

## Project Development Phase Model Performance Test

Date	19 November 2022
Team ID	PNT2022TMID51972
Project Name	Project - University Admit Eligibility Predictor
Maximum Marks	

### Model Performance Testing:

S.No	Parameter	Values
1.	Metrics	<b>Regression Model:</b> MAE – 0.0476 MSE – 0.0046 RMSE – 0.0682 R2 score – 0.78
2.	Tune the Model	Hyper-parameter Tuning – GridSearchCV and RandomizedSearchCV  Validation Method – Cross Validation with Ridge and Lasso
3.	Testing	Testing model: Total Request per Second, Response time, No of users
4.	Locus Testing Report	Request Statistics, Response Time Statistics, Final Ratio

### 1. Metrics :

```
In [55]: from math import sqrt
from sklearn.metrics import mean_absolute_error, mean_squared_error

print("Mean Absolute Error :", mean_absolute_error(y_test, y_pred))
print("Mean Squared Error :", mean_squared_error(y_test, y_pred))
print("Rooted Mean Squared Error :", sqrt(mean_squared_error(y_test, y_pred)))
print("R2 Score :", r2_score(y_test, y_pred))
```

```
Mean Absolute Error : 0.04766740707308981
Mean Squared Error : 0.004658892249358201
Rooted Mean Squared Error : 0.06825607847919628
R2 Score : 0.7777563100798979
```

## 2. Tune the Model :

### Hyperparameter Tuning

```
In [45]: def evaluate(models):
          results = {}
          for i, j in models.items():
              results[i] = [j.best_params_, j.best_estimator_, j.best_score_]
          return pd.DataFrame(results, index=['Best Parameter', 'Best Estimator', 'Best Score'])
```

```
In [46]: #Using GridSearchCV
          from sklearn.model_selection import GridSearchCV

          lasso_params = {'alpha':[0.002, 0.00024, 0.00025, 0.0026, 0.03]}
          ridge_params = {'alpha':[0.002, 0.0024, 0.0025, 0.0026, 0.03, 0.04]}

          lsgs = GridSearchCV(Lasso(), param_grid=lasso_params, cv=5)
          rdgs = GridSearchCV(Ridge(), param_grid=ridge_params, cv=5)

          models2 = {'OLS': LinearRegression(),
                     'Lasso': lsgs.fit(X, y).best_estimator_,
                     'Ridge': rdgs.fit(X, y).best_estimator_,}

          test(models2, X, y)
```

```
Out[46]:
```

	OLS	Lasso	Ridge
Training Results	0.796207	0.794771	0.794814
Testing Results	0.774031	0.783381	0.782087

```
In [47]: cv = {'Lasso': lsgs,
               'Ridge': rdgs}
          evaluate(cv)
```

```
Out[47]:
```

	Lasso	Ridge
Best Parameter	{'alpha': 0.00025}	{'alpha': 0.002}
Best Estimator	Lasso(alpha=0.00025)	Ridge(alpha=0.002)
Best Score	0.759762	0.759304

```
In [48]: #Using RandomizedSearchCV
          from sklearn.model_selection import RandomizedSearchCV

          lsrs = RandomizedSearchCV(estimator=Lasso(), param_distributions=lasso_params, cv = 3, n_iter = 5)
          rdrs = RandomizedSearchCV(estimator=Ridge(), param_distributions=ridge_params, cv = 3, n_iter = 5)

          models3 = {'OLS': LinearRegression(),
                     'Lasso': lsrs.fit(X, y).best_estimator_,
                     'Ridge': rdrs.fit(X, y).best_estimator_,}

          test(models3, X, y)
```

```
Out[48]:
```

	OLS	Lasso	Ridge
Training Results	0.796065	0.796405	0.794866
Testing Results	0.777465	0.776379	0.783031

```
In [49]: cv = {'Lasso': lsrs,  
             'Ridge': rdrrs}  
evaluate(cv)
```

Out[49]:

	Lasso	Ridge
<b>Best Parameter</b>	{'alpha': 0.00024}	{'alpha': 0.002}
<b>Best Estimator</b>	Lasso(alpha=0.00024)	Ridge(alpha=0.002)
<b>Best Score</b>	0.721576	0.721907

