

## Evaluating Educational Programs Impact on Texas students

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### Problem Statement

#### 1 Problem

TEA is the state agency in charge of public education in Texas. TEA designs programs to boost student performance and aims to measure its impact. However, Texas is diverse and causal inference is a challenge.

#### 2 Objective

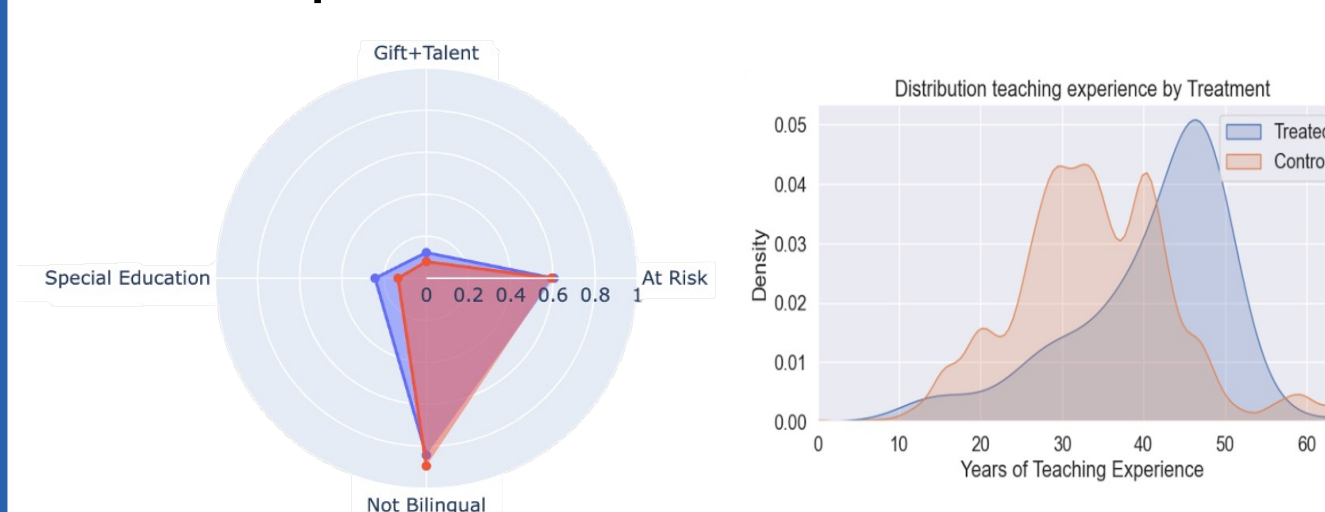
The objective of this project is to **measure the causal impact of TEA's initiatives on student success** by **optimally matching** classes that taught a new curriculum to control groups that didn't.

#### 3 Data & EDA

We have 4 data sources

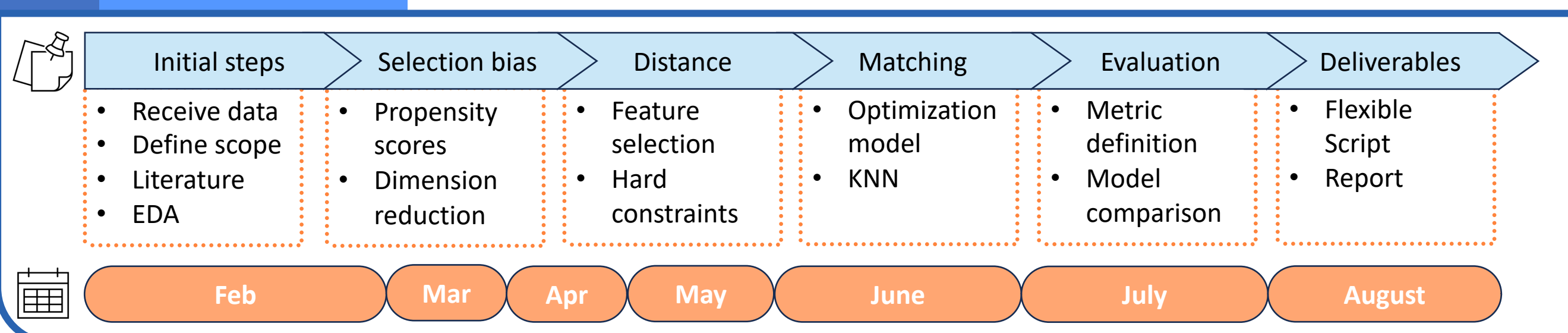
- Student demographics
- District data
- Teacher demographics
- Survey data

