

# PART 1

## distribution of integration and productivity scores

In [1]:

```
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns

file_path = '/Users/hjk2160@columbia.edu/Desktop/dataset.csv'
students_df = pd.read_csv(file_path)

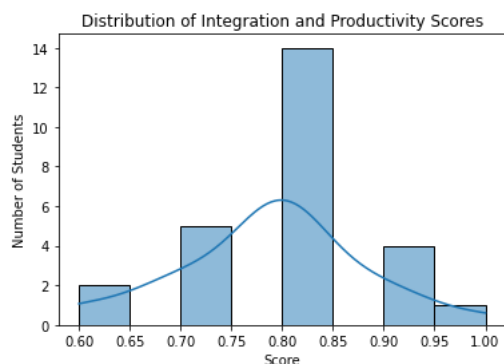
students_df.columns =
students_df.columns.str.strip().str.lower().str.replace(' ', '_')

students_df['integration_productivity_score'] =
students_df['graduation_points'] + students_df['language_scores'] +
students_df['prioritization_scores']
students_df['cultural_fit_score'] = students_df['community_involvement'] +
students_df['team_player_scores']

students_df['integration_productivity_score'] /=
students_df['integration_productivity_score'].max()
students_df['cultural_fit_score'] /=
students_df['cultural_fit_score'].max()