## LassoCV and RidgeCV Validation

```
In [50]: models4 = {'LassoCV': LassoCV(alphas=lasso_params['alpha']),  
                  'RidgeCV': RidgeCV(alphas=ridge_params['alpha'])}  
test(models4, X, y)
```

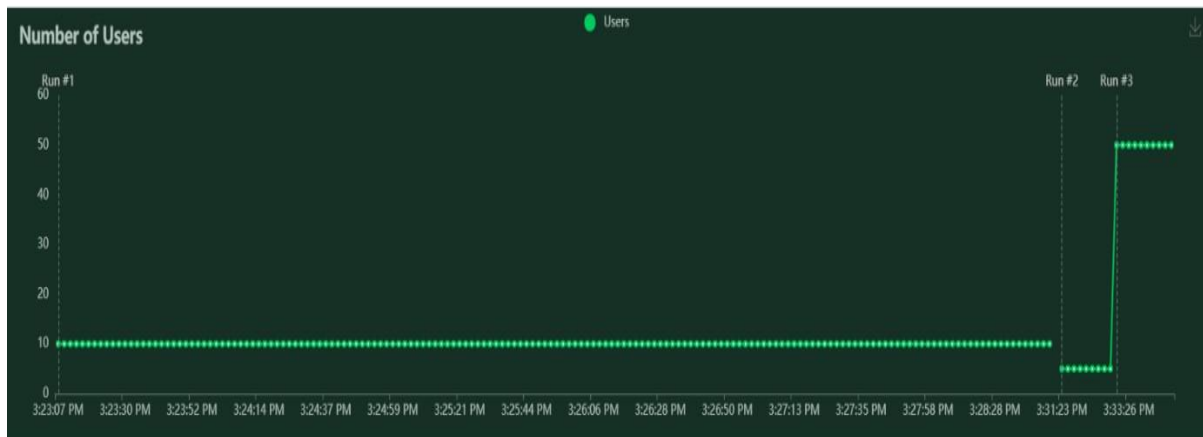
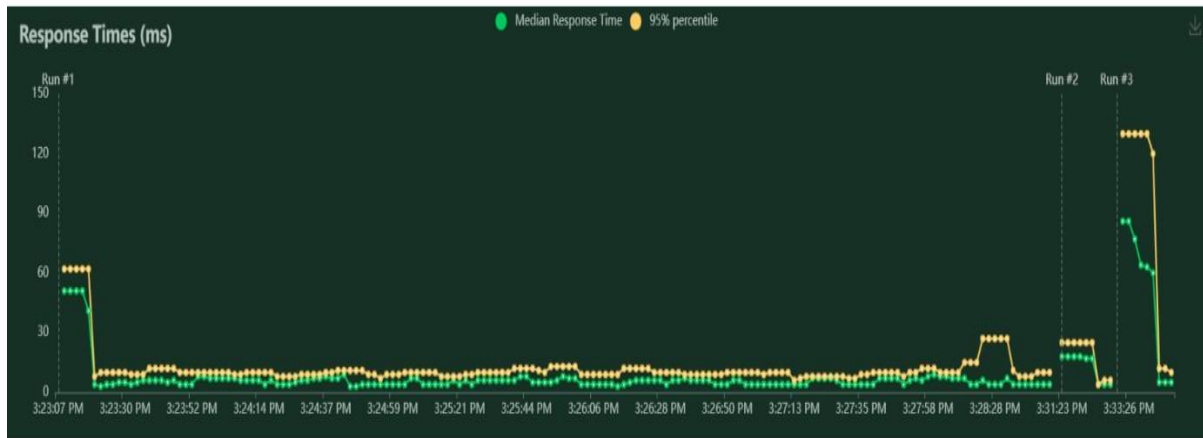
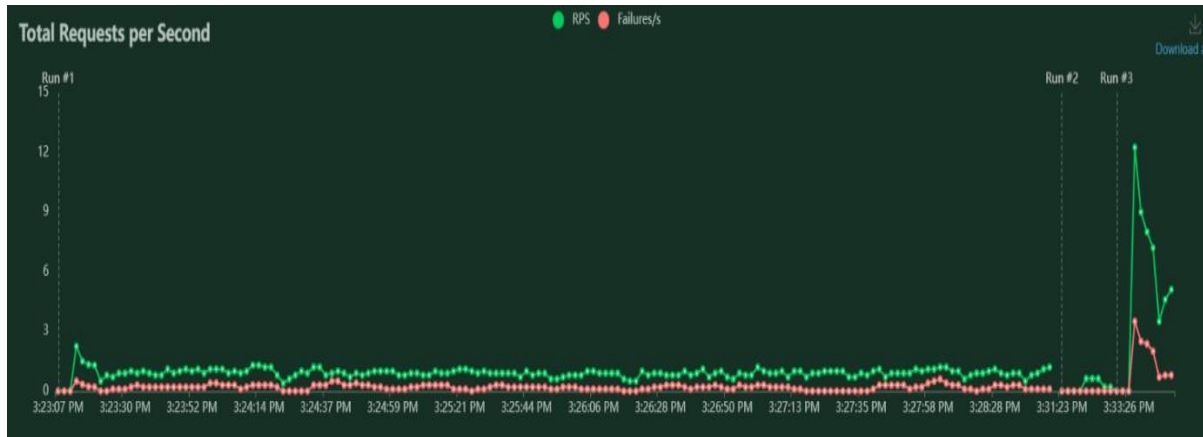
Out[50]:

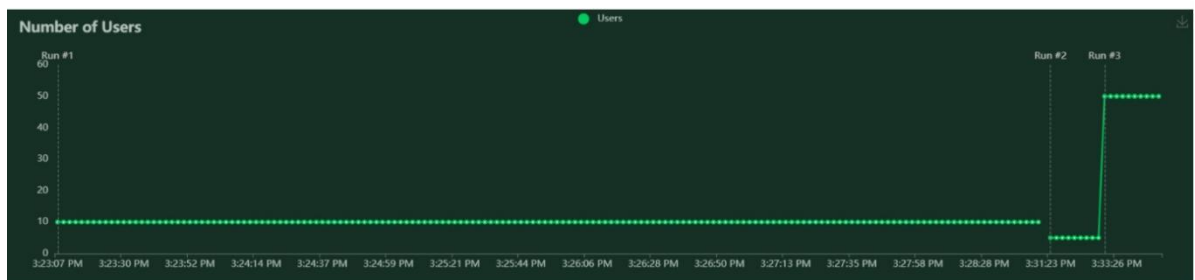
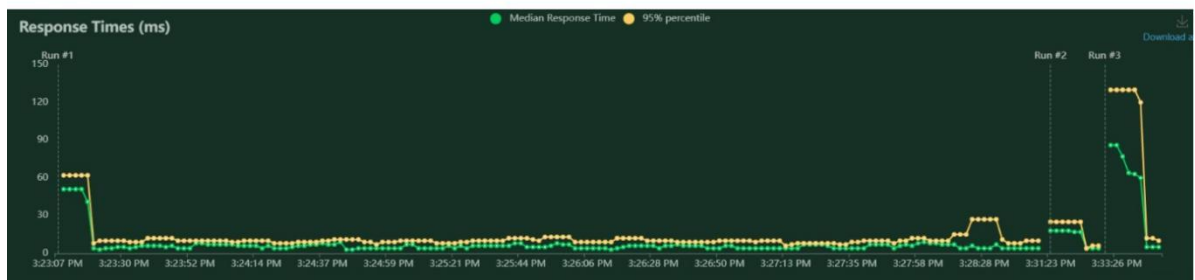
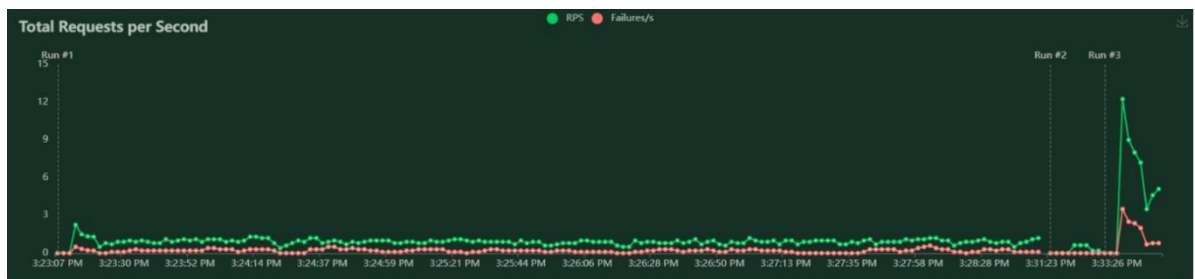
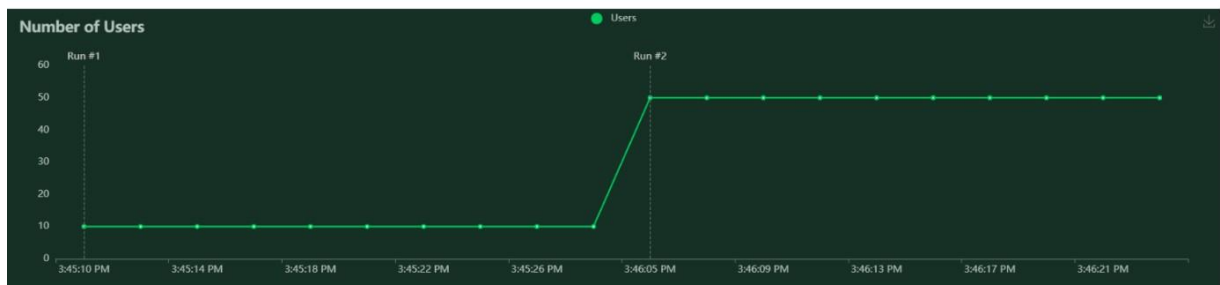
