Office of Foreign Labor Certification (OFLC) & EasyVisa

Machine Learning Based Solution using Ensemble Techniques

Julie Kistler _ January 18, 2024



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Executive Summary

Businesses in the United States face a growing demand for human resources, leading to challenges in identifying and attracting the right talent. The Immigration and Nationality Act (INA) permits foreign workers to address this demand. The Office of Foreign Labor Certification (OFLC) administers immigration programs, processing applications for temporary and permanent labor certifications.

To streamline the visa approval process, EasyVisa, hired by OFLC, seeks a machine learning solution. As a data scientist, the goal is to analyze provided data and recommend a classification model that facilitates visa approvals based on key influencing factors.

The model highlights the significance of three key features: education level, job experience, and prevailing wage. These factors play a crucial role in determining visa approval outcomes.

Key Features for Model Importance:

 The top factors influencing visa approval are the education level of the employee, job experience, and prevailing wage. These features significantly contribute to the model's decision-making process.

Executive Summary, Cont.

Insights cont:

- Best profile for Visa Approval:
 - Education Level: Higher the education the better most approvals require a Bachelors' degree or better
 Doctorate and Master's degree are highly preferred
 - Job Experience: Job experience is essential
 - Prevailing Wage: Average prevailing wage is around \$70k.
 - Additional Factors: Applicants from Europe, Africa, and Asia, with yearly unit of wage, and applying to the Mid-West region have higher chances of approval.
- Best profile for Visa Denied:
 - Education Level: High School Education and/or no degree
 - Job Experience: Lack of job experience
 - Prevailing Wage: Average prevailing wage is around \$65k.
 - Additional Factors: Applicants with hourly unit of wage, from Oceania, North America, and South America are more likely to face denial.



Executive Summary, Cont.

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Recommendations:

• Utilize the XGBoostClassifier Tuned model for its outstanding performance, achieving an F1 Score of 83% for the training set and 82% for the testing set. This model is recommended for accurate predictions in the visa approval process.

Recommended Further Analysis:

- Additional Data Collection: Gather more information from both employers and employees to extract deeper insights
 - **Job Type and Prevailing Wage Data**: Analyze prevailing wage data based on job types such as IT, service, administration, etc.
 - Required Education and Years of Experience: Understand the correlation between visa approval and specific education levels and work experience.
 - Industry Sector Analysis: Explore visa approval trends within different industry sectors.
 - Regional Analysis: Segment data based on the type of company (industry sector) and size in different regions.
 - **Visa Length Analysis**: Examine the impact of visa length on approval rates.
 - Applicant Segmentation by Company Size: Classify applicants based on company size (small, median, large) to discern patterns in approved/denied applications.

Business Problem Overview and Solution Approach

The Office of Foreign Labor Certification (OFLC) continues to grapple with a substantial surge in visa applications, leading to a growing backlog. The current manual review process is proving to be laborious and inefficient, hindering the timely processing of applications. This is impacting the overall effectiveness of OFLC's visa approval procedures.

To address the challenges faced by OFLC, EasyVisa has been hired to develop a machine learning-based solution. The proposed solution aims to streamline operations by leveraging a classification model that can analyze and categorize visa applications. EasyVisa's objectives are:

- Facilitate the Visa Approval Process: EasyVisa intends to enhance the efficiency of the visa approval process by automating the initial screening of applications. By leveraging machine learning algorithms, the system will identify and shortlist applicants who are more likely to receive visa approval.
- Recommend Suitable Profiles: The solution will go beyond mere automation and provide OFLC with actionable
 insights. EasyVisa's classification model will consider various factors influencing the visa approval process and
 recommend suitable profiles for certification or denial. This approach ensures that the decision-making process is
 not only expedited but also guided by data-driven insights.

In summary, the collaboration between OFLC and EasyVisa seeks to transform the visa application process by, introducing efficiency through automation and informed decision-making based on machine learning analysis. This solution is poised to alleviate the strain caused by the increasing number of applications, ultimately leading to a more streamlined and effective visa approval workflow.

Data Overview

	Data Dictionary
case_id	ID of each visa application
continent	Information of continent the employee
education_of_employee	Information of education of the employee
has_job_experience	Does the employee has any job experience? Y= Yes; N = No
requires_job_training	Does the employee require any job training? Y = Yes; N = No
no_of_employees	Number of employees in the employer's company
yr_of_estab	Year in which the employer's company was established
region_of_employment	Information of foreign worker's intended region of employment in the US
prevailing_wage	Average wage paid to similarly employed workers in a specific occupation in the area of intended employment. The purpose of the prevailing wage is to ensure that the foreign worker is not underpaid compared to other workers offering the same or similar service in the same area of employment
unit_of_wage	Unit of prevailing wage. Values include Hourly, Weekly, Monthly, and Yearly
full_time_position	Is the position of work full-time? Y = Full Time Position; N = Part Time Position
case_status	Flag indicating if the Visa was certified or denied

Data Overview cont...

