University of Maryland College Park Information Challenge 2022 Paycheck protection program: Analysis

Team – IC22-052
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Introduction:

U.S. Small Business Administration:

A United States government agency that provides support to entrepreneurs and small businesses.

Paycheck protection program:

An SBA-backed loan that helps businesses keep their workforce employed during the COVID-19 crisis.

Periodical release of data by SBA on 11.5 million approved applications, of which some applications were later on removed from the database.

Goal: through data analysis, develop an understanding of why these loans might have been removed.



Exploratory data analysis:

Plot: Loan approval date vs no. of applications (Time series graph)



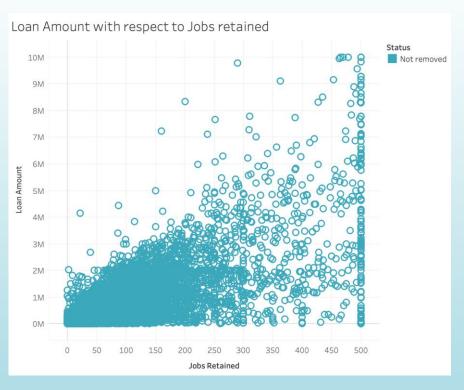
The Consolidated Appropriations Act, 2021 (CAA) extended the Paycheck Protection Program (PPP) through **March 31, 2021**. However, the PPP funds have been exhausted, meaning the PPP program is no longer available as of May 31, 2021.

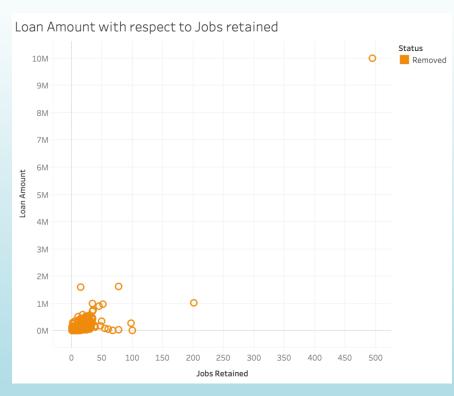
The amount of funds made available in the third round totaled \$284 billion. Maximum loans of \$10 million were available to first-draw borrowers, and loans up to \$2 million were offered to second-draw, small business owners.

Reference: https://www.investopedia.com/your-guide-to-the-paycheck-protection-program-ppp-and-how-to-apply-4802195#:~:text=Round%20three%20funding%20of%20%24284,rounds%20of%20PPP%20loan%20funding.

Exploratory data analysis:

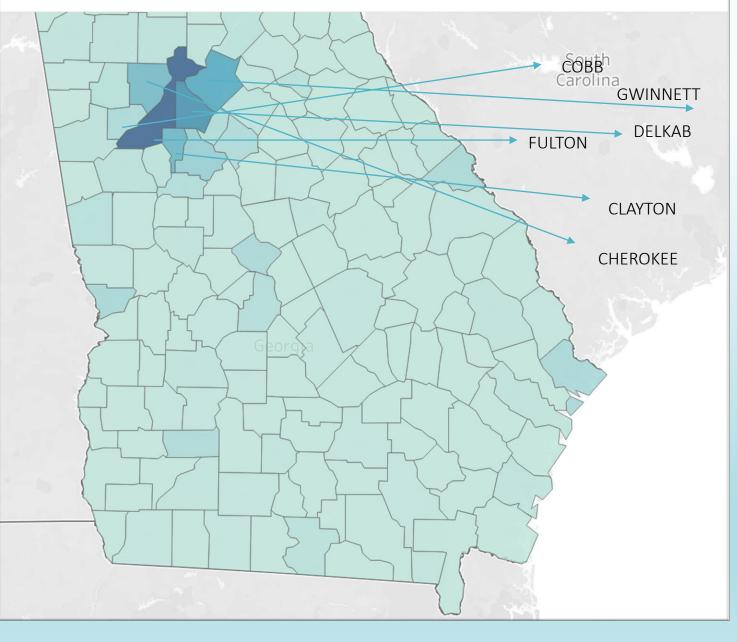
Plots: Jobs retained vs loan amount requested – Full data & removed data