```

In [2]:

```
sns.histplot(data=students_df, x='integration_productivity_score',
kde=True)
plt.title('Distribution of Integration and Productivity Scores')
plt.xlabel('Score')
plt.ylabel('Number of Students')
plt.show()
```

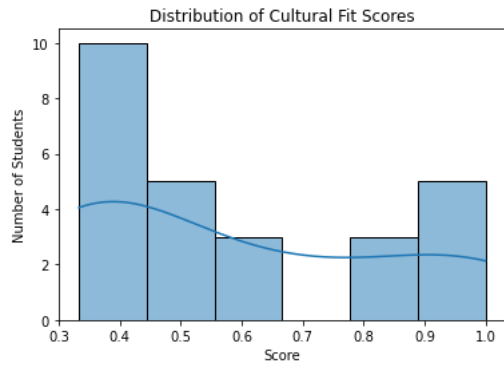


## Distribution of Cultural Fit Scores

In [3]:

```
sns.histplot(data=students_df, x='cultural_fit_score', kde=True)
plt.title('Distribution of Cultural Fit Scores')
```

```
plt.xlabel('Score')
plt.ylabel('Number of Students')
plt.show()
```



## Correlation Heatmap

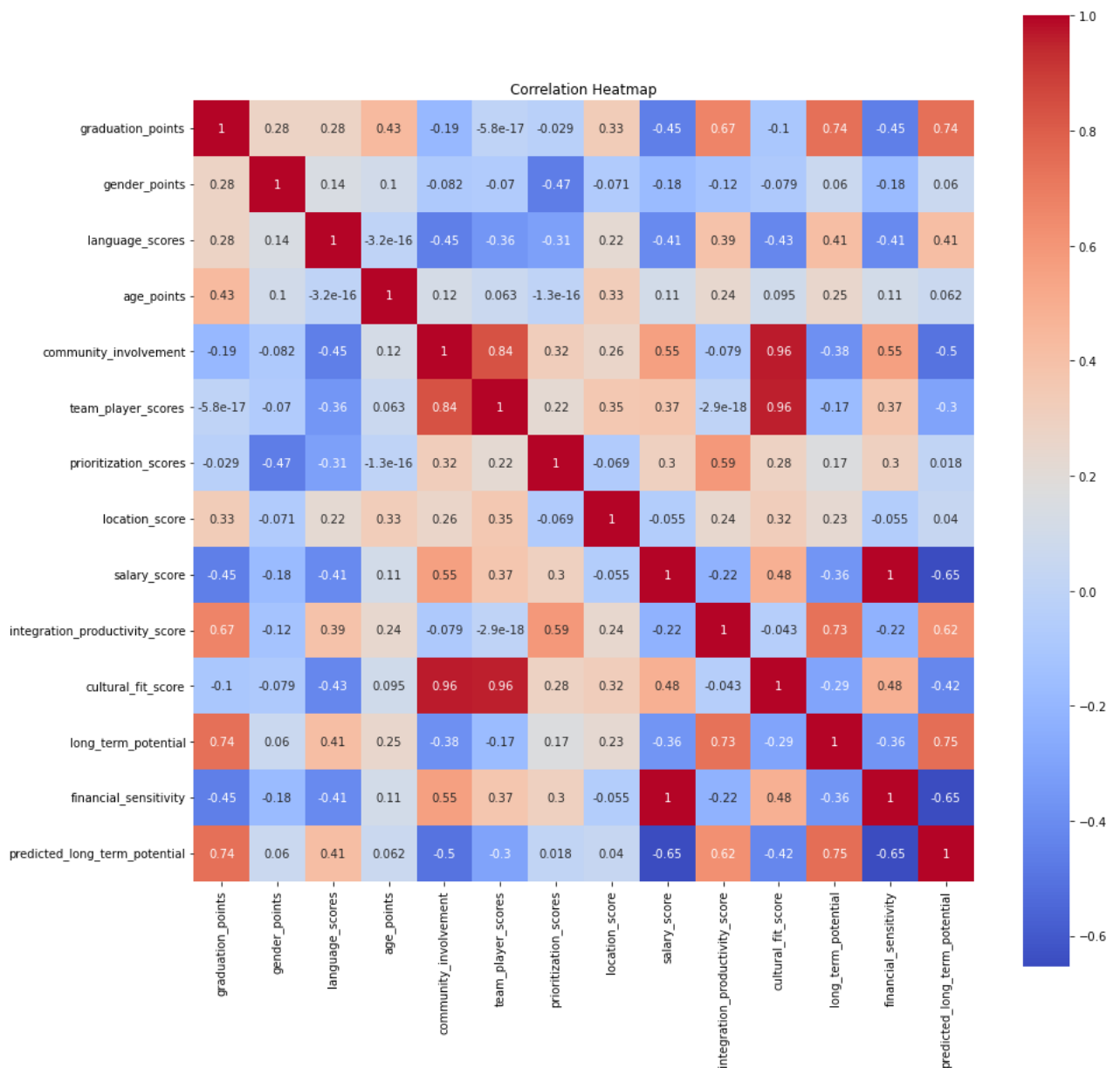
```
file_path2 = '/Users/hjk2160@columbia.edu/Desktop/dashboard_data.csv'
students_df2 = pd.read_csv(file_path2)
students_df2.columns =
students_df2.columns.str.strip().str.lower().str.replace(' ', '_')
```