Column	Dtype
case_id	object
continent	object
education_of_employee	object
has_job_experience	object
requires_job_training	object
no_of_employees	int64
yr_of_estab	int64
region_of_employment	object
prevailing_wage	float64
unit_of_wage	object
full_time_position	object
case_status	object

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Rows	Columns
25480	12

- There are no duplicate values
- There are no missing values

- 9 object data types (case_id, continent, education_of_employee, has_job_experience, requires_job_training, region_of_employment, unit_of_wage, full_time_position, case_status)
- 2 integer data types (no_of_employees, yr_of_estab)
- 1 float data type (prevailing_wage)

EDA Results



Statistical Summary

- no_of_employees, yr_of_estab, prevailing_wage columns are all numerical features the remaining columns are objects
- The no_of_employees has a mean of 5667 with a median of 2109 indicating the distribution of data may be skewed
- There are negative numbers in the no_of_employees this could indicate an error
- The yr_of_estab has a broad range from 1800 2016
- The most prevalent continent is Asia
- The most prevalent education level is a Bachelors Degree
- Most applicants do not require job training
- Most applicants do have job experience
- The most prominent region is the Northeast
- The average prevailing wage is ~ \$75,456 (Min: \$2.14 / Max: \$319210) data distribution is skewed
- Annual salary if the most prevalent unit of wage
- Most applicants are applying for a full time job
- A majority of applicants are certified ~ 67%

EDA Results

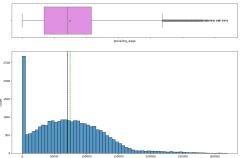


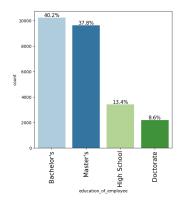
- Correct the negative values in the number of employees column
 - Assuming these negative numbers may be a result of a data entry error
 - There are 33 instances -- these were converted by using the absolute value of the numbers
- Number of observations for each unique category of the categorical variables
 - In the case_id category, there are 25,480 unique values
 - In the continent category, Asia has the most applicants at 16891 (66%).
 - o In the education of employee category, Bachelor's degree is the top with 10234 (40%)
 - In the has_job_experience category, over half have experience at 14802 (58%)
 - o In the requires_job_training catagory, a vast majority do not need training at 22525 (88%)
 - o In the region_of_employment catagory, the northeast the most requested at 7195 (28%) with the south close behind with 7017 (27%) over 50% of the applicants are applying in these two regions.
 - In the unit_of_wage category, the most prevalent wage is year at 22962 (90%)
 - In the full_time_position category, the majority of applicants are full time at 22773 (89%)
 - In the case_status category, approx. 2/3 are certified at 17018 (67%)
- Removed "case_id" from data as it is not needed for analysis

EDA Results _ Univariate Analysis (Observations)



- Number of employees data distribution is heavily skewed right with lots of outliers
 - O This may indicate there is a large variety of company sizes
- Prevailing wage data distribution is skewed right with lots of outliers
 - O There are wages above the 200k mark
 - There is a large disparity between wages
 - There appears to be several wages at or close to the 0 mark further analysis is recommended
 - There were 176 rows that have a prevailing hourly wage of less than 100
- Continent data indicated approximately 66% of all applications are originated out of Asia and 28% originated from Europe and North America
- The education of employee data reflected that approximately 87% of applicants had a higher education degree with the Bachelor's degree being the most common

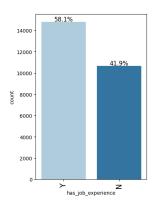


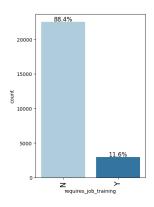


EDA Results _ Univariate Analysis (Observations)



Approximately 58% of applicants have job experience and approximately 88% of applicants do not require job training.

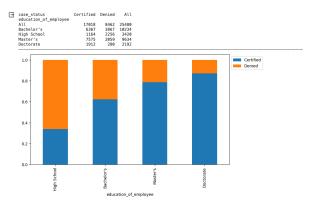


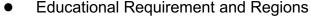


- The region of employment looks to be pretty evenly distributed across three regions (Northeast, South, and West). Approximately 82% of applications identify one of these three regions
- Approximately 90% of all applications have an hourly unit of wage
- Approximately 67% of the Visas were certified

EDA Results _ Bivariate Analysis (Observations)

- The level of education does appear to have an impact on visa certifications
 - Those with higher level of education appear to have a greater chance to be certified



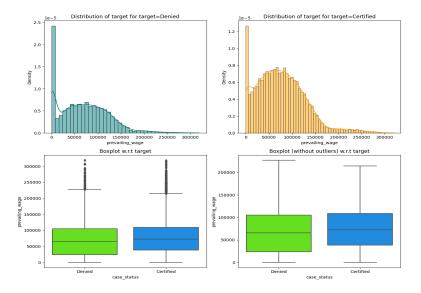


- O The west has the highest demand for applicants with a Doctoral education requirement
- O The northeast region has the highest demand for applicants with a Master's degree requirement
- O The south has the highest demand for applicants with a Bachelor's degree requirement
- The south has the highest demand for applicants with a High School education requirement
- The Midwest Region currently has the highest certification rate (~75%)
- Applicants from Europe appear to have the highest chance of certification while South America has the lowest
- Applicants with work experience are more likely get certified as they are less likely to need job training



EDA Results _ Bivariate Analysis (Wage Observations)

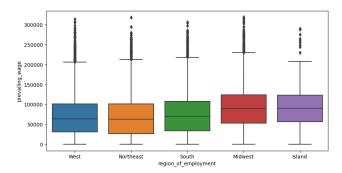
• It appears the median prevailing wage is slightly higher for applicants that obtain a visa certification



EDA Results _ Bivariate Analysis (Wage Observations)