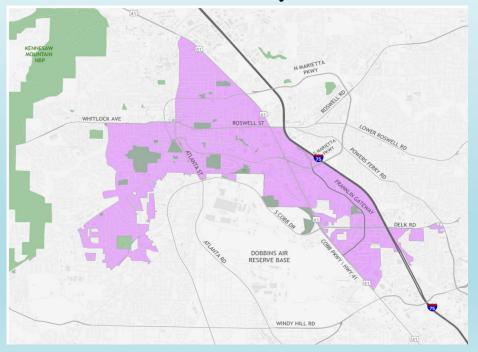
- The Georgia loans of \$14.4 billion as of June 27 represent about 55% of the 278,000 active small businesses in the state, according to the National Federation of Independent Business.
- About 18,300 companies in the state received loans of between \$150,000 and \$10 million, and another 139,000 state-based businesses obtained loans of less than \$150,000, according to the SBA.
- The loans helped companies retain more than 800,000 jobs.

Number of records for each county in Georgia



Georgia: Counties with most applications Historically Underutilized Business zones in pink

- Counties in Georgia with highest number of applications: Fulton, Dekalb, Gwinnett, Clayton, Cherokee, Cobb
- Counties characterized by many small businesses
- Highest no. of PPP applications from these counties as affected by COVID19



Applications breakdown w.r.t Loan status:

| | Loan Status | | | |
|-------------|------------------------|-------------|--------------|--|
| Status | Active Un-Disbursed | Exemption 4 | Paid in Full | |
| Not removed | 344 | 273,238 | 280,243 | |
| Removed | 23,331 | 2,505 | | |

number of removed records for each industry type Naics Code Def 7К 6К 5К Number of records 4K ЗК 2K 1K 0K **Educational Services** Finance and Insurance Manufacturing No value Other Services Public Administration Retail Trade Transportation and Wareh.. Utilities Wholesale Trade Agriculture, Forestry, Fish.. Construction Estate and Rental an.. Mining, Quarrying, and Oil.. Management of Companie. Health Care and Social Ass. Administrative and Suppo. Accommodation and Food Arts, Entertainment, and

Analysis on industry type vs number of applications



Data set: Applications removed from database:



Plot: Industry vs no. of applications

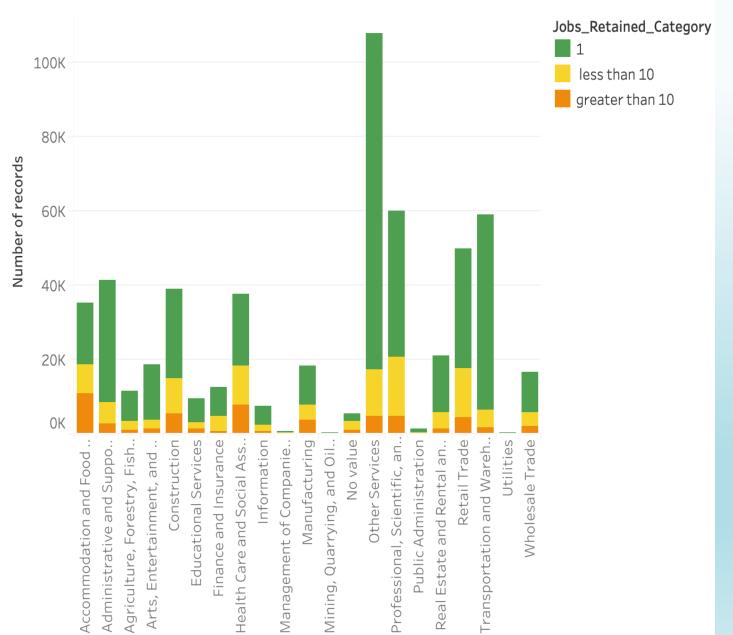


Observation: Highest number of applications filed from the industry 'Other services'

Analysis on industry type (continued):



jobs retained category proportion for each industry type for records not removed



Analysis on industry and jobs retained:





Data set: full data set

Plot: industry vs no. of applications, category as no. of jobs retained



Observation: no. of jobs retained are 1 majorly from the applications from 'other services' industry. other industry applications are also considerably high.

jobs retained category proportion for each industry type for removed records 7К Jobs_Retained_Category less than 10 6К greater than 10 Number of removed records 1K Educational Services Finance and Insurance Manufacturing Mining, Quarrying, and Oil.. No value Other Services Professional, Scientific, an. Wholesale Trade Real Estate and Rental an.. Agriculture, Forestry, Fish. Management of Companie.

