



## Analysis of Credit Default Data

Data Science Framework

Rory S. Langran, Credit One LLC

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## BACKGROUND AND GOALS

Project Context and Desired Outcomes

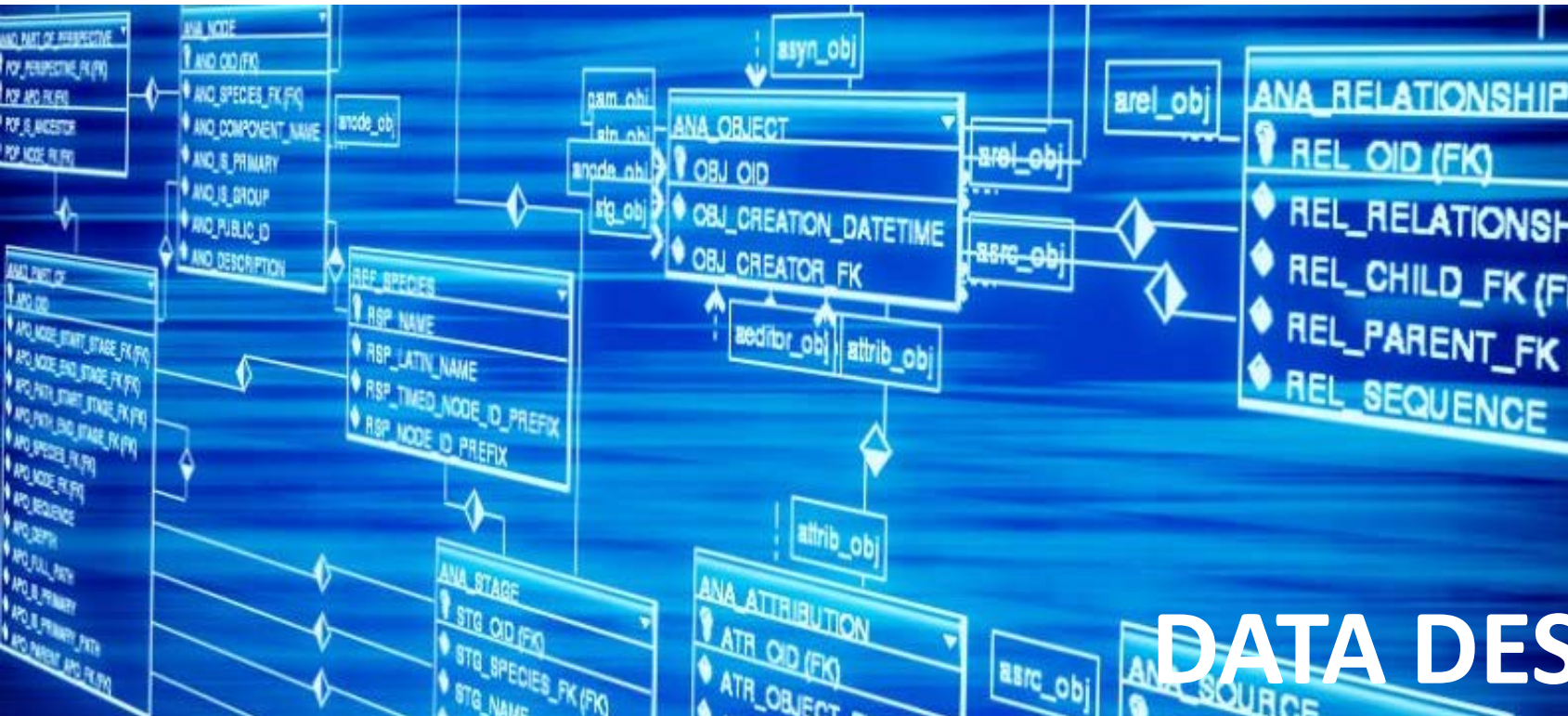
# Project Background

- Over the past year, there has been an increase in the number of customers who have defaulted on loans from various partners
- Credit One, as their credit scoring service, could risk losing business if the problem is not solved right away
- Management has engaged with the Data Science team in order to design and implement a creative, empirically sound solution for predicting credit default

# Project Goals

- Conduct an initial exploration of the current data feed
- Define the business problem using a Data Science framework
- Understand how Data Science will be used to create a model that will more accurately predict credit default
- Design and implement a data process flow model





02

## DATA DESCRIPTION

An Overview of the Dataset

# Data Description

**Background:** Customer profile with balance and payment details in \*.csv file format

**Period:** April to September 2005

**Records:** 30,000

**Dependent Variable:** Binary loan default indicator for next month – “1” default or “0” no default

**Independent Variables:** 23 total

**Variable Categories:**

- 1) Customer Profile (5): Credit Limit, Gender, Education Level, Marital Status, Age
- 2) Payment Status (6): Indicator for on-time payment, no payment, or # months late from April to September 2005
- 3) Balance Amount (6): Balance of account for each month from April to September 2005
- 4) Payment Amount (6): Payment for each month from April to September 2005