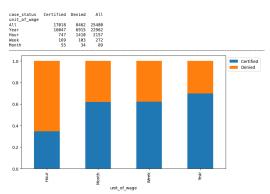






• It appears applicants with "Year" unit of wage have a higher chance to obtain visa certification vs. the

applicants that have "hour" unit of wage



Data Preprocessing

- There were no duplicates found. no treatment necessary
- There were no missing values founds no treatment necessary
- Outlier check completed all data appears to be in the dataset and appears to be valuable information no treatment necessary
- Feature engineering
 - The "case_status column was encoded a 1 for certified and 0 for all other values
- Data preparation for modeling
 - The "case status" column was dropped from the feature set X.
 - Dummy variables were created for categorical features in X using pd.get_dummies().
 - The data was split into training and testing sets using with a 70:30 ratio, and the stratify parameter was set to ensure that the class distribution is maintained in the splits.

Shape of Training and Testing Data Sets

Shape of Training Set	(17836, 21)
Shape of Testing Set	(7644, 21)

Percentage of Classes

Percentage of Classes	Training Set	Test Set
0	0.667919	0.667884
1	0.332081	0.332156

- The model_performance_classification_sklearn function was used to check the model performance of models
- The confusion matrix sklearn function was used to plot the confusion matrix
- The F1 Score was used as an evaluation metric because it considers both false positives and false negatives
- Balanced class weights were used so the model focuses equal on both classes



Model Performance Summary

It is recommended to use the XGBoost Classifier Tuned model as it appears to be performing
the best reflecting the best F1 scores for the Training performance and the testing performance

• XGBoost Classifier Tuned (Training) F1 Score: .832

XGBoost Classifier Tuned (Test) F1 Score: .821

- The top feature of importance for the model are:
 - 1) Education of Employee _ High School
 - 2) Has Job Experience _y
 - 3) Prevailing wage
 - 4) Education of Employee _ Masters
 - 5) Education of Employee _ Doctorate

Training Performance Comparison

index	Decision Tree	Tuned Decision Tree	Bagging Classifier	Tuned Bagging Classifier	Random Forest	Tuned Random Forest	Adaboost Classifier	Tuned Adaboost Classifier	Gradient Boost Classifier	Tuned Gradient Boost Classifier	XGBoost Classifier	XGBoost Classifier Tuned	Stacking Classifier
Accuracy	1.0	0.7125476564252075	0.9851984749943934	0.9961874859834043	1.0	0.7691186364655752	0.7382260596546311	0.7544292442251626	0.7588024220677282	0.7561673020856694	0.8508073559093967	0.7621103386409509	0.7647454586230097
Recall	1.0	0.9319231092084278	0.9859817006631411	0.9999160580878033	1.0	0.9186602870813397	0.8871820700075548	0.8839083354318812	0.8837404516074876	0.8852514060270293	0.9359523209938723	0.8881893729539159	0.8871820700075548
Precision	1.0	0.7200674536256324	0.9918095077260829	0.9944068787043994	1.0	0.7765557368906549	0.7606880667914208	0.7784431137724551	0.7830420230568985	0.7795683027794205	0.8545370938074801	0.7842425140824192	0.7874972058713956
F1	1.0	0.8124108155574256	0.988887018016501	0.9971538590323121	1.0	0.841651926478505	0.8190800945479909	0.8278301886792453	0.8303493966401135	0.8290554616563816	0.8933936941628942	0.8329856327494588	0.834372779663693

Testing Performance Comparison

index	Decision Tree	Tuned Decision Tree	Bagging Classifier	Tuned Bagging Classifier	Random Forest	Tuned Random Forest	Adaboost Classifier	Tuned Adaboost Classifier	Gradient Boost Classifier	Tuned Gradient Boost Classifier	XGBoost Classifier	XGBoost Classifier Tuned	Stacking Classifier
Accuracy	0.6648351648351648	0.706567242281528	0.6915227629513344	0.7242281527995814	0.727367870225013	0.7380952380952381	0.7343014128728415	0.7411041339612768	0.7447671376242805	0.7437205651491365	0.7299843014128728	0.7448979591836735	0.7422815279958137
Recall	0.7428011753183154	0.9308521057786484	0.764152791380999	0.8953966699314397	0.8472086190009794	0.898922624877571	0.8850146914789422	0.8760039177277179	0.8760039177277179	0.8787463271302645	0.8515181194906954	0.8777668952007835	0.8736532810969637
Precision	0.752231700059512	0.7154471544715447	0.7717111770524233	0.7438567941415786	0.7683425119914727	0.7553909465020576	0.75779939617578	0.7686490202818838	0.7723661485319516	0.769996567112942	0.768972227136034	0.7716548992595144	0.7709593777009507
F1	0.7474866942637493	0.8090576317357624	0.7679133858267716	0.812622222222222	0.805850568287684	0.8209302325581396	0.8164814312821903	0.8188226677652658	0.8209270307480495	0.8207849236117464	0.808142777467931	0.8212976539589443	0.8191000918273645



Insights and Recommendations



Insights:

- Key Features for Model Importance:
 - The top factors influencing visa approval are the education level of the employee, job experience, and prevailing wage. These features significantly contribute to the model's decision-making process.
- Best profile for Visa Approval:
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Insights and Recommendations



Recommendations:

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 - **Applicant Segmentation by Company Size**: Classify applicants based on company size (small, median, large) to discern patterns in approved/denied applications.