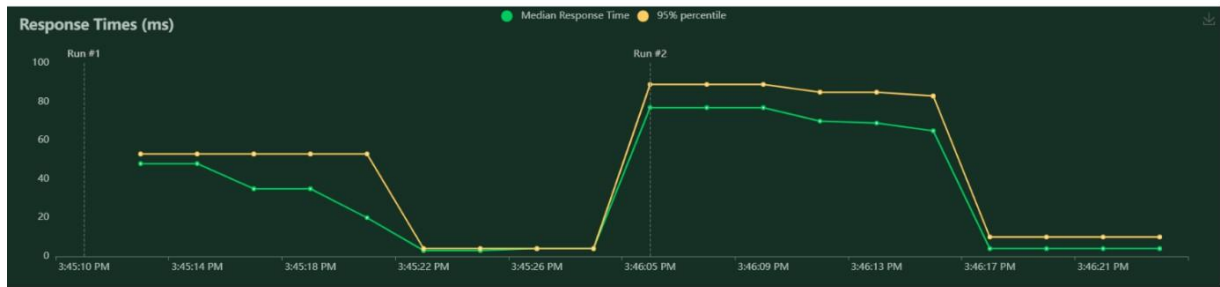
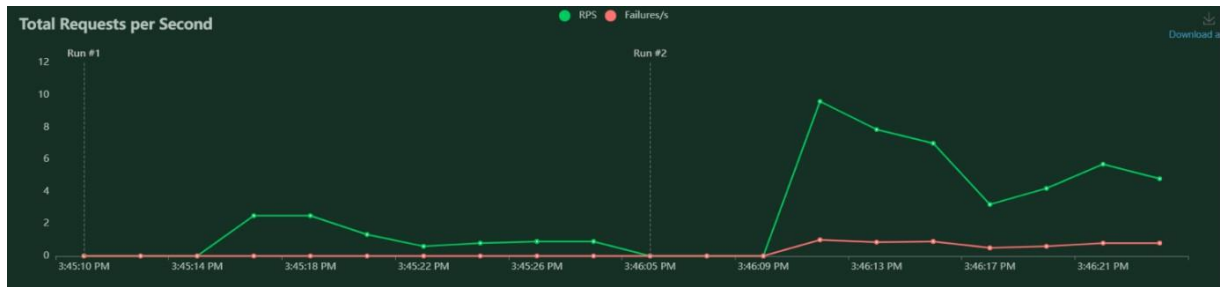
	LassoCV	RidgeCV
<b>Training Results</b>	0.794466	0.795869
<b>Testing Results</b>	0.780977	0.776775

