

# Negligence, Strict Liability, and Complexity

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# Introduction

- Strict liability versus negligence.
- Equivalence result in the basic accident model of Shavell.
  - ▶ negligence "puzzle"
- Literature explores many departures from the basic model.
- I and several others have explored ambiguity.
- David, Surajeet, and I are exploring unawareness.
- In this paper I will explore **complexity**.

# The Model

- Unilateral care model with fixed activity level.
- Two risk neutral agents: injurer and victim.
- Injurer engages in a risky activity.
- In the event of an accident the victim incurs a loss.
- Only the injurer can take precautions against an accident.

# The Model

## Standard Continuous/Divisible Care Model

- Continuous set of precautions,  $\mathbb{R}_+$ .
- Injurer chooses level of care  $x \in \mathbb{R}_+$  having cost  $c(x)$ .
  - ▶  $c : \mathbb{R}_+ \rightarrow \mathbb{R}_+$  is differentiable, increasing, and either linear or convex.
- Expected accident loss is  $l(x)$ .
  - ▶  $l : \mathbb{R}_+ \rightarrow \mathbb{R}_+$  is differentiable, decreasing, and convex.

# The Model

## Standard Continuous/Divisible Care Model

- The **social goal** is to

$$\underset{x \in \mathbb{R}_+}{\text{minimize}} \quad c(x) + I(x).$$

The solution  $x^*$  is given implicitly by  $c'(x^*) = -I'(x^*)$ .

- Under **strict liability**, the injurer's problem is

$$\underset{x \in \mathbb{R}_+}{\text{minimize}} \quad c(x) + I(x),$$

which is the social goal. Hence, the injurer takes socially optimal care  $x^*$ .

- Under **negligence**, the injurer's problem is

$$\underset{x \in \mathbb{R}_+}{\text{minimize}} \quad \begin{cases} c(x) & \text{if } x \geq \bar{x} \\ c(x) + I(x) & \text{otherwise} \end{cases},$$

where  $\bar{x}$  is the due care standard set by the court. If  $\bar{x} = x^*$ , then the injurer takes socially optimal care.

# The Model

## Discrete/Indivisible Care Model

- Discrete set of precautions,  $N$ .
  - ▶  $|N| = n$ .
- Injurer chooses subset  $X \subseteq N$  having cost  $c(X)$ .
  - ▶  $c : 2^N \rightarrow \mathbb{R}_+$  is monotone increasing.
  - ▶ monotone increasing:  $c(A) \leq c(B)$  for all  $A \subseteq B \subseteq N$ .
- Assume  $c$  is either modular or supermodular.
  - ▶ modular:  $c(A) = \sum_{i \in A} c_i$  for all  $A \subseteq N$ .
  - ▶ supermodular:  $c(B \cup \{i\}) - c(B) \geq c(A \cup \{i\}) - c(A)$  for all  $A \subseteq B \subseteq N$  and  $i \in N \setminus B$ .
- Expected accident loss is  $l(X)$ .
  - ▶  $l : 2^N \rightarrow \mathbb{R}_+$  is monotone decreasing.
- Assume initially that  $l$  is submodular.
  - ▶ submodular:  $l(B \cup \{i\}) - l(B) \leq l(A \cup \{i\}) - l(A)$  for all  $A \subseteq B \subseteq N$  and  $i \in N \setminus B$ .

# The Model

## Discrete/Indivisible Care Model

- The **social goal** is to

$$\underset{X \subseteq N}{\text{minimize}} \quad c(X) + I(X). \quad (\text{SOCIAL GOAL})$$

Let  $X^*$  denote the solution to SOCIAL GOAL.

- Under **strict liability**, the injurer's problem is

$$\underset{X \subseteq N}{\text{minimize}} \quad c(X) + I(X). \quad (\text{STRICT LIABILITY})$$

which is identical to SOCIAL GOAL.

- Under **negligence**, the injurer's problem is

$$\underset{X \subseteq N}{\text{minimize}} \quad \begin{cases} c(X) & \text{if } X \supseteq \bar{X} \\ c(X) + I(X) & \text{otherwise} \end{cases}. \quad (\text{NEGLIGENCE})$$

where  $\bar{X} \subseteq N$  is the due care standard set by the court.

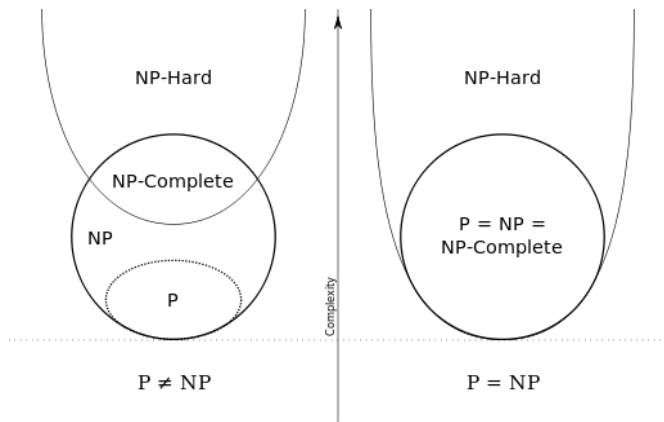
# Complexity

Four important computational complexity classes:

- $\mathcal{P}$  = problems that can be solved computationally in polynomial time.
- $\mathcal{NP}$  = problems for which a given solution can be verified in polynomial time.
  - ▶  $\mathcal{P} = \mathcal{NP}$ ? is a Millennium Prize Problem. It is conjectured that  $\mathcal{P} \neq \mathcal{NP}$ .
- $\mathcal{NP}$ -hard = problems with no known polynomial-time solutions.
- $\mathcal{NP}$ -complete = problems in  $\mathcal{NP}$  that are  $\mathcal{NP}$ -hard.