APPENDIX



EDA

Statistical Summary of the Data



	count	unique	top	freq	mean	std	min	25%	50%	75%	max
case_id	25480	25480	EZYV01	1	NaN	NaN	NaN	NaN	NaN	NaN	NaN
continent	25480	6	Asia	16861	NaN	NaN	NaN	NaN	NaN	NaN	NaN
education_of_employee	25480	4	Bachelor's	10234	NaN	NaN	NaN	NaN	NaN	NaN	NaN
has_job_experience	25480	2	Υ	14802	NaN	NaN	NaN	NaN	NaN	NaN	NaN
requires_job_training	25480	2	N	22525	NaN	NaN	NaN	NaN	NaN	NaN	NaN
no_of_employees	25480.0	NaN	NaN	NaN	5667.04321	22877.928848	-26.0	1022.0	2109.0	3504.0	602069.0
yr_of_estab	25480.0	NaN	NaN	NaN	1979.409929	42.366929	1800.0	1976.0	1997.0	2005.0	2016.0
region_of_employment	25480	5	Northeast	7195	NaN	NaN	NaN	NaN	NaN	NaN	NaN
prevailing_wage	25480.0	NaN	NaN	NaN	74455.814592	52815.942327	2.1367	34015.48	70308.21	107735.5125	319210.27
unit_of_wage	25480	4	Year	22962	NaN	NaN	NaN	NaN	NaN	NaN	NaN
full_time_position	25480	2	Υ	22773	NaN	NaN	NaN	NaN	NaN	NaN	NaN
case_status	25480	2	Certified	17018	NaN	NaN	NaN	NaN	NaN	NaN	NaN

Counts for each unique category (Categorical Variables)

```
EZYV01
EZYV16995
EZYV16993
EZYV16992
EZYV16991
EZYV8492
EZYV8491
EZYV8490
EZYV8489
EZYV25480
Name: case_id, Length: 25480, dtype: int64
Asia
                 16861
                  3732
Europe
                  3292
North America
South America
                   852
Africa
                   551
Oceania
                  192
Name: continent, dtype: int64
Bachelor's
              10234
Master's
                9634
High School
                3420
               2192
Doctorate
Name: education_of_employee, dtype: int64
     14802
     10678
Name: has_job_experience, dtype: int64
```

```
22525
      2955
Name: requires_job_training, dtype: int64
Northeast
             7195
South
             7017
West
             6586
Midwest
             4307
Island
             375
Name: region_of_employment, dtype: int64
         22962
Year
          2157
Hour
Week
Month
Name: unit of wage, dtype: int64
    22773
      2707
Name: full_time_position, dtype: int64
Certified
             17018
Denied
              8462
Name: case_status, dtype: int64
```



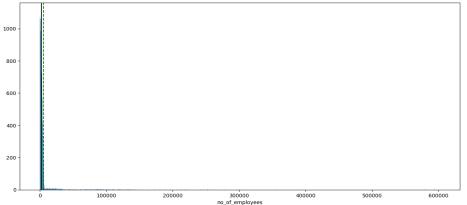


Univariate Analysis

EDA _Univariate Analysis

Number of Employees



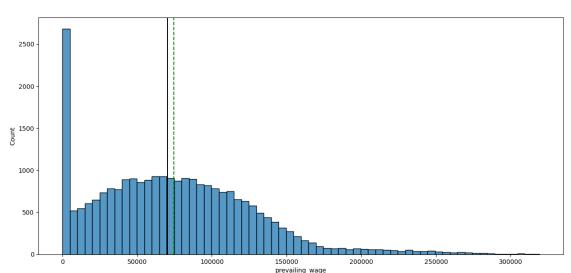




- Data appears heavily skewed right with lots of outliers
- Data indicates a large variety of company size

Prevailing Wage

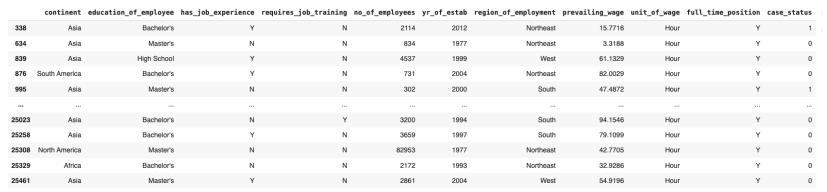






- Data appears skewed right with lots of outliers
- There is a large disparity within the prevailing wage data
- There is a large amount of wages around the 0 mark – this may require further analysis
- There may be a data entry error in the data using hourly data
- There are wages above the 200k mark

Prevailing Wage _ Analysis less than 100



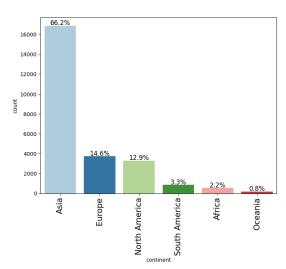
176 rows x 11 columns

176 rows had less than 100 in hourly prevailing wage



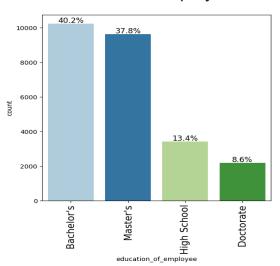


Continent Observations



- ~66% Asia
- ~15% Europe
- ~13% North America
- ~3% South America
- ~2% from Africa
- ~1% from Oceania

