

Individually Fair Learning with One-Sided Feedback

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Individual Fairness

“Similar individuals should be treated similarly.”

Meaningful guarantee at the individual level.

Problem: Metric often **unavailable**.

Auditor-based Approach

“Can you spot a pair of **similar** individuals who were treated **very differently**?”

“Yes. Individuals #5 and #17.”



Auditor “knows unfairness when he sees it.” Auditor

Issue #1: single auditors are prone to **biases**.

- Decision-makers less likely to entrust a single auditor with fairness-related judgements in high-stakes scenarios.
- How to reconcile cases disagreed upon by different auditors?

Auditing by Panels

- Fairness violation – only when a consensus is reached within a panel.
- Possible to alter the required fraction to **algorithmically** explore the fairness-accuracy frontier.

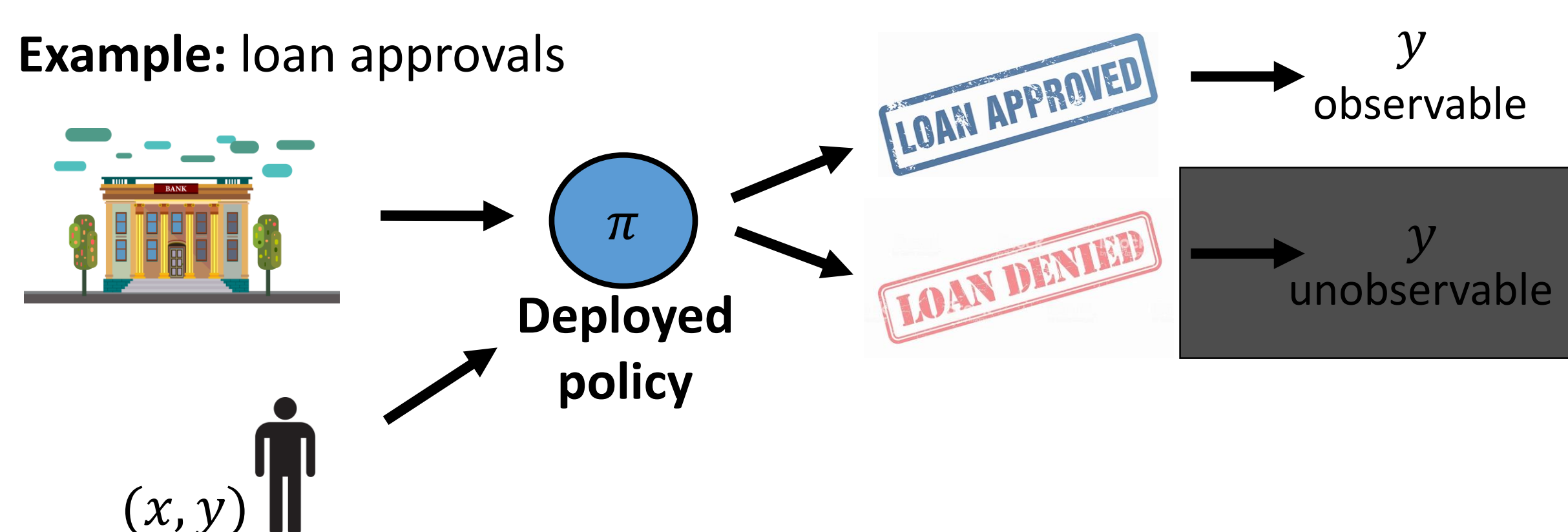


One-Sided Feedback

Issue #2: real-life feedback is often **one-sided**.