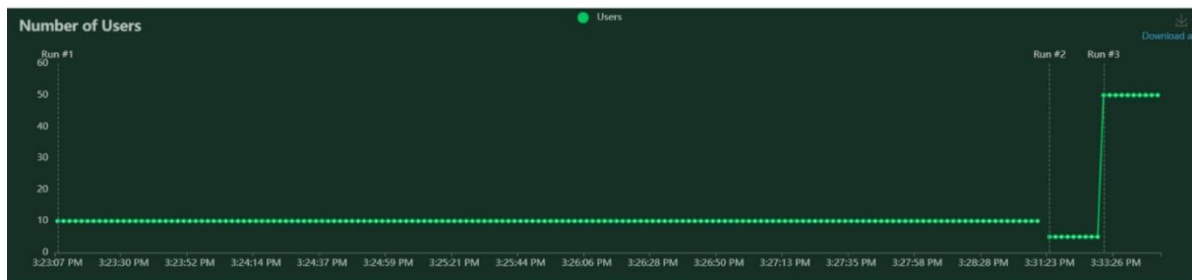
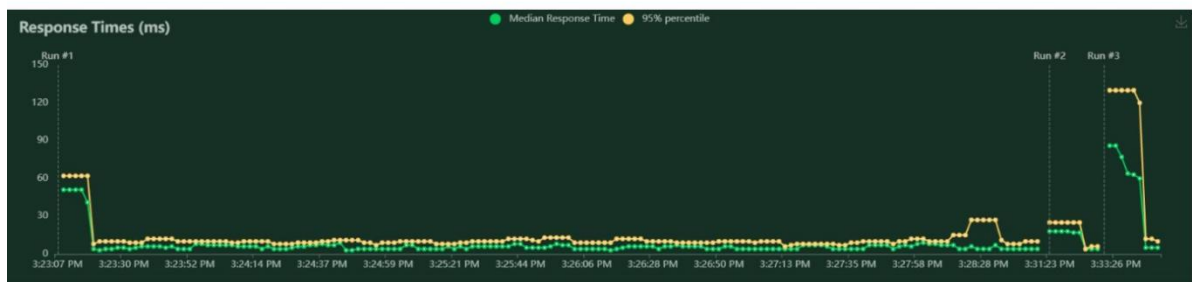
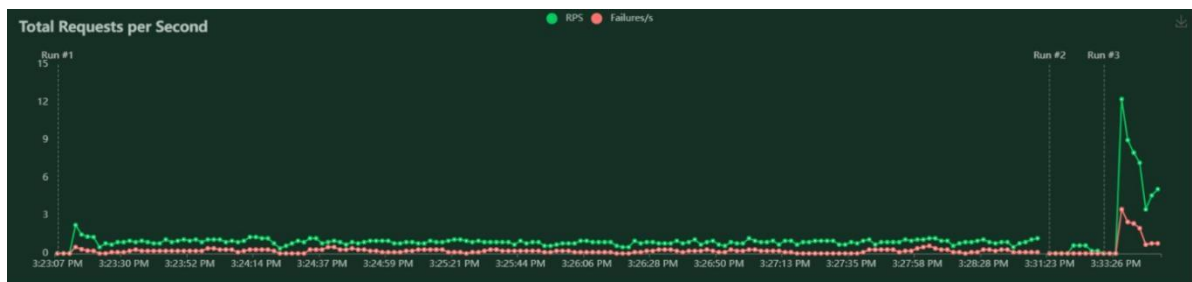
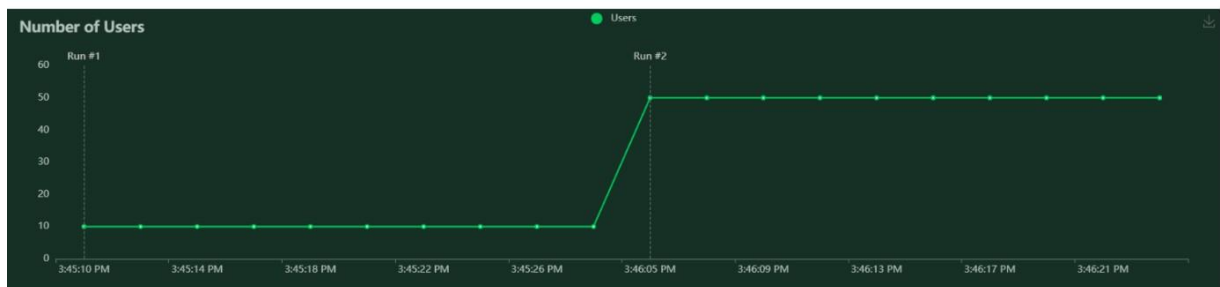
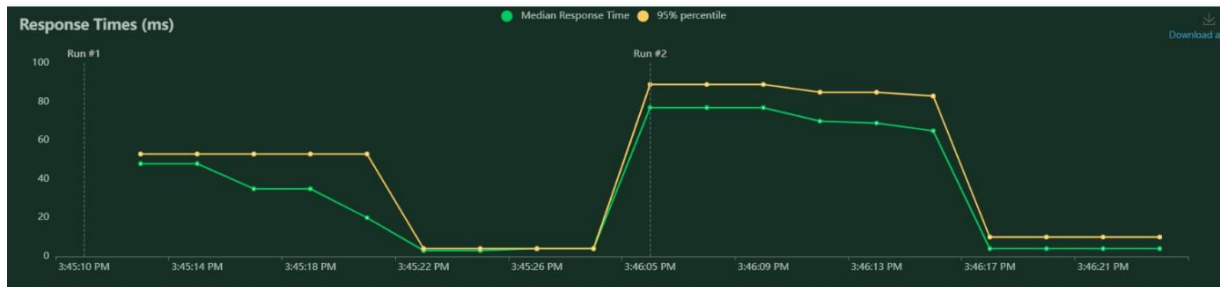
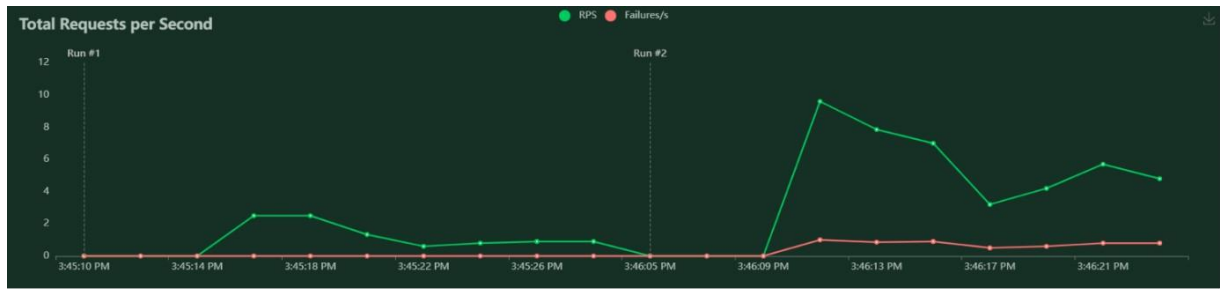
## Build, Train and Test the Best Model

```
In [51]: ridge_cv = RidgeCV(alphas=ridge_params['alpha'])  
  
ridge_cv.fit(X_train, y_train)  
  
y_pred_train = ridge_cv.predict(X_train)  
y_pred = ridge_cv.predict(X_test)
```

