



Introduction to fraud detection

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Meet your instructor



Hi my name is Charlotte and I am a Data Scientist

What is Fraud?

Examples of fraud: insurance fraud, credit card fraud, identify theft, money laundering, tax evasion, product warranty, healthcare fraud

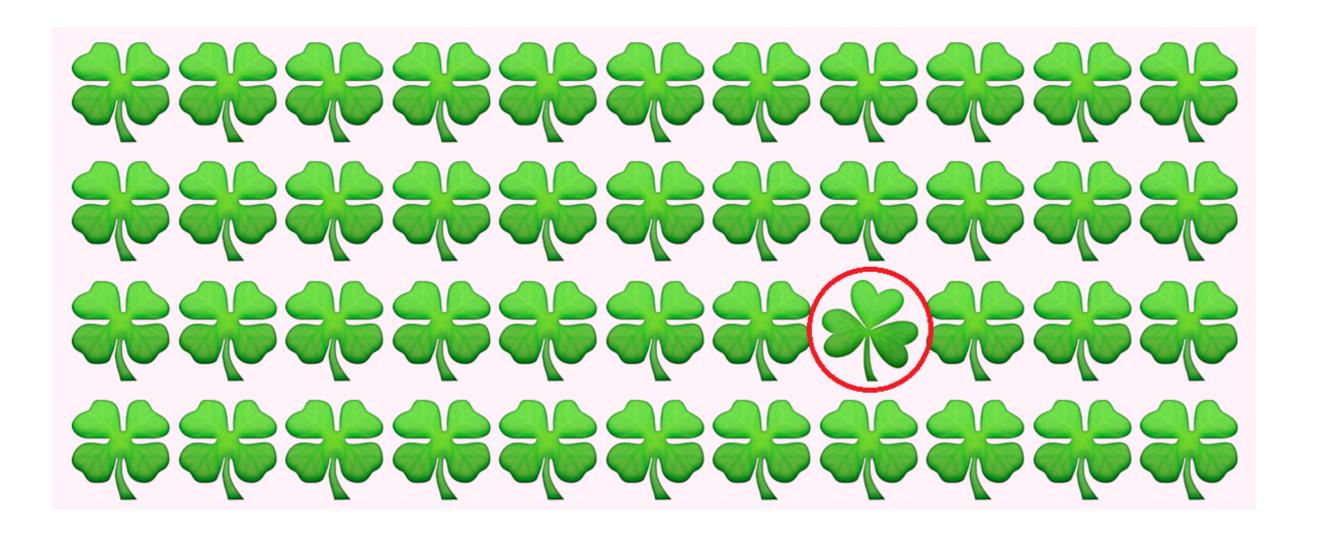
Fraud is

- uncommon
- concealed
- changing over time
- organized











16	80	44	12
24	96	20	32
8	28	36	26
40	56	68	4



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How companies deal with fraud

Fraud analytics teams:

- 1. Often use rules based systems, based on manually set thresholds and experience
- 2. Check the news
- 3. Receive external lists of fraudulent accounts and names
- 4. Sometimes use machine learning algorithms to detect fraud or suspicious behaviour



Let's have a look at some data

```
df.shape
(5050, 30)
```





Let's practice!