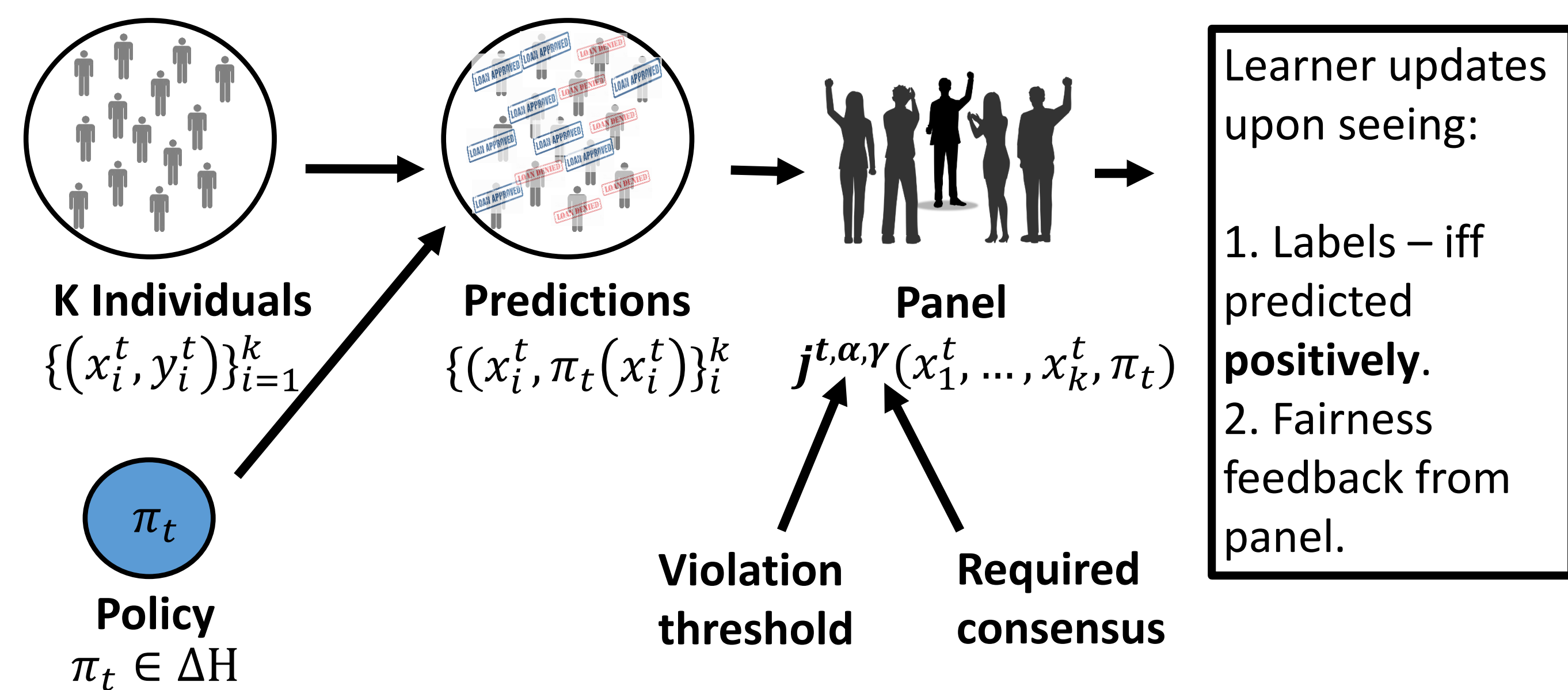
- “Hidden outcomes” of rejected individuals.
- Uncareful treatment may result in feedback loops.

Example: loan approvals



Our Setting

Online Learning with One-Sided Feedback + Feedback from Dynamically-Chosen Panels



Results

Result #1: Reduction from online learning with **one-sided feedback** and feedback from **dynamically-chosen panels** to Contextual Combinatorial Semi-Bandit.

Result #2: Multi-Criteria No-Regret Guarantees
Using regret bound of any algorithm for Contextual Combinatorial Semi-Bandit, upper bounding, simultaneously:

1. **Accuracy:** sub-linear regret vs. best fair policy.
2. **Fairness:** sub-linear number of rounds on which fairness violations exist.

Accuracy + Fairness Guarantees

Thm. 1 (simplified.): Using Exp2 algorithm,

$$\text{Accuracy: } \text{Regret}(\text{Exp2}, T, Q_{\alpha-\epsilon}) \leq O(k^{\frac{3}{2}} T^{\frac{4}{5}} \log |H|^{\frac{1}{2}})$$

$$\text{Fairness: } \sum_{t=1}^T \text{Unfair}^{\alpha, \gamma}(\pi_t, \bar{x}^t, \bar{j}^t) \leq O\left(\frac{1}{\epsilon} k^{\frac{3}{2}} T^{\frac{4}{5}} \log |H|^{\frac{1}{2}}\right)$$

Thm. 2 (simplified.): Using (adapted) Context-Semi-Bandit-FTPL,

$$\text{Accuracy: } \text{Regret}(\text{CSB-FTPL-WR}, T, Q_{\alpha-\epsilon}) \leq \tilde{O}\left(k^{\frac{11}{4}} s^{\frac{3}{4}} T^{\frac{41}{45}} \log |H|^{\frac{1}{2}}\right)$$

$$\text{Fairness: } \sum_{t=1}^T \text{Unfair}^{\alpha, \gamma}(\pi_t, \bar{x}^t, \bar{j}^t) \leq \tilde{O}\left(\frac{1}{\epsilon} k^{\frac{11}{4}} s^{\frac{3}{4}} T^{\frac{41}{45}} \log |H|^{\frac{1}{2}}\right)$$