# Data Attributes Details

## Customer Profile Attributes (5):

- “LIMIT\_BAL” – Integer data type; amount of given credit (NT dollars)
- “SEX” – Factor data type; “1” male / “2” female
- “EDUCATION” – Integer data type; “1” graduate school / “2” university / “3” high school / “0, 4, 5, 6” all other school
- “MARRIAGE” – Integer data type; “1” married / “2” single / “3” divorced / “0” all others
- “AGE” – Integer data type; current age of account holder

## Payment Status Attributes (6):

- “PAY\_0, PAY\_2, PAY\_3, PAY\_4, PAY\_5, PAY\_6” – Payment indicator for six month period from Apr 2005 through Sep 2005. Integer data type; “-2” no payment required / “-1” paid in full / “0” partial payment / “1,2,3,...” number of months payment is past due



# Data Attributes Details

## Balance Amount Attributes (6):

- “BILL\_AMT1, BILL\_AMT2, BILL\_AMT3, BILL\_AMT4, BILL\_AMT5, BILL\_AMT6” – Billing statement amount for six month period from Apr 2005 through Sep 2005. Integer data type.

## Payment Amount Attributes (6):

- “PAY\_AMT1, PAY\_AMT2, PAY\_AMT3, PAY\_AMT4, PAY\_AMT5, PAY\_AMT6” – Monthly payment amount for six month period from Apr 2005 through Sep 2005. Integer data type.

## Default Indicator:

- “default payment next month” – Binary data type classifier; Status of account indicating if it has defaulted “0” or not defaulted “1”

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## DATA SCIENCE FRAMEWORK

Our Approach to Managing the Project

# Framework One - Zumel and Mount (5 Steps)

1

Define Goals – Understand desired customer outcomes and evaluate feasibility based on the available data

2

Manage Data – Conduct exploratory data analysis and determine areas to improve quality and any other data preparation

3

Develop Model – Design and test predictive models and compare results - refine models for higher accuracy and lower errors

# Framework One - Zumel and Mount (5 Steps)

4 Present Results – Present results of the selected model to customer and solicit feedback about confidence in results – may require additional rework based on feedback

5 Deploy Model – Provide knowledge transfer to final model custodian and ensure thorough understanding of structure and process for follow-on maintenance or enhancements

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## DATA MANAGEMENT

Identification of Data Issues

# Data Management

- Ground truth data will be sourced from \*.csv file
- We will train and test predictive models based on ground truth data
- We can leverage \*.csv / \*.txt / \*.xlsx data files or data tables for unclassified data for final testing
- Upon implementation, Python model should be setup to receive data from a database table, such as SQL (PostgreSQL or Microsoft), MySQL (Oracle) , or NoSQL (Apache Cassandra)
- Data ETL and preparation and wrangling will be conducted via Python script(s) and final output to database table (preferred)



# Data Management – Any Known Issues

- Verify that final data is available through database
- Test and validate access to database table for read / write
- Ensure that the database table datatypes are the same in \*.csv file
- Determine if there are any other available data attributes, that can be leveraged such as credit score, liabilities, etc.
- Evaluate other data sources that can be leveraged in addition to the bank details, such as credit reporting data
- Identify and address any Personally Identifiable Information (PII) issues that could occur with the data collection and reporting

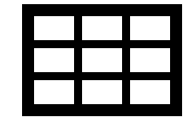
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## DATA PROCESS FLOW

Diagram of How Data Will be Processed

# Data Process Flow

## Possible Data Sources



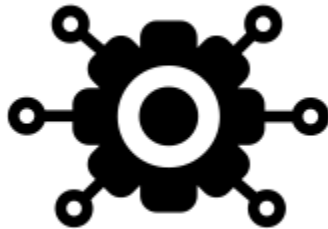
Data Table



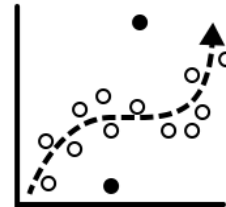
## Data Extraction, Transformation, and Loading (ETL)



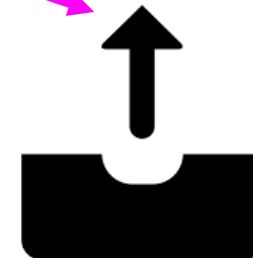
Extraction



Preparation

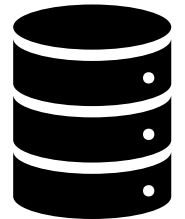


Predictive Model



Load Output

## Possible Data Destinations



Data Warehouse



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## ADDITIONAL INSIGHTS

Other Ideas Based on Review of Data

# Additional Insights of Data

- Extreme Outliers in Billing Statement Amount – (\$334K), (\$170K), and (\$151K) balances (credits) in April and June 2005 – we will determine if these should be eliminated during model training and evaluation
- Ensure that binary indicator for Default Payment classifier is a String / Factor for classification model training in Python
- Consolidate additional indicators for Education – “0,4,5,6” should be consolidated into one indicator for simplification of model
- Determine viability of creating new attributes that can summarize multiple fields, such as a binary string indicator for any late payments during the six months – “0” any late payments or “1” no late payments