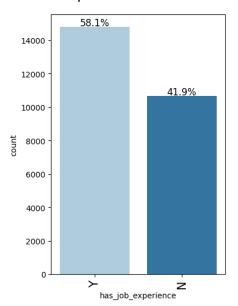
Education of Employee Observations



- Bachelor's degree is the top education (~40%)
- Master's degree is second highest (~38%)
- Approximately 87% have a degree in higher education

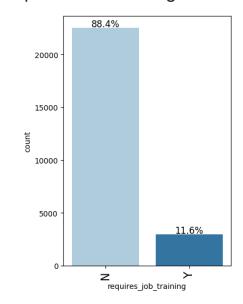


Job Experience Observations



- ~58% of applicants have job experience
- ~42% do not have job experience

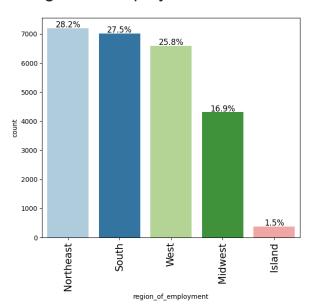
Requires Job Training Observations



- ~88% of applicants do not need job training
- ~12% of applicants will need job training

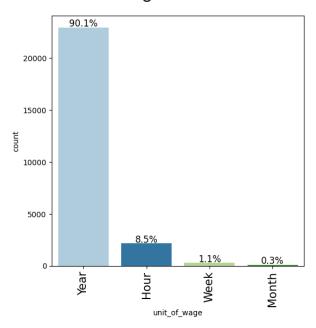


Region of Employment Observations



- Data looks pretty evenly distributed with the top three regions
- ~82% of applications designate the Northeast, South and West regions.
- ~17% of applicants designate the Midwest
- The island region is the lest designated at 1.5%

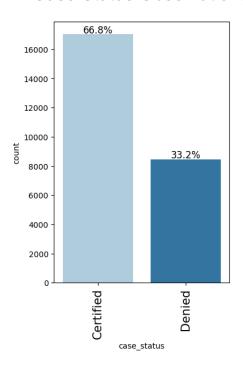
Unit of Wage Observations



~90% of applicants have a yearly unit of wage



Case Status Observations



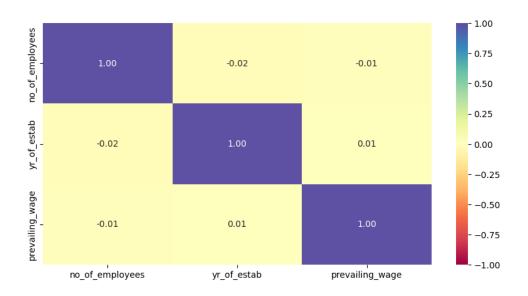
- ~67% of all visas are certified
- ~33% of all visas are denied



Bivariate Analysis

EDA _Bivariate Analysis



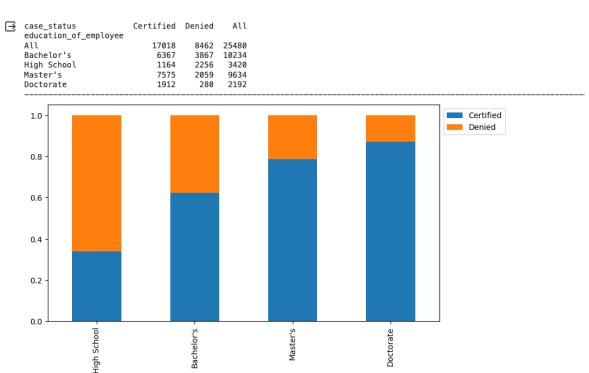


 There does not appear to be any correlation among the numerical variables





Does education level impact visa certification?



education of employee

 Yes, it appears the higher the level of education to better chance to have the visa certified





Regions vs. diverse talent and educational backgrounds

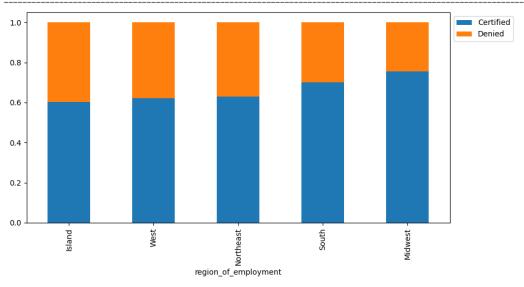


- High school education requirement if highest in the South Region followed by the Northeast Region
- Bachelor's Degree education requirement if highest in South Region followed by the West Region
- Master's Degree education requirement if highest in the Northeast Region followed by the South Region
- Doctorate Degree education requirement is highest in the West Region followed by the Northeast Region



Number of visa certification across each region

case_status	Certified	Denied	All
region_of_employment			
All	17018	8462	25480
Northeast	4526	2669	7195
West	4100	2486	6586
South	4913	2104	7017
Midwest	3253	1054	4307
Island	226	149	375



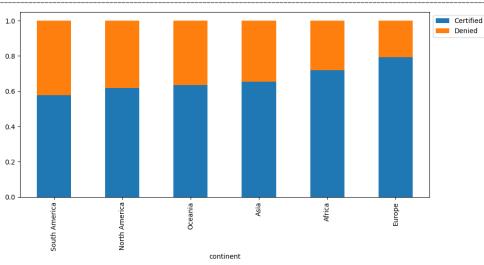
- The Midwest region has the highest certification rate
- The Island region has the lowest certification rate