In [4]:

#히트맵을 통한 데이터 상관관계 확인하기

```
plt.figure(figsize=(15, 15))
corr = students_df2.corr()
sns.heatmap(corr, annot=True, cmap='coolwarm', square=True)
plt.title('Correlation Heatmap')
plt.show()
```

In [5]:

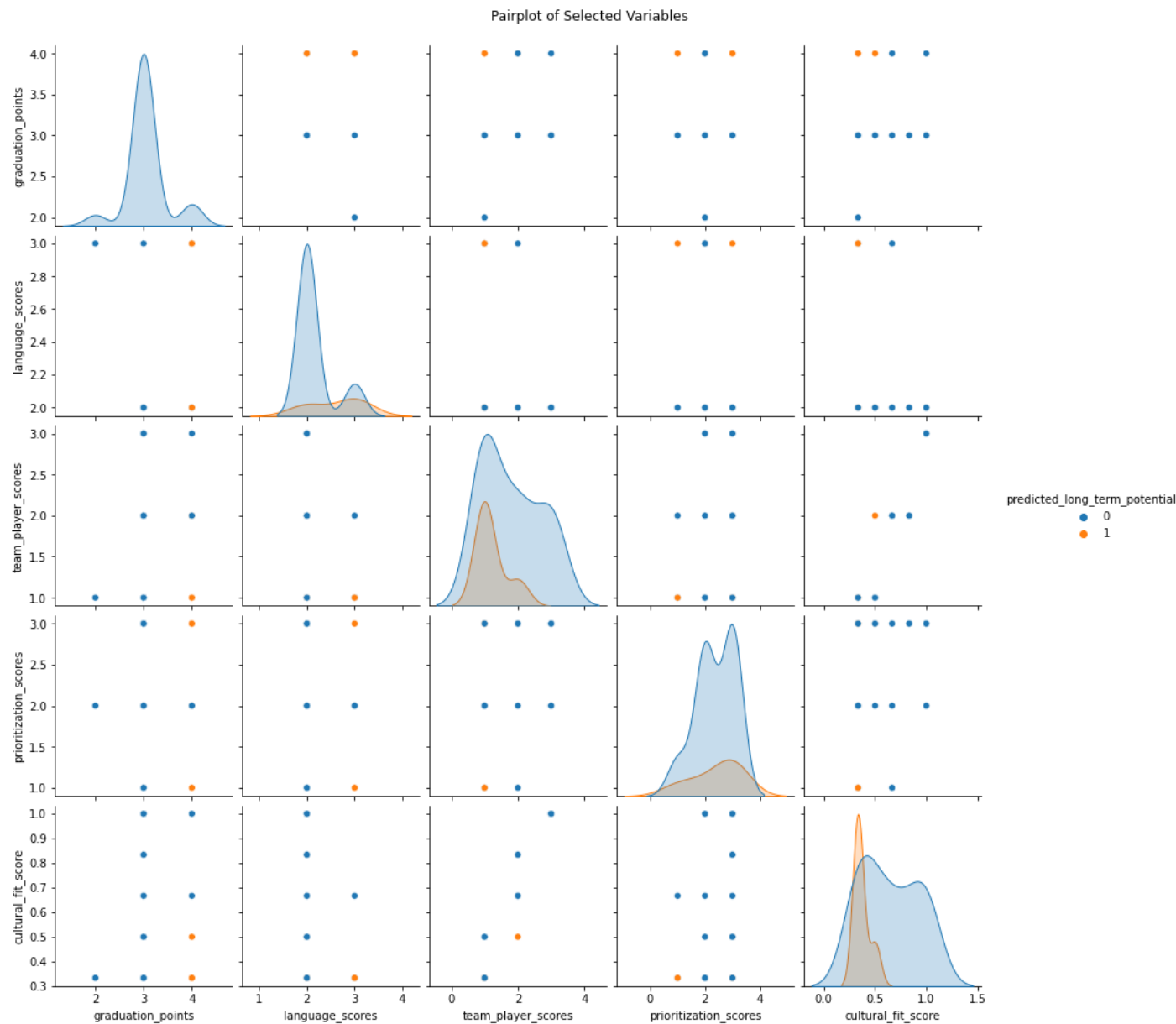


## Pairplot of Selected Variables

In [6]:

```
import seaborn as sns
import matplotlib.pyplot as plt

selected_columns = ['graduation_points', 'language_scores',
                    'team_player_scores', 'prioritization_scores', 'cultural_fit_score']
sns.pairplot(students_df2[selected_columns +
['predicted_long_term_potential']], hue='predicted_long_term_potential')
plt.suptitle('Pairplot of Selected Variables', y=1.02)
plt.show()
```



## PART 2

In [7]:

```
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
from sklearn.linear_model import LogisticRegression
from sklearn.metrics import classification_report, confusion_matrix,
roc_curve, roc_auc_score
import matplotlib.pyplot as plt
import seaborn as sns

# Determine the 80th percentile score to identify the top 20%
threshold = students_df['integration_productivity_score'].quantile(0.80)
```

```
# Create a new column to segment students
students_df['long_term_potential'] =
students_df['integration_productivity_score'].apply(lambda x: 1 if x >=
threshold else 0)

# Feature Engineering
students_df['financial_sensitivity'] = students_df['salary_score'] * 0.7

features = ['community_involvement', 'team_player_scores',
'prioritization_scores',
            'graduation_points', 'salary_score', 'age_points',
'gender_points',
            'location_score', 'financial_sensitivity']
```

```
X = students_df[features]
y = students_df['long_term_potential']
```

```
# Splitting the data into training and test sets
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3,
random_state=42)
```

```
# Feature scaling
scaler = StandardScaler()
X_train = scaler.fit_transform(X_train)
X_test = scaler.transform(X_test)
```

In [9]:

```
from sklearn.metrics import confusion_matrix, roc_curve, roc_auc_score
```

```
# Train the logistic regression model
model = LogisticRegression()
model.fit(X_train, y_train)
```

```
# Predict on test data
y_pred = model.predict(X_test)
```