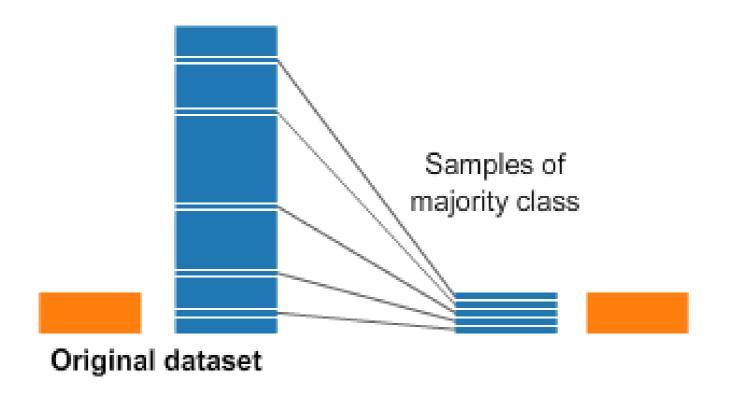


Increasing succesfull detections using data resampling

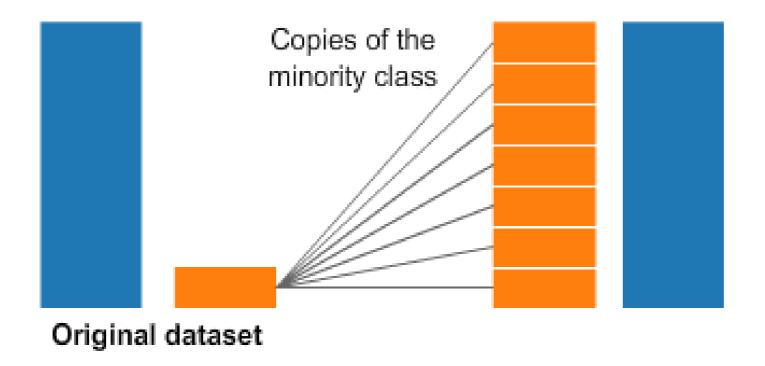
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Undersampling



Oversampling

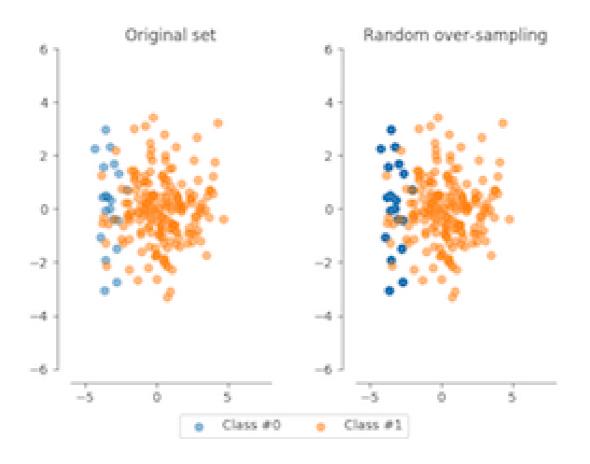


Oversampling in Python

```
from imblearn.over_sampling import RandomOverSampler

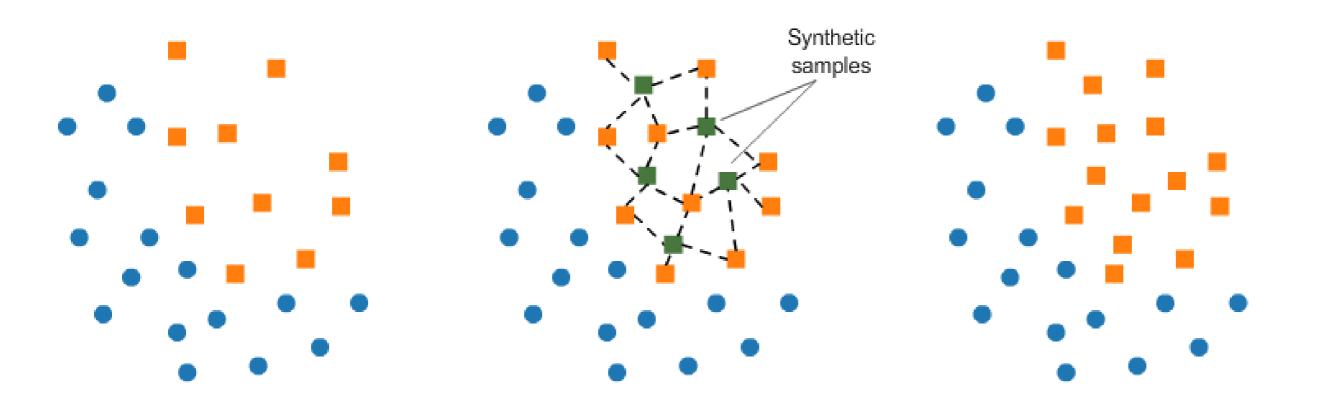
method = RandomOverSampler()
X_resampled, y_resampled = method.fit_sample(X, y)

compare_plots(X_resampled, y_resampled, X, y)
```





Synthetic Minority Oversampling Technique (SMOTE)



Source: https://www.kaggle.com/rafjaa/resampling-strategies-for-imbalanced-datasets



Which resampling method to use?

- Random Under Sampling (RUS): throw away data, computationally efficient
- Random Over Sampling (ROS): straightforward and simple, but training your model on many duplicates
- Synthetic Minority Oversampling Technique (SMOTE): more sophisticated and realistic dataset, but you are training on "fake" data



When to use resampling methods

Use resampling methods on your training set, never on your test set!

```
# Define resampling method and split into train and test
method = SMOTE(kind='borderline1')
X_train, X_test, y_train, y_test = train_test_split(X, y,
    train_size=0.8, random_state=0)

# Apply resampling to the training data only
X_resampled, y_resampled = method.fit_sample(X_train, y_train)

# Continue fitting the model and obtain predictions
model = LogisticRegression()
model.fit(X_resampled, y_resampled)

# Get your performance metrics
predicted = model.predict(X_test)
print (classification_report(y_test, predicted))
```





Let's practice!