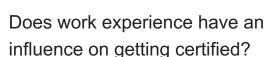
Number of visa certification across continents

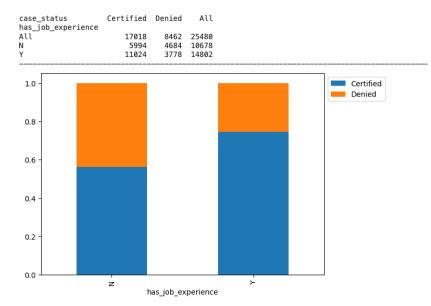
case_status	Certified	Denied	All
continent			
All	17018	8462	25480
Asia	11012	5849	16861
North America	2037	1255	3292
Europe	2957	775	3732
South America	493	359	852
Africa	397	154	551
Oceania	122	70	192



- Europe has the highest certification rate
- South America has the lowest certification rate



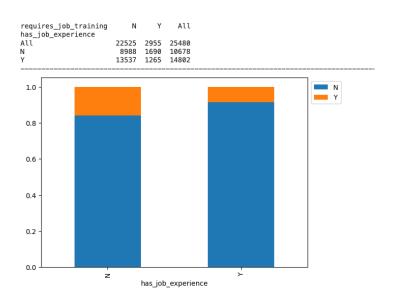




Having work experience increases the applicants chances to get certified



Do the employees who have prior work experience require any job training?

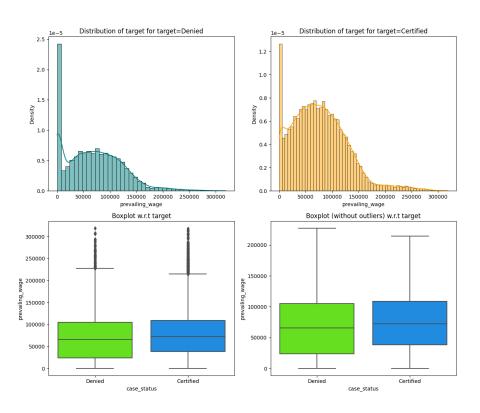


 Applicants that do have job experience and less likely going to need job training

EDA _Bivariate Analysis, cont...

AST ISA

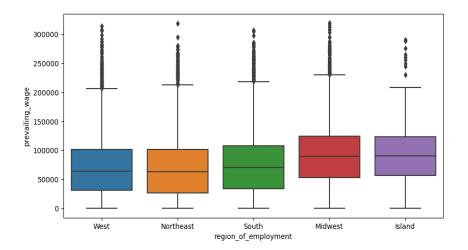
Does Visa status change with prevailing wage?



 It appears the median prevailing wage of applicants that have obtained a visa certification is slightly higher than those applicates that were denied.

EDA _Bivariate Analysis, cont...

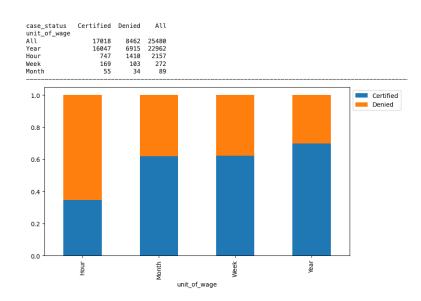
 Is the prevailing wage is similar across all the regions of the US?



 The prevailing wages appear to higher in the Midwest and Island regions



 Does the type of prevailing wage unit type have any impact on certification?



 It appears applicants with "year" units of wage have a higher can for certification where applicants with a "hour" unit of wage have the lowest chance for certification

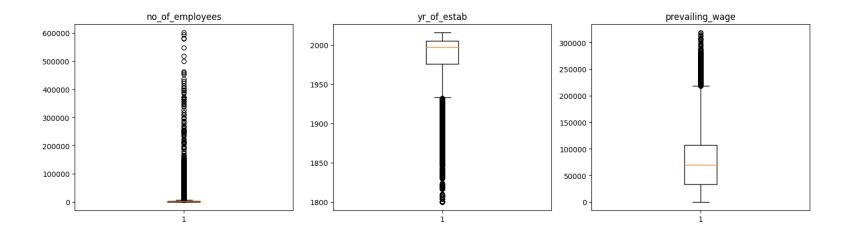


Data Preprocessing

Data Preprocessing _ Outlier Check



- There are quite a few outliers in the data- these all seem to appear in the dataset
- However, we will not treat them as they provide valuable information



Data Preprocessing _ Train and Test Set



- The data was split into training and testing sets using with a 70:30 ratio
- The stratify parameter was set to ensure that the class distribution is maintained in the splits

```
Shape of Training set: (17836, 21)
Shape of test set: (7644, 21)
Percentage of classes in training set:
1  0.667919
0  0.332081
Name: case_status, dtype: float64
Percentage of classes in test set:
1  0.667844
0  0.332156
Name: case_status, dtype: float64
```

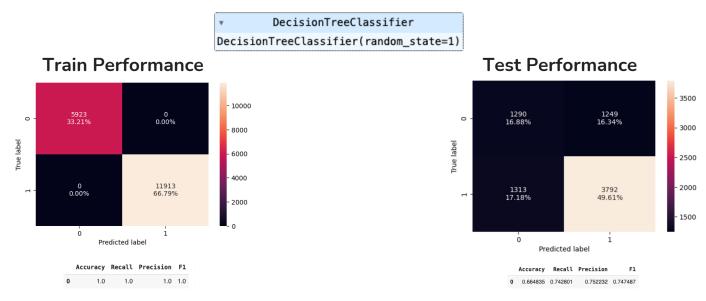


Model Building

Model Building - Bagging (Decision Tree)

Decision Tree Model building steps

• Used the DecisionTreeClassifier function with random state = 1



- The model is overfitting in the train data (F1 is 1.0)
- The test data is not performing as well (F1 is .747)



Model Building – Bagging (Decision Tree) cont...