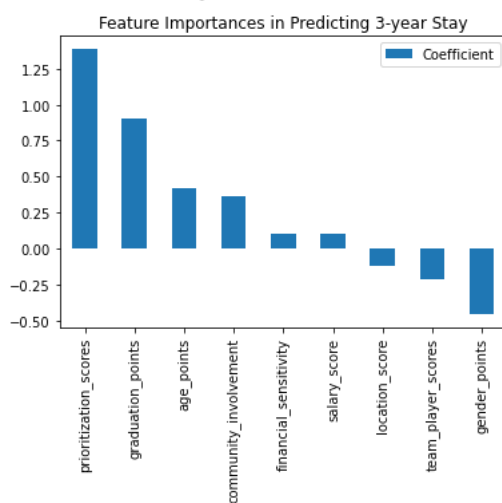
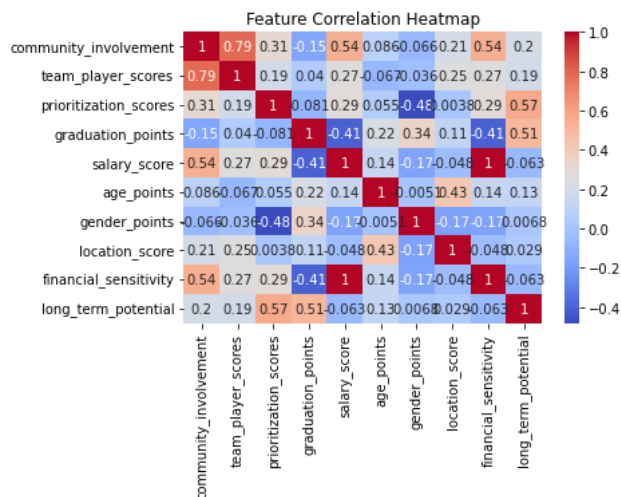
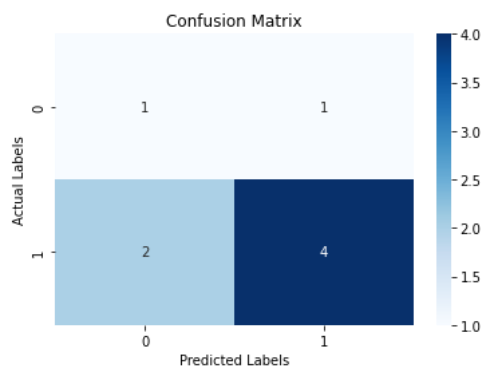
```
# Confusion Matrix
cm = confusion_matrix(y_test, y_pred)
sns.heatmap(cm, annot=True, cmap='Blues')
plt.title('Confusion Matrix')
plt.ylabel('Actual Labels')
plt.xlabel('Predicted Labels')
plt.show()
```

```
# Calculate correlations
corr = students_df[features + ['long_term_potential']].corr() # Changed to
'long_term_potential'
```

```
# Plot heatmap
sns.heatmap(corr, annot=True, cmap='coolwarm')
```

```
plt.title('Feature Correlation Heatmap')
plt.show()
```

```
# Displaying the coefficient of each feature
feature_importance = pd.DataFrame(model.coef_[0], index=features,
columns=['Coefficient'])
feature_importance.sort_values(by='Coefficient', ascending=False,
inplace=True)
feature_importance.plot(kind='bar')
plt.title('Feature Importances in Predicting 3-year Stay')
plt.show()
```



## 3 년 이상 계약 유지 예측 60% 이상 매칭된 지원자들 이름 및 확률

In [8]:

```
X_scaled = scaler.transform(students_df[features])

students_df['predicted_long_term_potential'] = model.predict(X_scaled)

predicted_long_term_students =
students_df[students_df['predicted_long_term_potential'] == 1]['name']

print("Names of Students Predicted to Have Long-Term Potential:")
print(predicted_long_term_students)
Names of Students Predicted to Have Long-Term Potential:
0          Omer Yetiskul
1              Juan
2          Jason Zhai
5      Gabrielle tsuker
7          Sevan Kejejian
8              Amy Kang
9              Kate Yeo
10             lynn
12          Radhika Shah
15             Kevin Hyun
17      Danial Mohammed
19      Mustafa yawary
20      Stephanie Fang
21      Edison Rodriguez
22      Sharona Hatanian
23          Miguel Ruiz
24          Heily Shin
25          Ashli Phan
Name: name, dtype: object
```

In [10]:

```
predicted_probabilities = model.predict_proba(X_scaled)[:, 1]

students_df['predicted_probability'] = predicted_probabilities

predicted_students =
students_df[students_df['predicted_long_term_potential'] == 1][['name',
'predicted_probability']]

predicted_students['predicted_probability'] =
(predicted_students['predicted_probability'] * 68).round(2)

print("Names and Predicted Probabilities (in %) of Students Predicted to
Have Long-Term Potential:")
print(predicted_students)
```

Names and Predicted Probabilities (in %) of Students Predicted to Have Long-Term Potential:

	name	predicted_probability
0	Omer Yetiskul	63.47
1	Juan	55.18
2	Jason Zhai	64.76
5	Gabrielle tsuker	57.35
7	Sevan Kejejian	64.45
8	Amy Kang	47.59
9	Kate Yeo	57.92
10	lynn	58.49
12	Radhika Shah	59.31
15	Kevin Hyun	67.15
17	Danial Mohammed	63.47
19	Mustafa yawary	62.68
20	Stephanie Fang	65.59
21	Edison Rodriguez	66.27
22	Sharona Hatanian	62.74
23	Miguel Ruiz	64.90
24	Heily Shin	35.49
25	Ashli Phan	57.67

In []:

In []: