{W}_R^*$ at the same time.

We first observe that due to the Inada condition C'(0) = 0, the optimal effort is interior, i.e. $e_R^* > 0$. We next show that there is no over-enforcement, i.e. $\pi p(e_R^*) f^* < h$. Suppose to the contrary that $\pi p(e_R^*) f^* \ge h$. Then a slight reduction of e_R^* , while keeping f^* constant, would weakly increase social benefits and strictly decrease the costs. Suppose now that the optimal fine f^* was not maximal, i.e. $f^* < F$. By continuity, the induced level of deterrence, $p(e_R^*) f^*$, can also be reached by a lower effort $e' < e_R^*$ and a higher fine $f' > f^*$. This increases welfare since the increase in fine is costless, while the decrease in effort saves costs. Hence, the optimal fine must be maximal: $f^* = F$. Consequently, the optimal effort equals the optimal effort of the baseline model when we set fine \bar{f} (of the baseline model) to F.

Part (ii): In regime H, the authority chooses its enforcement effort e and the fine f to maximize welfare.

$$\max_{e \ge 0, f \in [0,F]} W_H(e,f) = (1-a) \cdot \left[\int_{p(e)f}^{\infty} (\pi g - h) G'(g) dg \right] + a \cdot \left[\int_{p(\hat{e})f}^{\infty} (\pi g - h) G'(g) dg \right] - C(e).$$
(17)

Observe first that f = 0 is never optimal. (Indeed, for every effort e and perceived effort \hat{e} , there is a (small) fine $f^{\epsilon} > 0$ with $\pi p(e)f^{\epsilon} < h$ and $\pi p(\hat{e})f^{\epsilon} < h$ such that $W_H(e,0) < W_H(e,f^{\epsilon})$). Hence the optimal fine f^* is either interior or maximal.

By assumption of part (ii) of Proposition 4, the optimal fine is maximal: $f^* = F$. In this case, the solution for e_H^* is implicitly given by the first-order condition

$$(1-a) [(h - \pi p(e)F) \cdot G'(p(e)F) \cdot p'(e)F] = C'(e).$$

Observe that this condition coincides with Eq. (3) that determines the optimal effort in regime H of the baseline model, when we set $\bar{f} \equiv F$.

Part (iii): By assumption of part (iii) of Proposition 4, the optimal fine is interior, i.e. $f^* < F$. In this case, the maximization problem (17) has a solution (e^*, f^*) that satisfies the following two first order conditions:

$$(1-a)\left[(h-\pi p(e)f) \cdot G'(p(e)f) \cdot p'(e)f\right] = C'(e),$$
(18)

$$(1-a)\left[(h-\pi p(e)f) \cdot G'(p(e)f) \cdot p(e)\right] = a\left[(\pi p(\hat{e})f - h) \cdot G'(p(\hat{e})f) \cdot p(\hat{e})\right].$$
(19)

Both equations (18) and (19) follow from Leibniz's rule. The RHS of Eq. (18), C'(e), is positive for $e = e_H^* > 0$. Hence its LHS is also positive. The LHS of Eq. (18) is only positive if $\pi p(e)f < h$. Now, observe that $\pi p(e)f < h$ implies that the LHS of Eq. (19) is also positive. In turn, the RHS of Eq. (19) must also be positive, which implies that $(\pi p(\hat{e})f - h) > 0$. Hence, for the optimal effort e^* and the optimal fine f^* , the two first order conditions imply $\pi p(e^*)f^* < h < \pi p(\hat{e})f^*$. And finally, $e^* < \hat{e}$.

Part (iv): For the baseline model, let us denote by e_R^b and e_H^b the respective optimal efforts in regime R and in regime H. By assumption, $W_R(e_R^b, \bar{f}) < W_H(e_H^b, \bar{f})$ for the fixed fine \bar{f} . By part (i) of this proposition, the optimal fine in regime R is maximal, i.e. equal to F. Together with $F = \bar{f}$, this yields $\max_{e>0, f \in (0, F=\bar{f}]} W_R(e, f) = W_R(e_R^b, \bar{f})$. Hence,

$$\max_{e>0, f\in(0,\bar{f}]} W_R(e,f) = W_R(e_R^b,\bar{f}) < W_H(e_H^b,\bar{f}) \le \max_{e>0, f\in(0,\bar{f}]} W_H(e,f). \quad \Box$$

G Proof of Proposition 5

Welfare $W_R(e)$ under regime R is unaffected from heterogeneity of perceptions and hence neither is optimal welfare $W_R(e_R^*) = W_R^*$ in this case. Welfare under regime H is given by Eq. (2) in the baseline model and by Eq. (10) in the heterogeneity extension. Observe that the optimal effort in both Eq. (2) and Eq. (10) then only depends on the first and on the last term, which coincide in both equations for $1 - a = 1 - \sum_{l=1}^{L} a_l$. Thus, optimal effort e_H^* is the same in both scenarios. For $a = \sum_{l=1}^{L} a_l$, the difference between the two becomes

$$(10) - (2) = \sum_{l=1}^{L} a_l \cdot \left[\int_{p(\hat{e}_l)f}^{\infty} (\pi g - h) G'(g) dg \right] - a \cdot \left[\int_{p(\hat{e})f}^{\infty} (\pi g - h) G'(g) dg \right].$$

Observe that this difference is independent of the actual effort e. By assumption on the exogenous perceptions \hat{e} and \hat{e}_l , both expressions in brackets are negative (cf. Assumption 1). For $\hat{e} \equiv 0 (< \min\{\hat{e}_1, ..., \hat{e}_L\})$, the left term is larger in absolute terms than the right one such that the difference is positive. For $\hat{e} \equiv e^{max}(> \max\{\hat{e}_1, ..., \hat{e}_L\})$, the difference is negative. Since the difference is a continuously decreasing function in \hat{e} , there must exist a unique level $\hat{e} \equiv \tilde{e}$ that satisfies that the difference is just zero.

We have thus constructed a model with homogeneous perceptions \tilde{e} that is welfare equivalent to the given model with heterogeneous perceptions. \Box

H Proof of Proposition 6

Welfare under regime H is given by the surplus function (10), which for two groups becomes

$$W_H(e) = (1 - a_1 - a_2) \cdot \left[\int_{p(e)f}^{\infty} (\pi g - h) G'(g) dg \right] + \sum_{l=1}^{2} a_l \cdot \left[\int_{p(\hat{e}_l)f}^{\infty} (\pi g - h) G'(g) dg \right] - C(e).$$

Setting $\hat{e}_1 = \hat{e} - \sigma$ and $\hat{e}_2 = \hat{e} + \sigma$ and applying Leibniz's rule yields $\frac{\partial W_H(e)}{\partial \sigma} =$

$$a_{1} \cdot \left[-[\pi p(\hat{e} - \sigma)f - h]G'(p(\hat{e} - \sigma)f) \frac{\partial p(\hat{e} - \sigma)f}{\partial \sigma} \right] + a_{2} \cdot \left[-[\pi p(\hat{e} + \sigma)f - h]G'(p(\hat{e} + \sigma)f) \frac{\partial p(\hat{e} + \sigma)f}{\partial \sigma} \right] + a_{2} \cdot \left[-[\pi p(\hat{e} - \sigma)f - h]G'(p(\hat{e} - \sigma)f) \frac{\partial p(\hat{e} - \sigma)f}{\partial \sigma} \right] + a_{2} \cdot \left[-[\pi p(\hat{e} - \sigma)f - h]G'(p(\hat{e} - \sigma)f) \frac{\partial p(\hat{e} - \sigma)f}{\partial \sigma} \right] + a_{2} \cdot \left[-[\pi p(\hat{e} - \sigma)f - h]G'(p(\hat{e} - \sigma)f) \frac{\partial p(\hat{e} - \sigma)f}{\partial \sigma} \right] + a_{2} \cdot \left[-[\pi p(\hat{e} - \sigma)f - h]G'(p(\hat{e} - \sigma)f) \frac{\partial p(\hat{e} - \sigma)f}{\partial \sigma} \right] + a_{2} \cdot \left[-[\pi p(\hat{e} - \sigma)f - h]G'(p(\hat{e} - \sigma)f) \frac{\partial p(\hat{e} - \sigma)f}{\partial \sigma} \right] + a_{2} \cdot \left[-[\pi p(\hat{e} - \sigma)f - h]G'(p(\hat{e} - \sigma)f) \frac{\partial p(\hat{e} - \sigma)f}{\partial \sigma} \right] + a_{2} \cdot \left[-[\pi p(\hat{e} - \sigma)f - h]G'(p(\hat{e} - \sigma)f) \frac{\partial p(\hat{e} - \sigma)f}{\partial \sigma} \right] + a_{2} \cdot \left[-[\pi p(\hat{e} - \sigma)f - h]G'(p(\hat{e} - \sigma)f) \frac{\partial p(\hat{e} - \sigma)f}{\partial \sigma} \right] + a_{2} \cdot \left[-[\pi p(\hat{e} - \sigma)f - h]G'(p(\hat{e} - \sigma)f) \frac{\partial p(\hat{e} - \sigma)f}{\partial \sigma} \right] + a_{2} \cdot \left[-[\pi p(\hat{e} - \sigma)f - h]G'(p(\hat{e} - \sigma)f) \frac{\partial p(\hat{e} - \sigma)f}{\partial \sigma} \right] + a_{2} \cdot \left[-[\pi p(\hat{e} - \sigma)f - h]G'(p(\hat{e} - \sigma)f) \frac{\partial p(\hat{e} - \sigma)f}{\partial \sigma} \right] + a_{2} \cdot \left[-[\pi p(\hat{e} - \sigma)f - h]G'(p(\hat{e} - \sigma)f) \frac{\partial p(\hat{e} - \sigma)f}{\partial \sigma} \right] + a_{2} \cdot \left[-[\pi p(\hat{e} - \sigma)f - h]G'(p(\hat{e} - \sigma)f) \frac{\partial p(\hat{e} - \sigma)f}{\partial \sigma} \right] + a_{2} \cdot \left[-[\pi p(\hat{e} - \sigma)f - h]G'(p(\hat{e} - \sigma)f) \frac{\partial p(\hat{e} - \sigma)f}{\partial \sigma} \right] + a_{2} \cdot \left[-[\pi p(\hat{e} - \sigma)f - h]G'(p(\hat{e} - \sigma)f) \frac{\partial p(\hat{e} - \sigma)f}{\partial \sigma} \right] + a_{2} \cdot \left[-[\pi p(\hat{e} - \sigma)f - h]G'(p(\hat{e} - \sigma)f) \frac{\partial p(\hat{e} - \sigma)f}{\partial \sigma} \right] + a_{2} \cdot \left[-[\pi p(\hat{e} - \sigma)f - h]G'(p(\hat{e} - \sigma)f) \frac{\partial p(\hat{e} - \sigma)f}{\partial \sigma} \right] + a_{2} \cdot \left[-[\pi p(\hat{e} - \sigma)f - h]G'(p(\hat{e} - \sigma)f) \frac{\partial p(\hat{e} - \sigma)f}{\partial \sigma} \right] + a_{2} \cdot \left[-[\pi p(\hat{e} - \sigma)f - h]G'(p(\hat{e} - \sigma)f) \frac{\partial p(\hat{e} - \sigma)f}{\partial \sigma} \right] + a_{2} \cdot \left[-[\pi p(\hat{e} - \sigma)f - h]G'(p(\hat{e} - \sigma)f) \frac{\partial p(\hat{e} - \sigma)f}{\partial \sigma} \right] + a_{2} \cdot \left[-[\pi p(\hat{e} - \sigma)f - h]G'(p(\hat{e} - \sigma)f) \frac{\partial p(\hat{e} - \sigma)f}{\partial \sigma} \right] + a_{2} \cdot \left[-[\pi p(\hat{e} - \sigma)f - h]G'(p(\hat{e} - \sigma)f) \frac{\partial p(\hat{e} - \sigma)f}{\partial \sigma} \right] + a_{2} \cdot \left[-[\pi p(\hat{e} - \sigma)f - h]G'(p(\hat{e} - \sigma)f) \frac{\partial p(\hat{e} - \sigma)f}{\partial \sigma} \right] + a_{2} \cdot \left[-[\pi p(\hat{e} - \sigma)f - h]G'(p(\hat{e} - \sigma)f) \frac{\partial p(\hat{e} - \sigma)f}{\partial \sigma} \right] + a_{2} \cdot \left[-[\pi p(\hat{e} - \sigma)f - h]G'(p(\hat{e}$$

Using $a_1 = a_2$, we get $\frac{\partial W_H(e)}{\partial \sigma} < 0$ if and only if

$$[h - \pi p(\hat{e} - \sigma)f]G'(p(\hat{e} - \sigma)f)f\frac{\partial p(\hat{e} - \sigma)}{\partial \sigma} + [h - \pi p(\hat{e} + \sigma)f]G'(p(\hat{e} + \sigma)f)f\frac{\partial p(\hat{e} + \sigma)}{\partial \sigma} < 0.$$

which is (by rearranging such that every factor is positive)

$$[h - \pi p(\hat{e} + \sigma)f]G'(p(\hat{e} + \sigma)f) \cdot \frac{\partial p(\hat{e} + \sigma)}{\partial \sigma} < [(h - \pi p(\hat{e} - \sigma)f]G'(p(\hat{e} - \sigma)f) \cdot (-\frac{\partial p(\hat{e} - \sigma)}{\partial \sigma}) + (-\frac{\partial p(\hat{e} - \sigma)}{\partial \sigma})]G'(p(\hat{e} - \sigma)f) \cdot (-\frac{\partial p(\hat{e} - \sigma)}{\partial \sigma}) = (-\frac{\partial p(\hat{e} - \sigma)}{\partial \sigma}) + (-\frac{\partial p(\hat{e} - \sigma)}{\partial \sigma}) = (-\frac{\partial p$$

The last inequality is equivalent to the condition stated in the proposition that G and p(e) are "not too convex." Hence, this condition implies that $\frac{\partial W_H(e)}{\partial \sigma} < 0$ for any e. Since the optimal effort e_H^* is independent of σ , also the maximum welfare $W_H^* = W_H(e_H^*)$ is decreasing in σ . \Box

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