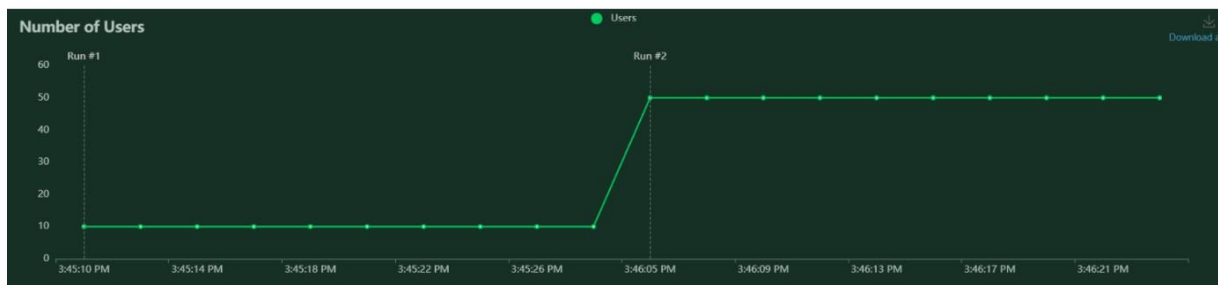
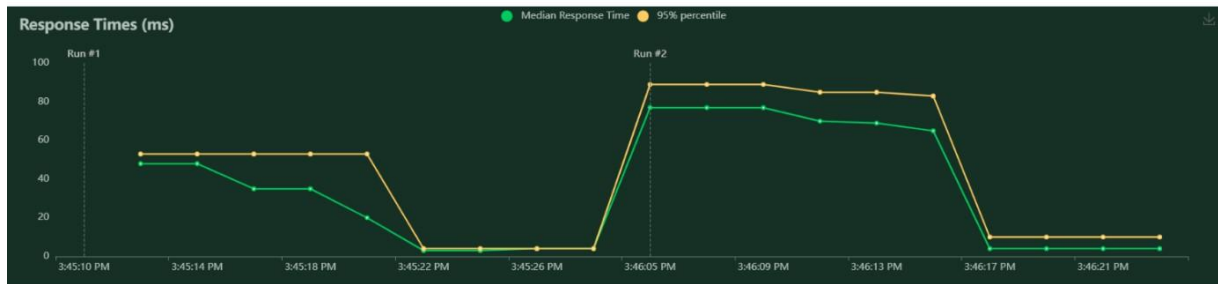
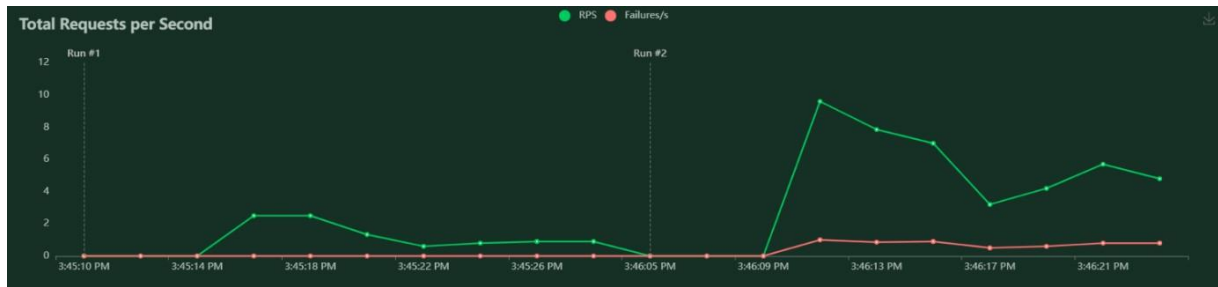
### 3. Testing:











**LOCUST**

HOST: <http://127.0.0.1:5000/>

STATUS: **STOPPED**  
New test

RPS: **5.1**

FAILURES: **21%**

Statistics Charts Failures Exceptions Current ratio Download Data

Type	Name	# Requests	# Fails	Median (ms)	90%ile (ms)	99%ile (ms)	Average (ms)	Min (ms)	Max (ms)	Average size (bytes)	Current RPS	Current Failures/s
GET	/	25	0	8	130	130	36	6	130	3919	1.7	0
GET	/home	34	0	4	99	110	33	3	110	2959	1.4	0
GET	/register	26	26	63	110	110	55	8	115	265	0.6	0.6
GET	/university	37	0	4	93	110	33	3	109	19010	1.4	0
Aggregated		122	26	9	110	130	38	3	130	7450	5.1	0.6

**LOCUST**

HOST: [http://127.0.0.1:5000](http://127.0.0.1:5000/)

STATUS: **STOPPED**  
New test

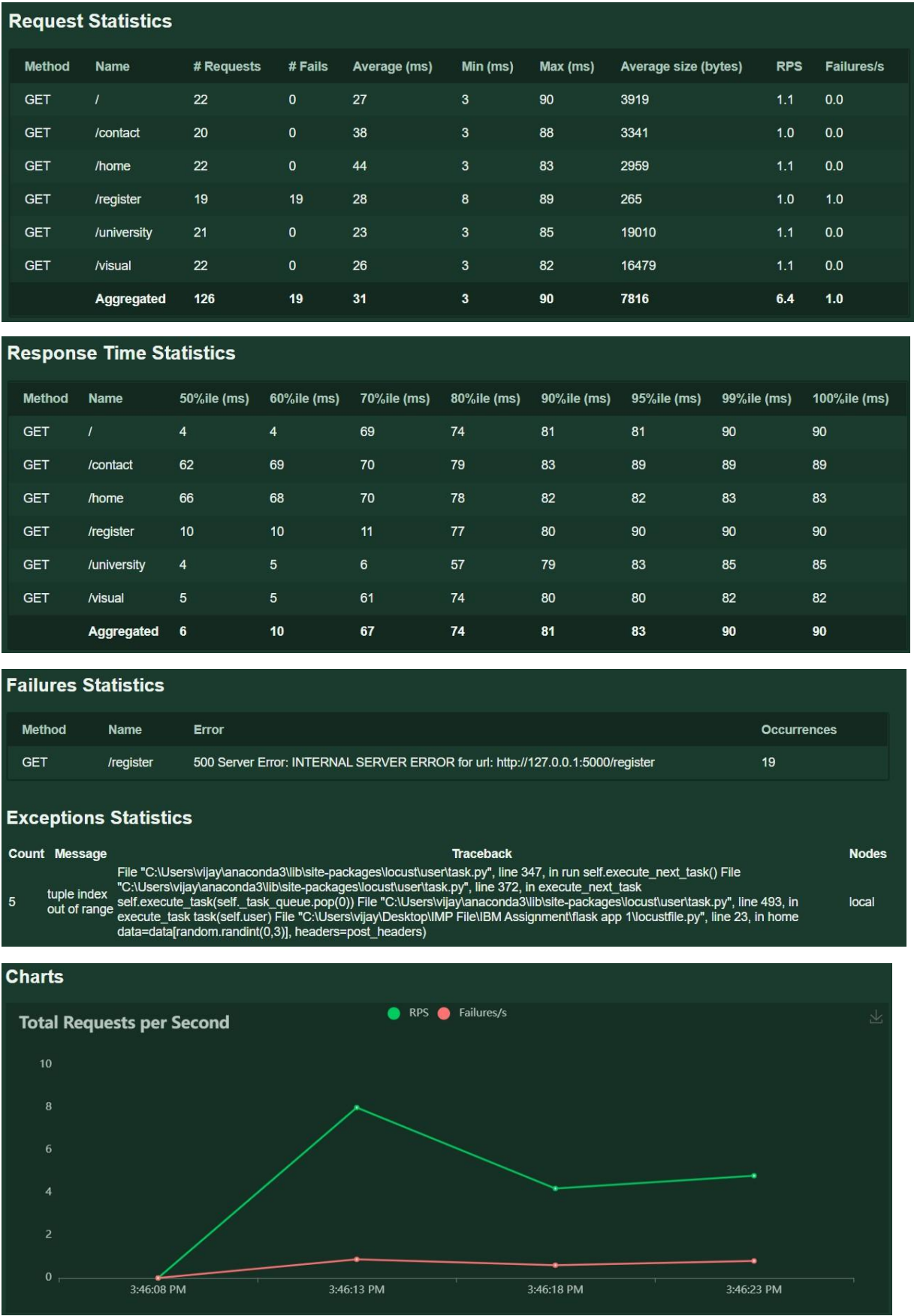
RPS: **4.8**

FAILURES: **15%**

Statistics Charts Failures Exceptions Current ratio Download Data

Type	Name	# Requests	# Fails	Median (ms)	90%ile (ms)	99%ile (ms)	Average (ms)	Min (ms)	Max (ms)	Average size (bytes)	Current RPS	Current Failures/s
GET	/	22	0	4	81	90	27	3	90	3919	0.7	0
GET	/contact	20	0	4	83	89	39	3	89	3341	0.4	0
GET	/home	22	0	65	82	83	44	3	83	2959	0.6	0
GET	/register	19	19	10	80	90	28	9	90	265	0.8	0.8
GET	/university	21	0	4	79	85	23	3	85	19010	1.1	0
GET	/visual	22	0	4	80	82	26	3	82	16479	1.2	0
Aggregated		126	19	6	81	90	31	3	90	7817	4.8	0.8

4. Locus Testing Report:







## Final ratio

### Ratio per User class

- 100.0% WebsiteUser
  - 16.7% index
  - 16.7% home
  - 16.7% home1
  - 16.7% visual
  - 16.7% visual1
  - 16.7% visual2

### Total ratio

- 100.0% WebsiteUser
  - 16.7% index
  - 16.7% home
  - 16.7% home1
  - 16.7% visual
  - 16.7% visual1
  - 16.7% visual2