We compared treatment to control units



The units overlap, but there is selection bias

#### 4 Timeline

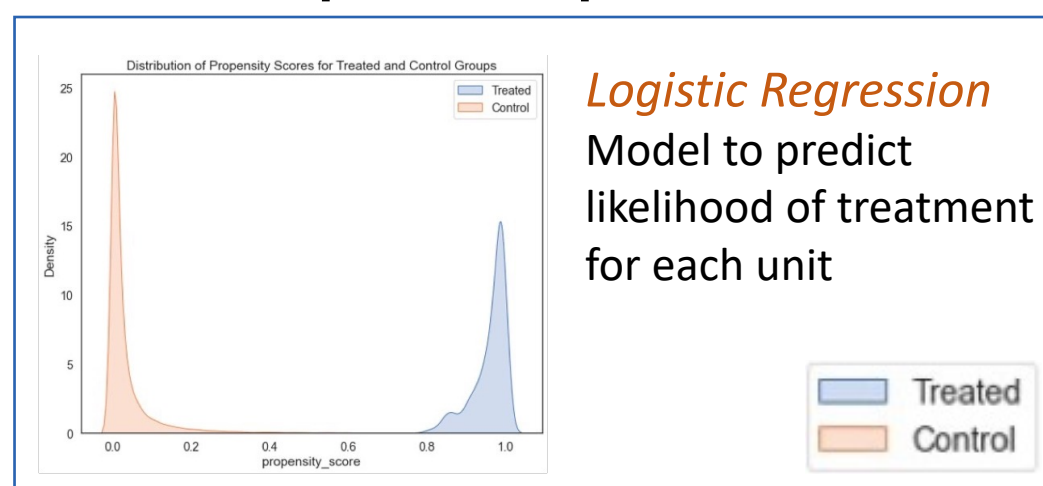


### Methodology

#### 5 Propensity Scores (PS)

Propensity scores (PS) account for selection bias in observational studies

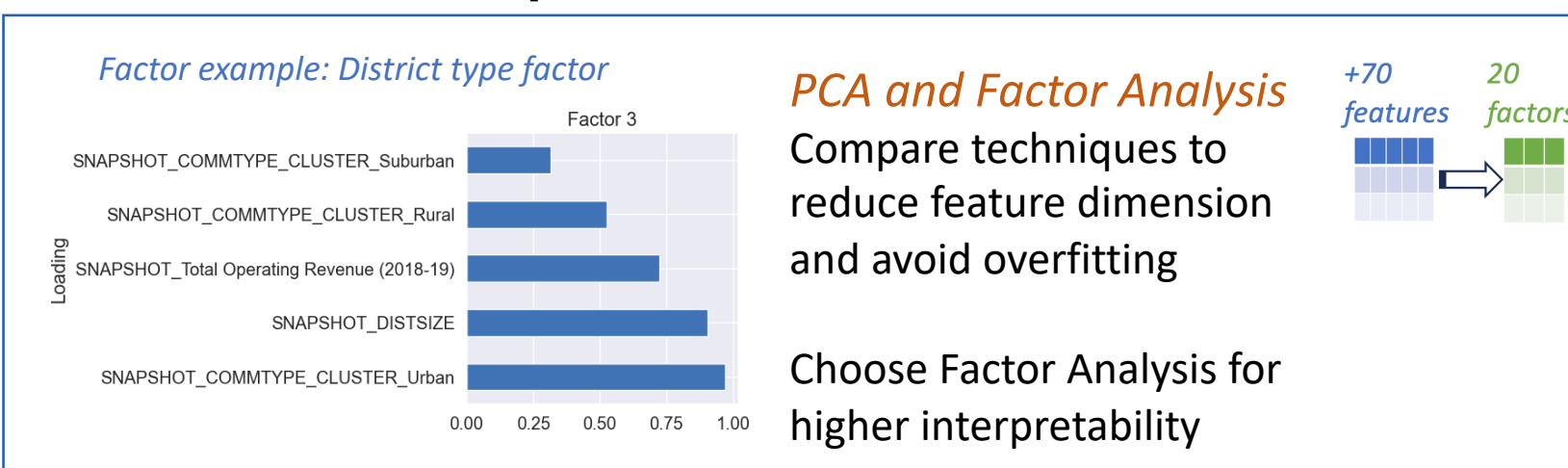
##### Step 1: Compute PS



**Logistic Regression Model** to predict likelihood of treatment for each unit

⚠️ **Overfitting** +70 features for 200 points

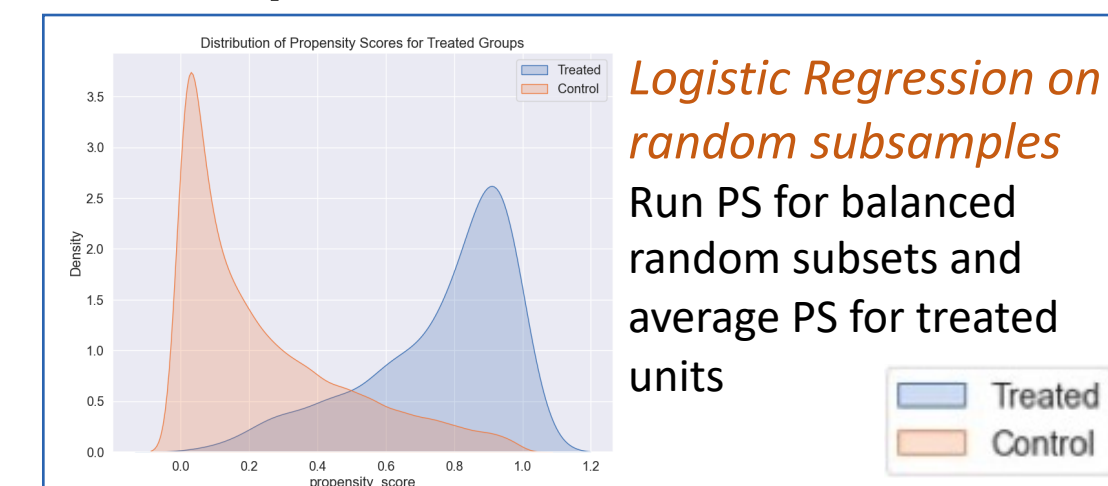