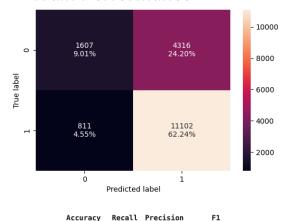
(SEPTER)

Hyperparameter Tuning - Decision Tree

• Used the DecisionTreeClassifier function (class_weight='balanced', random state = 1)

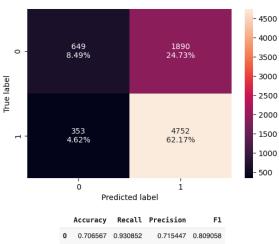
Train Performance

0.712548 0.931923



0.720067 0.812411

Test Performance



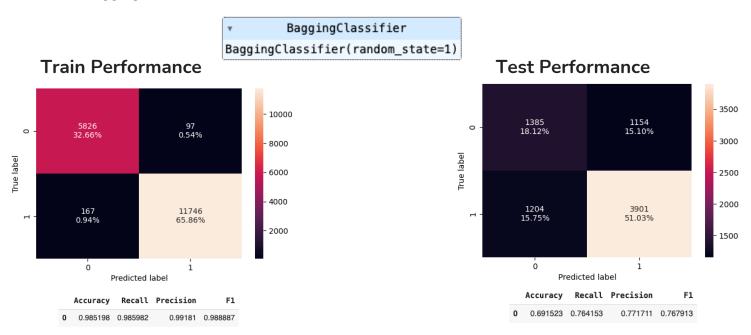
The model is demonstrating a good fit and is not overfitting

Model Building - Bagging (Bagging Classifier)



Bagging Classifier Model building steps

Used the BaggingClassifier function with random state = 1



Model Building - Bagging (Bagging Classifier) cont...



- 4500 - 4000

- 3500

- 3000

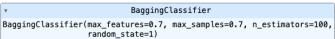
- 2500 - 2000

1500

1000

Hyperparameter Tuning – Bagging Classifier

Used the BaggingClassifier function with random state = 1



Train Performance Test Performance 10000 965 5856 0 12.62% 20.59% 0.38% 32.83% 8000 True label 6000 4000 4571 1 0.01% 11912 6.99% 59.80% 66.79% 2000 0 Predicted label Predicted label Recall Precision F1 Recall Precision F1 0.994407 0.997154 0.743857 0.812622 0 0.996187 0.999916 0.895397

The model appears to be overfitting even after tuning

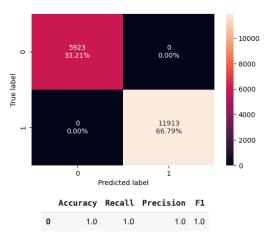
Model Building – Bagging (Random Forest)

Random Forest Model building steps

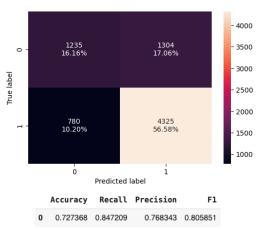
• Used the RandomForestClassifier (random state = 1, class_weight='balanced',)



Train Performance



Test Performance



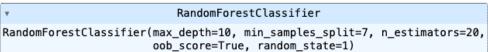
- The random forest classifier appears to be overfitting in the train data (F1 is 1.0)
- Test data is not performing as well (F1 is .805)



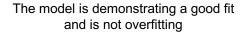
Model Building – Bagging (Random Forest), cont...

Hyperparameter Tuning - Random Forest

• Used the RandomForestClassifier (random state = 1, obb_score=True, bootstrap=True)



Train Performance Test Performance 4500 - 10000 4000 1486 2774 3149 0 13.78% 19.44% 3500 15.55% 17.66% 8000 - 3000 True label 6000 - 2500 2000 4000 10944 4589 1500 61.36% 6.75% 60.03% 1000 2000 0 1 Predicted label Predicted label Recall Precision F1 Accuracy Recall Precision F1 0.755391 0.82093 0.769119 0.91866 0.776556 0.841652



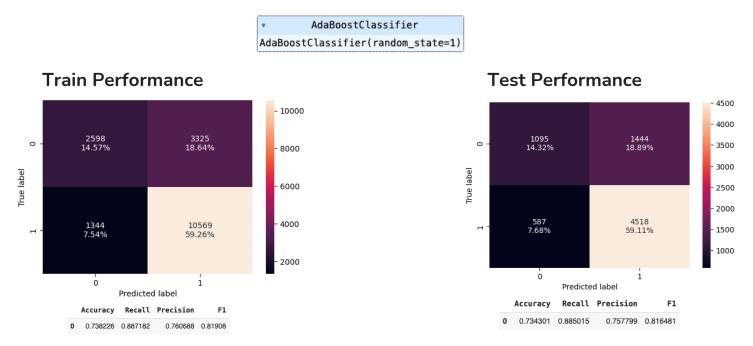


Model Building – Boosting (AdaBoost)