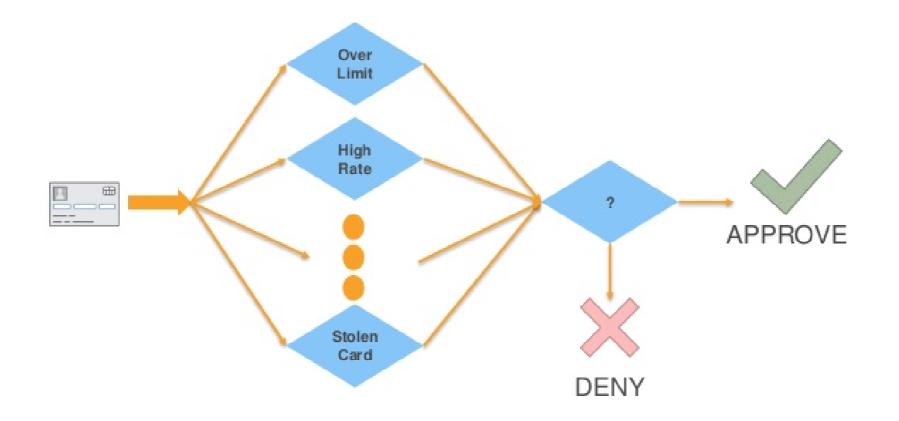
Fraud detection algorithms in action

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Traditional fraud detection with rules based systems

Rule-Based Fraud Detection





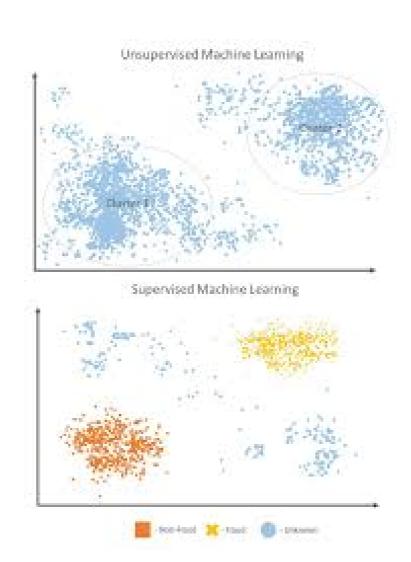
Drawbacks of using rules based systems

Rules based systems have their limitations:

- 1. Fixed thresholds per rule to determine fraud
- 2. Limited to yes/no outcomes
- 3. Fail to capture interaction between features

Why use machine learning for fraud detection?

- Machine learning models adapt to the data, and thus can change over time
- 2. Uses all the data combined rather than a threshold per feature
- 3. Can give a score, rather than a yes/no
- 4. Will typically have a better performance and can be combined with rules



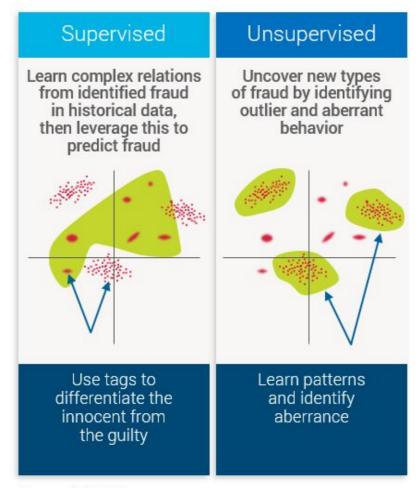


Refresher on machine learning models

```
from sklearn.linear model import LinearRegression
from sklearn.model selection import train test split
from sklearn import metrics
# Step 1: split your features and labels into train and test data
X train, X test, y train, y test = train test split(X, y, test size=0.2)
# Step 2: Define which model you want to use
model = LinearRegression()
# Step 3: Fit the model to your training data
model.fit(X train, y train)
# Step 4: Obtain model predictions from your test data
y predicted = model.predict(X test)
# Step 5: Compare y test to predictions and obtain performance metrics
print (metrics.r2 score(y test, y predicted))
0.821206237313
```

What you'll be doing in the upcoming chapters

- Chapter 2. Supervised learning: train a model using existing fraud labels
- Chapter 3. Unsupervised learning: use your data to determine what is 'suspicious' behaviour without labels
- Chapter 4. Fraud detection using text data: Learn how to augment your fraud detection models with text mining and topic modelling



Source: FICO Blog





Let's practice!