##### Step 2: Dimension Reduction



**Factor example: District type factor**  
**PCA and Factor Analysis** Compare techniques to reduce feature dimension and avoid overfitting  
 Choose Factor Analysis for higher interpretability

⚠️ **Class imbalance** 200 treatment vs +35K control

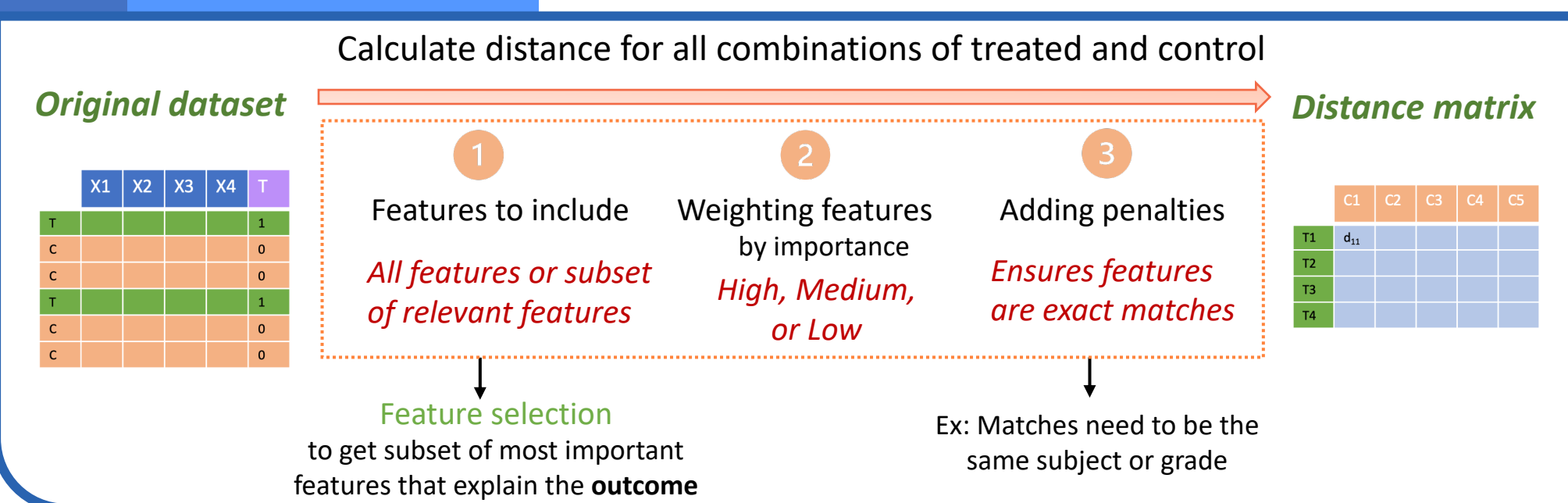
##### Step 3: PS on random subsets



**Logistic Regression on random subsamples**  
 Run PS for balanced random subsets and average PS for treated units

✅ **Balanced scores**

#### 6 Distance



#### 7 Matching

Wide range of matching options: 1:1 or 1:k, with or w/o replace, PS, Distance or PS Calipers, Optimal or greedy