# Complexity



## Theorem

*If  $c$  is modular, then STRICT LIABILITY and NEGLIGENCE are computable in polynomial time.*

- Idea of proof:
- If  $c$  is modular then  $f(X) = c(X) + l(X)$  is submodular.
- Observe that  $f : 2^N \rightarrow \mathbb{R}_+ \Leftrightarrow f : \{0, 1\}^n \rightarrow \mathbb{R}_+$ .
- Construct Lovász extension  $f^L : [0, 1]^n \rightarrow \mathbb{R}_+$ .
- Invoke Lovász theorem:  $f^L$  is convex and  $\min_{X \subseteq N} f(X) = \min_{x \in [0, 1]^n} f(x)$ .
- Construct  $X^*$  from  $x^*$ .

## Theorem

*If  $c$  is supermodular, then STRICT LIABILITY and NEGLIGENCE are  $\mathcal{NP}$ -hard.*

- Proof is by reduction from MAX CUT, a well-known  $\mathcal{NP}$ -hard problem.
- MAX CUT is the problem of finding a subset of the vertex set of a graph that maximizes the number of edges between the subset and the complementary subset.

# Results

- Many other reasonable variations result in computational intractability.
- For instance:
  - ▶  $f$  is supermodular (reduction from MAX CUT).
  - ▶ cardinality constraint:  $|X| \geq k$  (reduction from MINKP).
  - ▶ "knapsack" constraint:  $c(X) \geq m$  (reduction from MINKP).

# Results

- Upshot is that neither strict liability nor negligence can achieve socially optimal deterrence, if the precaution set is too large (i.e., if  $n$  is too large).
- For strict liability to achieve socially optimal deterrence, the injurer must solve STRICT LIABILITY.
- For negligence to achieve socially optimal deterrence:
  - ▶ The court must set  $\bar{X} = X^*$ , which requires solving STRICT LIABILITY.
  - ▶ The injurer must solve NEGLIGENCE.
- But in many cases neither the injurer nor the court, even if armed with a non-deterministic Turing machine (i.e., a computer that can perform an unbounded number of parallel computations), can solve STRICT LIABILITY or NEGLIGENCE.

# Results

## Theorem

*If  $c$  and  $l$  are computable in polynomial time, then the decision problem forms of STRICT LIABILITY and NEGLIGENCE are  $\mathcal{NP}$ -complete.*

- Idea of proof:
- Transform each problem into a decision problem (i.e., a yes-no problem).
- For example, STRICT LIABILITY becomes: Given a threshold value  $t$ ,

$$\begin{array}{l} \text{there exists } X \subseteq N \\ \text{such that } c(X) + l(X) \leq t. \end{array}$$

- Show that, given any threshold value  $t$  and any proposed solution  $\hat{X} \subseteq N$ , a polynomial-time algorithm exists that verifies whether  $\hat{X}$  is a solution.
- Algorithm is trivial: It simply checks whether  $c(\hat{X}) + l(\hat{X}) \leq t$ . If  $c$  and  $p$  are computable in polynomial time, then the algorithm runs in polynomial time.

# Results

- Upshot is that negligence can move the law in the direction of optimal deterrence far more rapidly than strict liability.
- Imagine that the circumstances of an accident suggest to the parties a subset of precautions  $\hat{X}$  that would have prevented the accident.
- Suppose we have a negligence regime where the due care standard is  $\bar{X}$ .
  - ▶ For instance, suppose the court can compute  $\bar{X}$  using a polynomial-time approximation algorithm for STRICT LIABILITY.
- Because STRICT LIABILITY is  $\mathcal{NP}$ -complete, the court can check whether  $c(\hat{X}) + l(\hat{X}) \leq c(\bar{X}) + l(\bar{X})$ .
- If yes, the court resets the due care standard to  $\hat{X}$ . **This leads all injurers to move in the direction of optimal care.**
- If we have a strict liability regime, the injurer in question learns that  $\hat{X}$  is superior to  $\bar{X}$ . **But this knowledge is not disseminated to all injurers.**

# Results

- In some cases, it may be possible to approximate solutions to STRICT LIABILITY and NEGLIGENCE in polynomial time.
  - ▶ No known approximation for the cases where  $c$  or  $l$  is supermodular.
- But the approximations often have poor performance guarantees.
  - ▶ For the case of a cardinality constraint, it appears there may be a  $\sqrt{n}$ -approximation.
  - ▶ It is unclear to me at this point if there is an approximation for the case of a knapsack constraint.



# Conclusion

- Paper is a work in progress!
- It will develop the ideas outlined above.
- I welcome your comments, ideas, and suggestions!