Analysis on industry and jobs retained:



Dataset:

Removed applications from database



Plot: Industry vs total number of applications with category as no. of jobs retained.



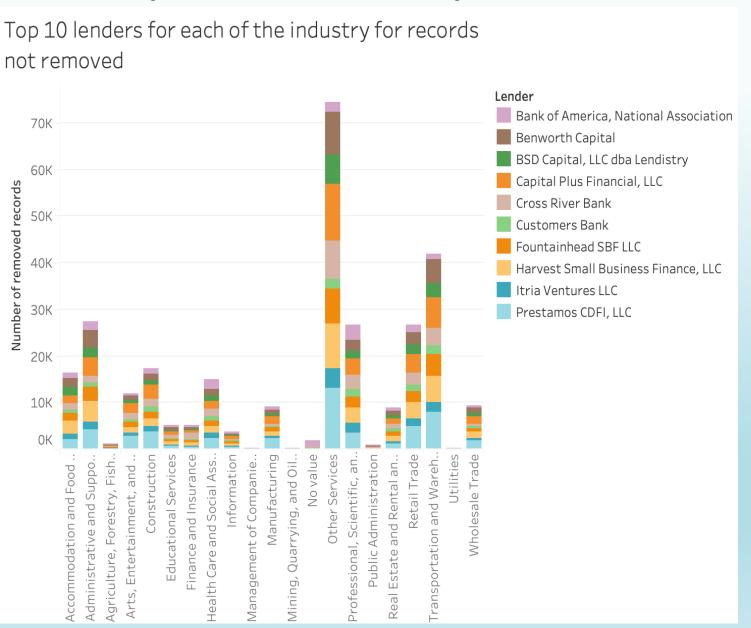
Observation:

Across all of the industries, most of the number of jobs retained =1.



One job is saved for each of the applications in 'Other services' industry

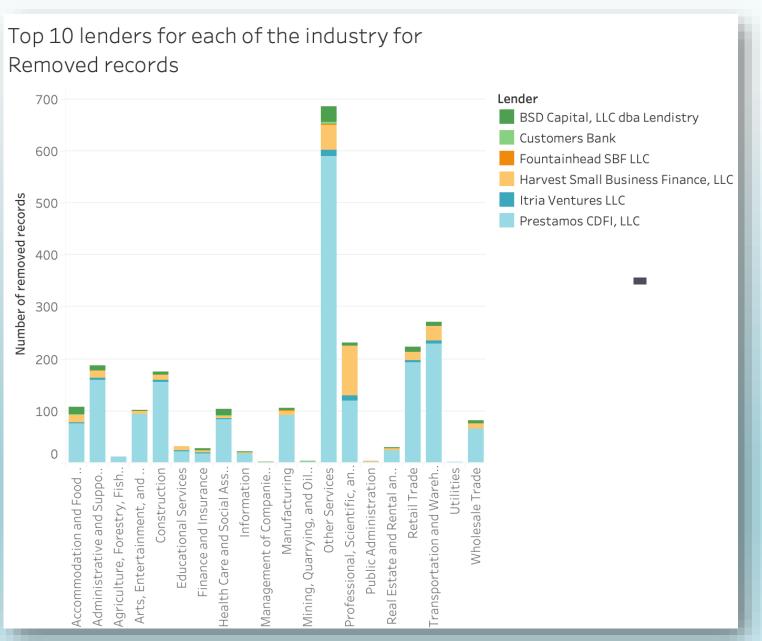
Analysis on Industry vs lender:



- **Dataset:** Full dataset
- **Plot:** Industry vs no. of applications, categorized by lender
- **Filter:** Loan status: Exempt 4 (disbursed but not paid in full or charged off) and paid in full
- **Observation:** Top 10 lenders are as listed.
- Major lenders:

Benworth Captial
Capital Plus Financial LLC
Prestamos CDFI LLC

Analysis on Industry vs lender (continued):



Dataset: Applications removed from

database

Plot: Industry vs no. of applications

categorized by lenders

Filter: Loan status as Exempt 4

(disbursed but not paid in full or charged

off)

