# Auto-Encoder Reconstruction Cost for Anomalies Detection scenarios

Comparing Auto-Encoder approach for Anomalies Detection to traditional Upsampling + Classification techniques

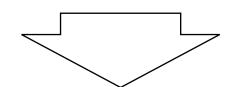
### Dataset

#### **Credit Card Fraud Detection** Anonymized credit card transactions labeled as fraudulent or genuine

https://www.kaggle.com/mlg-ulb/creditcardfraud

#### 2 days real European credit cards transactions: 492 frauds out of 284,807 (0.172%)

- 28 Numerical input variables which are the result of a PCA transformation to anonymize customer info
- Transaction Amount
- Time in seconds from T0
- Label Class: 0/1 (1 is the anomaly)

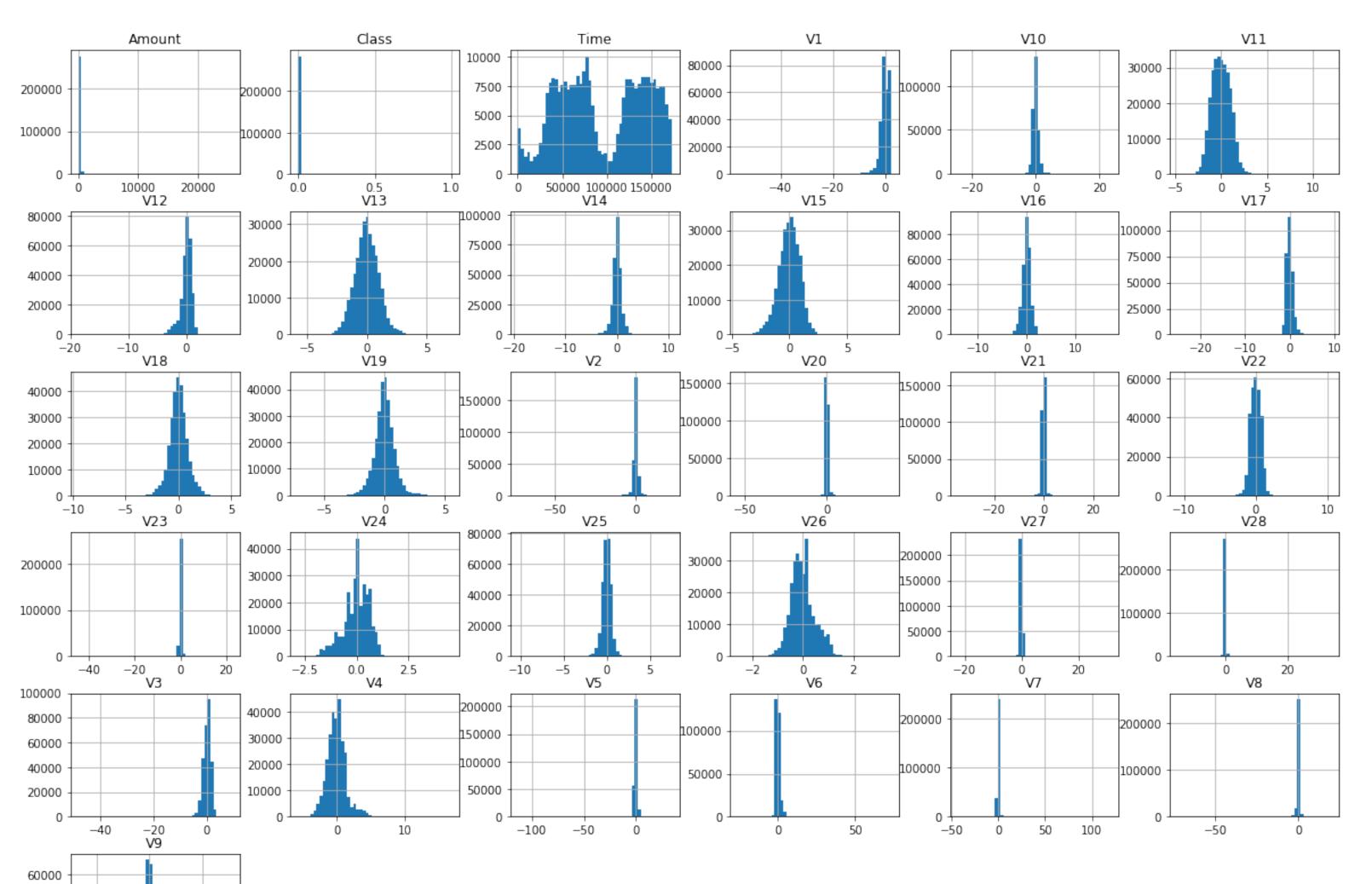


- 29 Numerical Features
  - Original 28 from PCA
- Transaction Amount (StandardScaler)
- 1 Categorical Feature
- Time of the transaction using 24 Hour beens



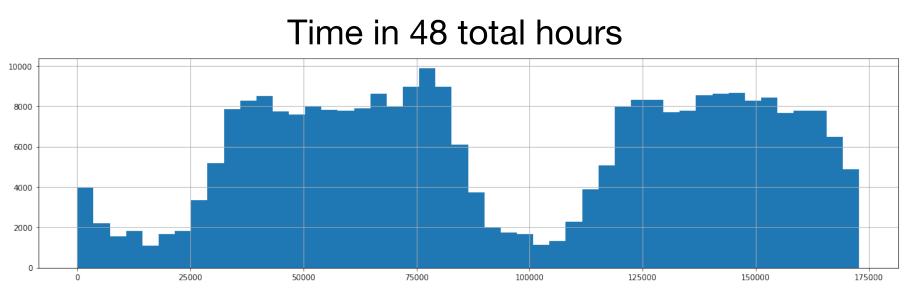
- 29 Numerical Features -> Tensor Length 29
- Categorical Feature values 0-24 —> Embedding Tensor 12

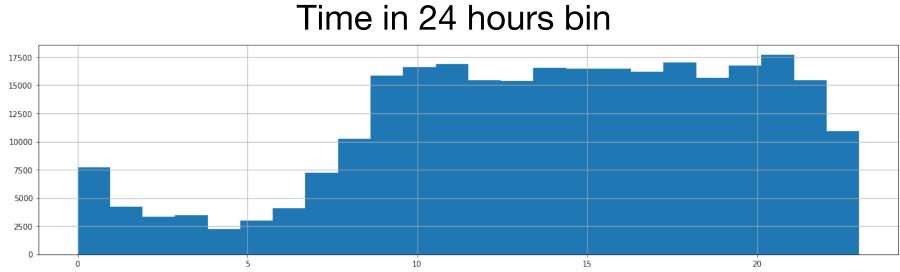
## Plot



40000 -

20000 -





# Training Approach

#### Classifier

- Feed 80% of both anomaly (+) and non-anomaly (-) transactions
- Upscale the anomaly transactions in the train dataset
- Handle Multi-Input for Numeric features and Categories
- Use Embedding for Categories
- Build a classic DNN classifier

#### **Auto-Encoder**

- Feed 80% of only the non-anomaly (-) transactions
- Handle Multi-Input for Numeric features and Categories
- Use Embedding for Categories
- Use an Auto-Encoder to learn compression/ decompression on non-anomaly transactions
- Use a custom layer to calculate the reconstruction cost
- Train to reduce this cost to 0

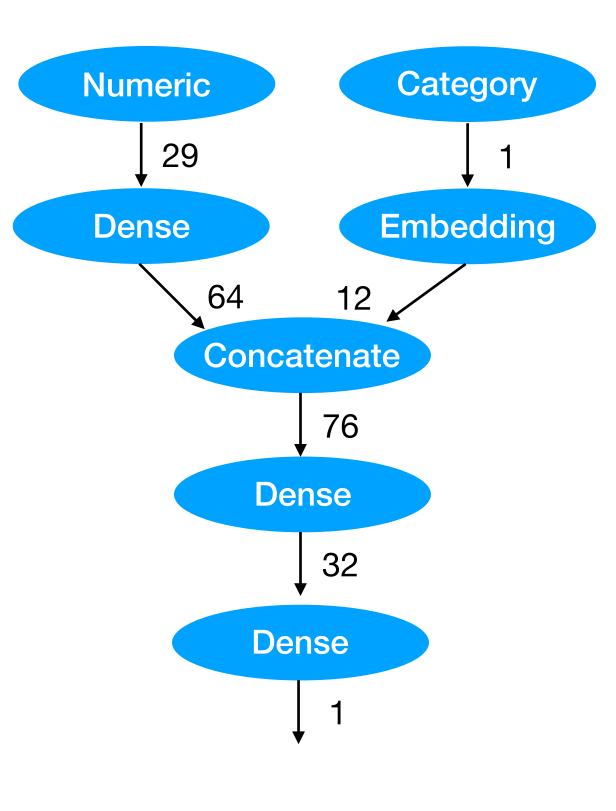


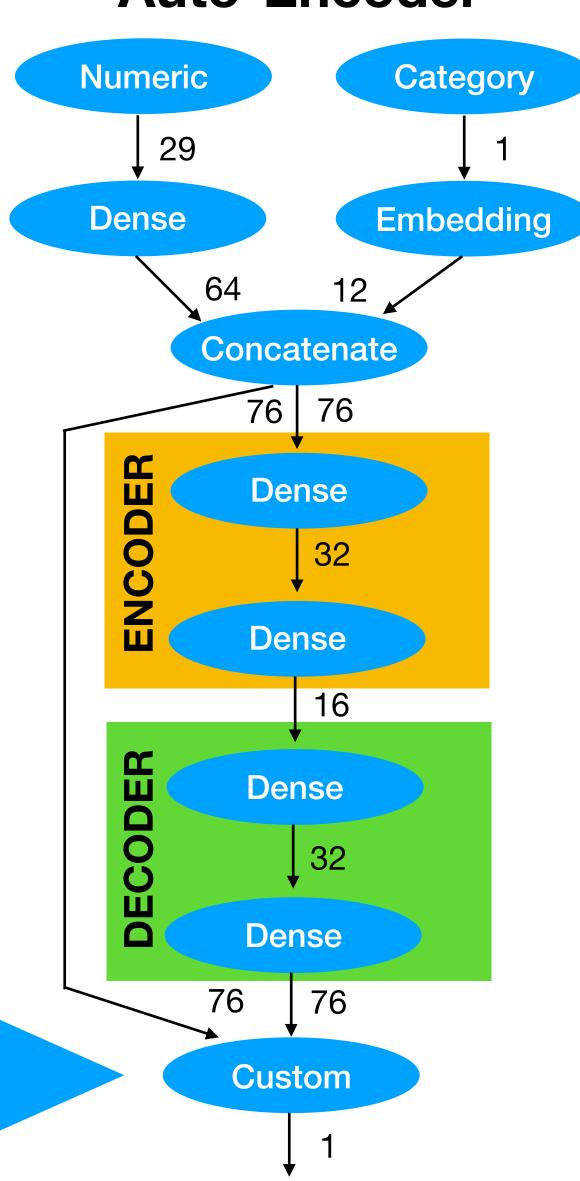
#### Classifier





#### **Auto-Encoder**





**Reconstruction Cost** \( \sum\_{\text{Encoder\_Input}} \) (Encoder\_Input - Decoder\_Output)<sup>2</sup>

#### Classifier

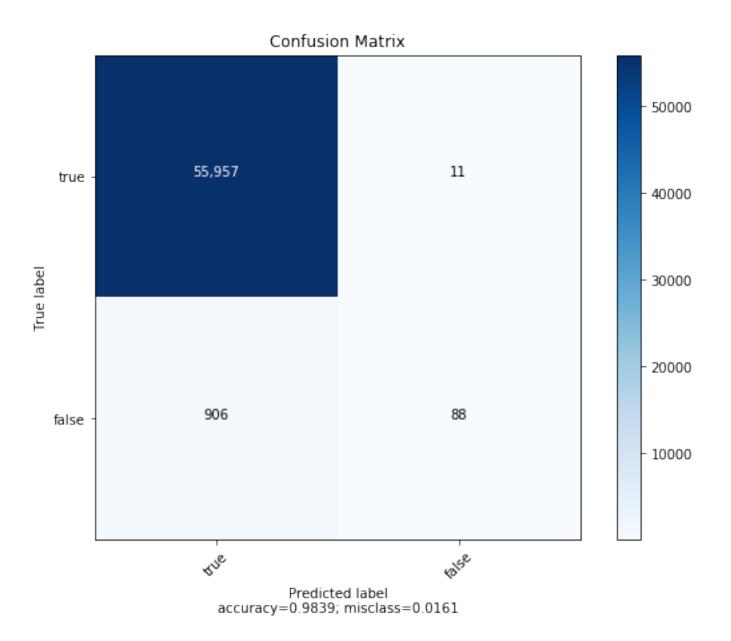
accuracy: 0.9839015484006882

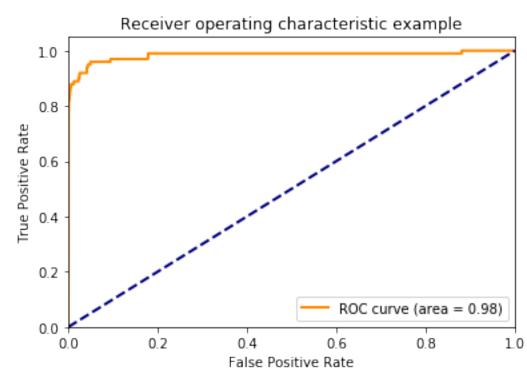
precision: [0.99980346 0.08853119]

recall: [0.98406697 0.88888889]

fscore: [0.9918728 0.1610247]

support: [56863 99]





### Results

Feed 20% + -

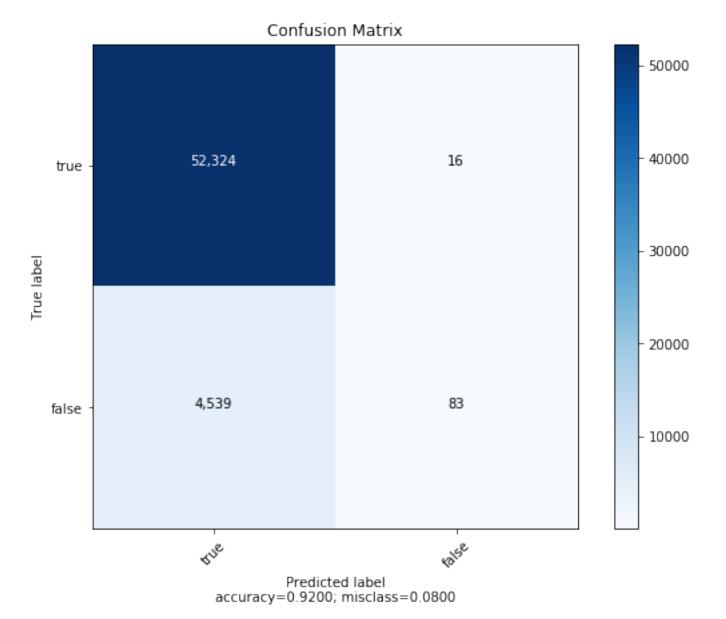
# Auto-Encoder (THRESHOLD = 0.17)

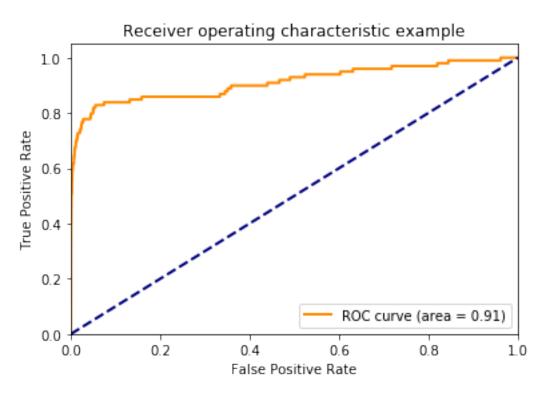
accuracy: 0.9200344089041818

precision: [0.99969431 0.01795759]
recall: [0.92017656 0.83838384]

fscore: [0.95828869 0.03516204]

support: [56863 99]





# Interactivity test

THRESHOLD -

0.17

Feed 80%

Feed 20%

SCORING

(-)

SCORING

TRAINED NEGATIVE

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Greater Than Threeshold: 13932 over 227452

Accuracy: 0.94

Mean: 0.05370073765516281 Std: 0.09088834375143051 Min: 5.205785419093445e-05

Max: 3.160778045654297

NEW NEGATIVE

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Greater Than Threeshold: 4809 over 56863

Accuracy: 0.92

Mean: 0.07014649361371994 Std: 0.1701817512512207 Min: 8.863704715622589e-05 Max: 26.871849060058594

NEW POSITIVE

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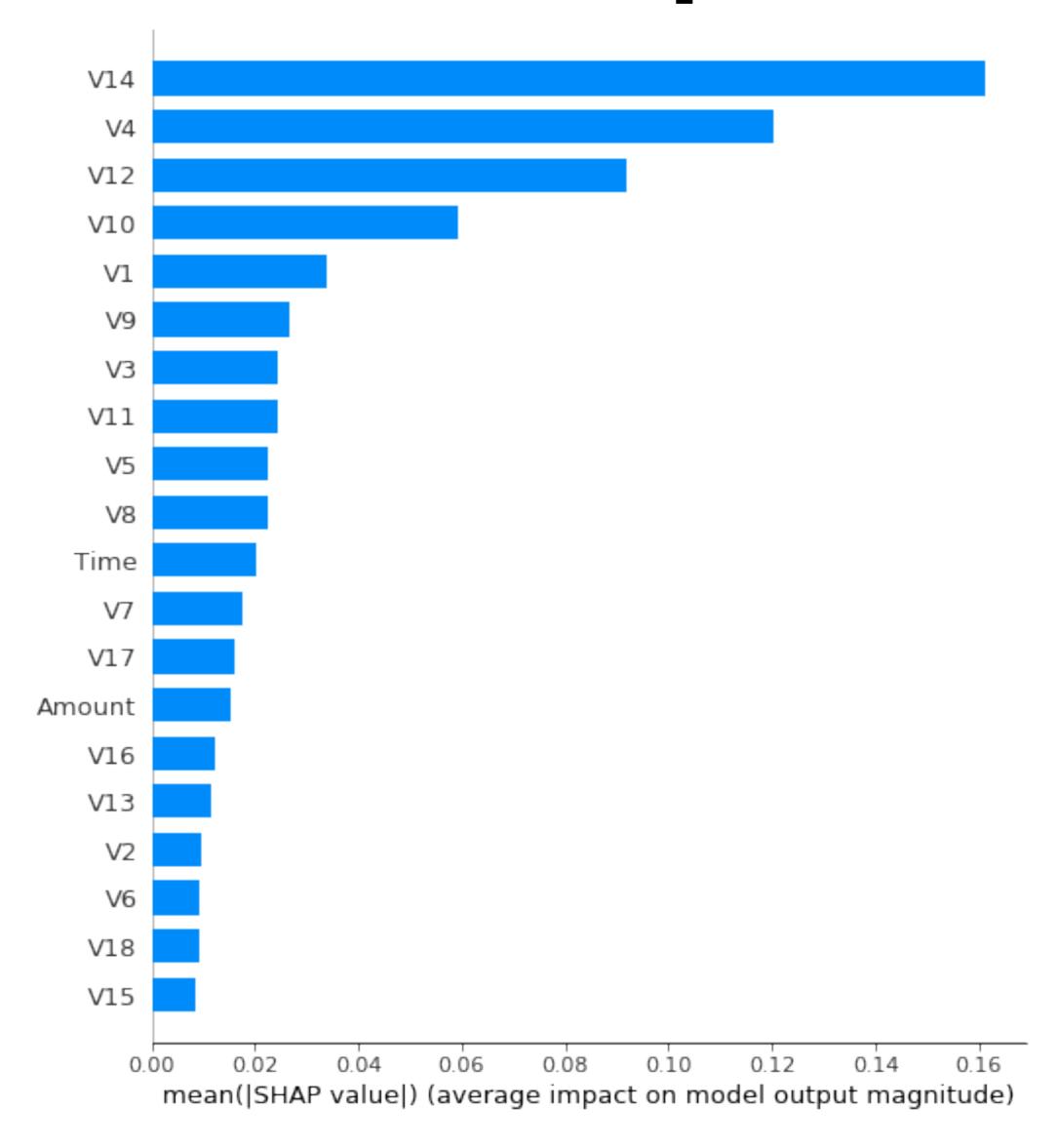
Greater Than Threeshold: 416 over 492

Accuracy: 0.85

Mean: 3.5444180965423584 Std: 7.813464164733887 Min: 0.001192888943478465 Max: 55.32634735107422

Feed 100% (+)
SCORING

# Features Interpretability



### Source Code

https://github.com/JacopoMangiavacchi/AutoEncoderForAnomalyDetection