JET SA

Boosting Classifier Model building steps

Used the AdaBoostClassifier (random state = 1)



The model is demonstrating a good fit and is not overfitting

Model Building – Boosting (AdaBoost)



Hyperparameter Tuning – AdaBoost Classifier

Used the AdaBoostClassifier (random state = 1)



The model is demonstrating a good fit and is not overfitting – The F1 score has increased in both the train and test data

Model Building - Boosting (GradientBoosting)

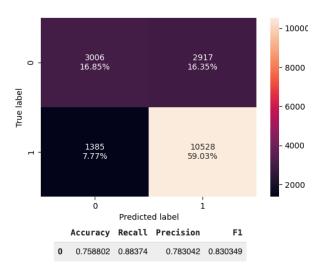
SET SA

Boosting Classifier Model building steps

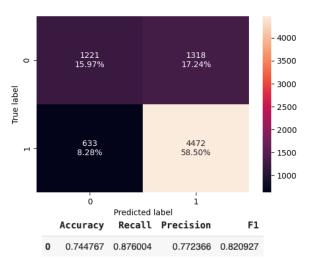
• Used the GradientBoostingClassifier (random state = 1)

v GradientBoostingClassifier
GradientBoostingClassifier(random_state=1)

Train Performance



Test Performance



The model is demonstrating a good fit and is not overfitting

Model Building – Boosting (GradientBoosting)



Hyperparameter Tuning – Gradient Boosting Classifier

• Used the GradientBoostingClassifier(init=AdaBoostClassifier(Random_State=1), Random_State=1



The model is demonstrating a good fit and is not overfitting – The F1 score has decreased in both the train and test data

Model Building - Boosting (XGBoost)

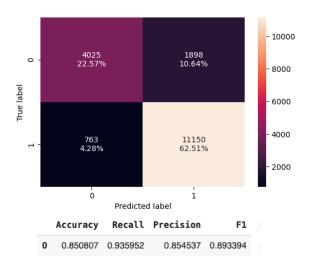
XGBoost Classifier Model building steps

Used the XGBClassifier(random_state=1, eval_metrics='logloss')

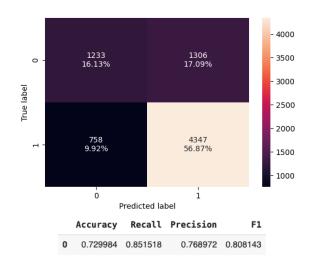


XGBClassifier(base_score=None, booster=None, callbacks=None, calsample_bylevel=None, colsample_bylorde=None, colsample_bylorde=None, colsample_bylorde=None, calsample_bylorde=None, early_stopping_rounds=None, enable_categorical=False, eval_metric='logloss', feature_types=None, gamma=None, grow_policy=None, importance_type=None, interaction_constraints=None, learning_rate=None, max_bin=None, max_cat_the=None, howe, max_cat_to_onehot=None, max_delta_step=None, max_delthe=None, max_delthe=None, max_delthe=None, max_delthe=None, max_delthe=None, missing=nan, monotone_constraints=None, multi_strategy=None, n_estimators=None, n_fobs=None, number_allel_tree=None, random_state=1,...)

Train Performance



Test Performance

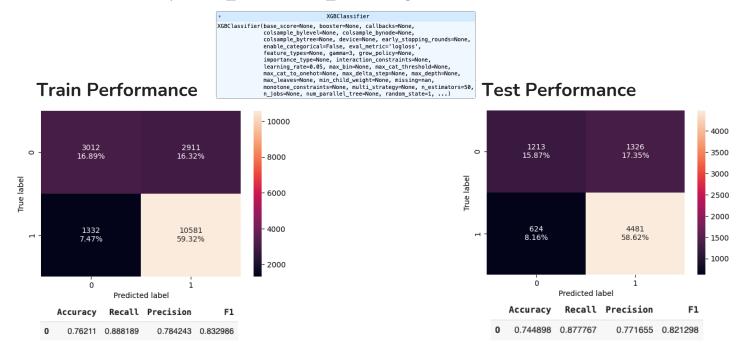


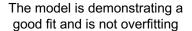
It appears the XGBoost may be overfitting

Model Building – Boosting (XGBoost)

Hyperparameter Tuning – XGBoost Classifier

Used the XGBClassifier(random_state=1, eval_metric='logloss'

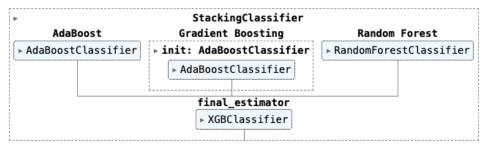


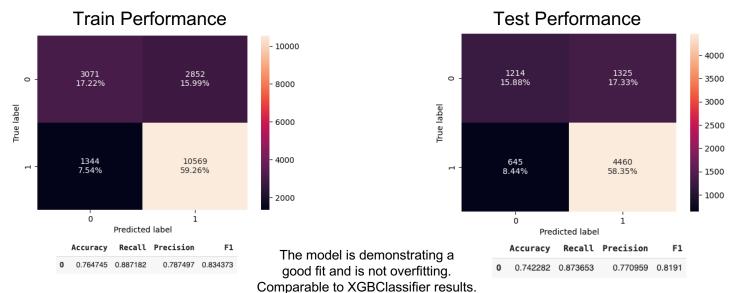




Stacking Classifier







F1 Scores are Train (.83) and Test (.81)

Feature Importance