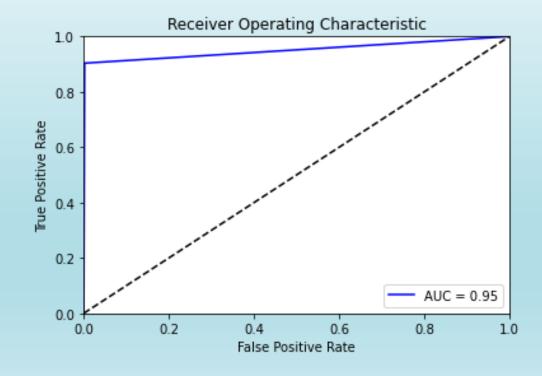
Major lender: Prestamos CDFI LLC



Machine learning: Decision tree

- Predictive model to traverse from observations (branches) to conclusions about the item's target value (leaf node).
- •White-box model, works well on large datasets, little data preparation, in-built feature selection
- •Accuracy & error rate computed through confusion matrix
- •Gini impurity, information gain, variance reduction, measure of goodness

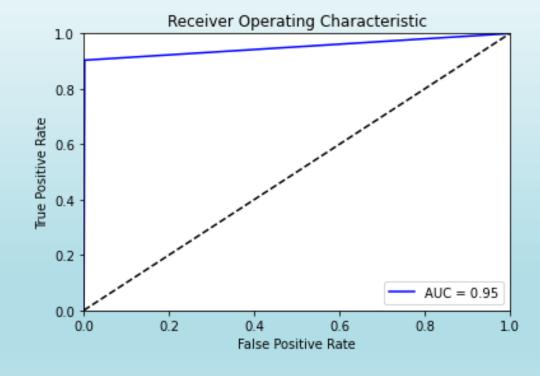
| Test data | Predicted | | |
|-----------|-------------|-------------|---------|
| | Category | Not removed | Removed |
| Actual | Not removed | 165059 | 341 |
| | Removed | 755 | 6994 |



Machine learning: Random forest

- Ensemble learning method builds multitude of decision trees at training time.
- Output for classification -> class selected by most trees.
- Bagging(reduces overfitting), ranking of variable importance, reduces variance, improves accuracy

| Test data | Predicted | | |
|-----------|-------------|-------------|---------|
| | Category | Not removed | Removed |
| Actual | Not removed | 165154 | 246 |
| | Removed | 712 | 7037 |



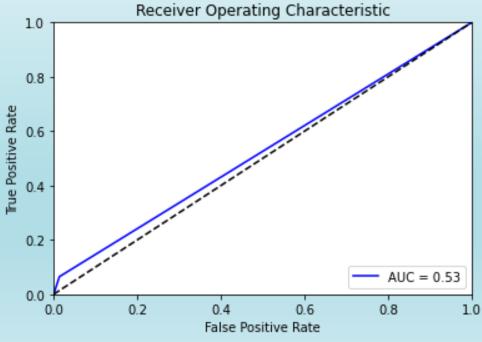
Feature importance:

Bias observed in the data due to the fields 'Loan Status', 'undisbursed amount':

| | Loan Status | | | |
|-------------|------------------------|-------------|--------------|--|
| Status | Active Un-Disbursed | Exemption 4 | Paid in Full | |
| Not removed | 344 | 273,238 | 280,243 | |
| Removed | 23,331 | 2,505 | | |

Due to possible bias, explored model excluding top 3 features. Resulting model is just as good as random guessing.





Literature survey - evidence

References:

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https://www.bizjournals.com/atlanta/news/feature/on-leadership/2021

https://www.forbes.com/advisor/business-loans/how-many-jobs-were-saved-because-of-ppp-loans/

https://www.justice.gov/archives/oip/foia-guide-2004-edition-exemption-4

https://projects.propublica.org/coronavirus/bailouts/faq

 $\frac{https://www.forbes.com/sites/brockblake/2021/04/06/why-ppp-loan-applications-are-getting-stuck-and-how-to-keep-yours-moving/?sh=2e86468e4872$

https://data.onlineathens.com/paycheck-protection-program-loans/

Conclusion:

Considering additional important demographic information from the PPP application form could help us uncover better insights.

Ex: gender, ethnicity etc.