

## **DSCI 510 Final Project Progress Report**

**Student:** Jasmine Adams

**Project:** *E-Rate Market Opportunity Analysis: Identifying Underserved Schools and Libraries Using Multi-Source Data Integration*

**Date:** November 13, 2025

### Project Scope Update

The project scope remains consistent with the proposal. I am developing a data-driven system to identify underserved schools and libraries for E-Rate consulting opportunities by integrating multiple public datasets. Key objectives include:

1. Predicting E-Rate application approval probability using a classification model
2. Performing geographic clustering (K-means) to locate underserved regions
3. Building a market opportunity scoring system

The project structure is modular, separating data collection, preprocessing, modeling, and visualization, to support both academic goals and future business use.

### Data Sources

#### **1. USAC E-Rate Open Data API**

- **Endpoint:** <https://opendata.usac.org/resource/avi8-svp9.json>
- **Status:** Fully implemented via Python
- **Details:** Provides E-Rate funding data (entity name, funding year, discount rate, service type, etc.).
- **Implementation:** Created an ERateDataCollector class to automate pagination, error handling, and data export.
- **Progress:** Successfully fetched and validated California FY2024 data.

#### **2. NCES Common Core of Data (Urban Institute API)**

- **Status:** Next phase
- **Goal:** Collect school demographic and geographic data for model enrichment.

#### **3. IMLS Public Libraries Survey**

- **Status:** Next phase
- **Goal:** Add library operational data for library-focused opportunity analysis.

## Issues / Difficulties

### **Resolved:**

- **API rate limits:** Added batch fetching, pagination, and delay controls.
- **Project structure:** Organized modules for scalability and reuse.

### **Current Challenges:**

- **Large data volume:** Managing ~1M records via incremental collection and potential Dask integration.
- **Data consistency:** Addressing missing or inconsistent fields across years.

### **Expected Challenges:**

- Complex feature engineering (historical success rates, time trends)
- Merging datasets with differing identifiers and formats
- Balancing interpretability and accuracy during model selection

## Next Steps

1. Complete NCES and IMLS data integration
2. Conduct EDA and feature engineering
3. Train baseline logistic regression model
4. Validate using 2024 holdout data

**GitHub:** [github.com/princessmanifest/erate-prospector](https://github.com/princessmanifest/erate-prospector)