Optimization model	
Objective function	Minimize total distance
Decision variable	Binary {1 if pair is selected, 0 otherwise}
Constraints	<b>C1:</b> Each treatment is matched with at least k control units <b>C2:</b> Each control is matched at most 1

Compare different algorithms, evaluate trade-offs, and recommend TEA which method to use while ensuring flexibility for future projects

### Results and Impact

#### 8 Results

##### Balance

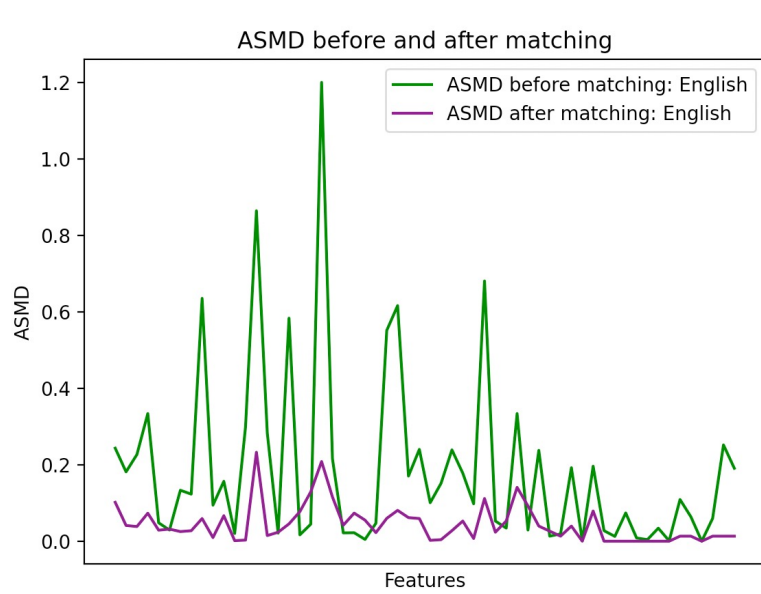
Our matching results have proven to significantly reduce bias and increase balance across features, which allows TEA to confidently evaluate programs and communicate results to stakeholders.

##### Treatment effect

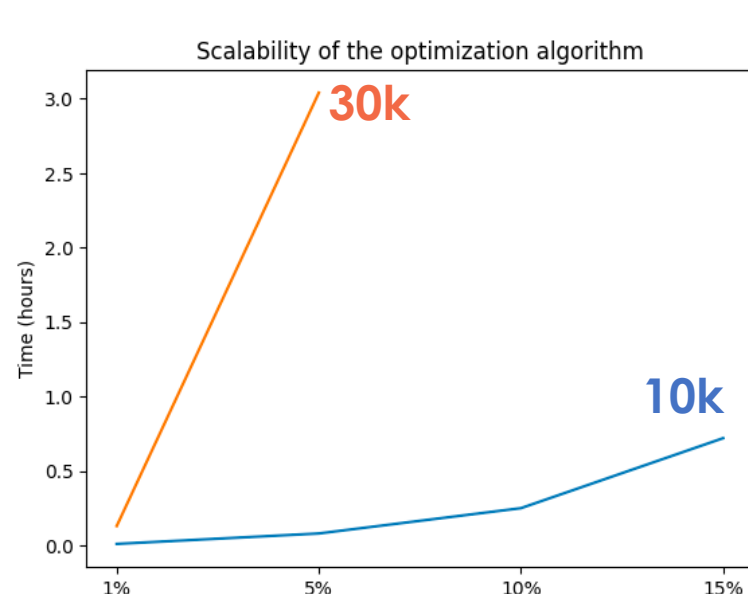
Our tool performs a difference-in-difference analysis, where applicable, to measure the treatment effect. This automation will allow TEA to easily gauge the effect and significance of their initiatives.

##### Scalability

We have a scalable optimization model and a greedy model that can perform with varying data set sizes and different ratios between control and treatment.



*"Continuous improvement is a core value at TEA. By rigorously evaluating our programs, we can encourage the field to adopt what works and improve what doesn't, to improve outcomes for all students in Texas."*



● **89% Bias Reduction** from unmatched data

● **80% Bias Reduction** from last year's baseline matching

● **100+** Different ways our script can be used to match

#### 9 Deliverables

We presented these deliverables for TEA to summarize our project

- Flexible Script**: Flexible script to perform the matching and evaluation
- Documentation**: Explanation on code, installation, and optimization basics
- Project overview**: Project overview slides targeted to different TEA audiences

#### 10 Next Steps

The next steps for our project are the following ones

- Implementation**: Implementation on other use cases
- Selection**: Choosing a matching method to evaluate 2023 CRIMS results
- Communication**: Communicate project with stakeholders