## Loan Delinquencies in Analysis

# Understanding the roles of credit, inflation, and interest rates on loan delinquency

Client: Federal Student Aid

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```
import pandas as pd
import numpy as np
```

Importing the credit data Columns we are keeping:

- userID: unique respondent in-survey identification number used to match respondents between different surveys
- date: data survey was conducted (2013 2024)
- N1\_3: whether they possess student loan or not (1 for possess)
- N1\_4: whether they possess home-based loan or not (1 for possess)
- N1 5: whether they possess auto-based loan or not (1 for possess)
- N1\_6: whether they possess other personal loan or not (1 for possess)
- N2\_1: current balance of credit card loans
- N2\_3: current balance of student loans
- N10\_7: Dollar amount granted by lender for student loan
- N15: Loan payments late by more than 30 days in past 12 months (1 for late payment)
- N16: Loan payments late by more than 90 days in past 12 months (1 for late payment)
- N17a\_5: How likely will request credit card limit increase in next 12 months
- N22: Credit score (1 for below 620; 2 for 620 679; 3 for 680 719; 4 for 720 760; 5 for Above 760; 6 for don't know)
- N23: Last time checked credit score (1 for <1 mon; 2 for between 1 6 mon ago; 3 for 6-12 mon ago; 4 for 1-2 yrs ago; 5 for >2yrs ago; 6 for don't know)

• N25: Likelihood of being able to come up with \$2000 in the next month given an unexpected need (0-100 percent)

#### **267**31**5**.3820**1**00 0 NaN 3600.0NaN 1.0 NaN NaN NaN NaN NaN **341**31**0**.5570**1**00 0 20000100000NJaN 0.0 NaN NaN NaN NaN NaN NaN **202**31**0**.8680000 0 NaN NaN NaN NaN NaN NaN NaN NaN **205**31**0**.422**00**00 0 **N**aN NaN 0.0 NaN NaN NaN NaN NaN NaN NaN NaN 0.0 NaN NaN NaN NaN NaN **26\$**31**0**.638**00**00 0 250.0 50000**M**aN 0.0 NaN 2.0 4.0 93.0 **262**40**2**.7417**0**7 1 4.0**280**40**2**.262**0**60 0 8000.(NaN NaN 0.0 NaN NaN 5.0 1.0 100.0 **202**40**2**.4673**0**6 0 10000108000101aN 0.0 NaN NaN 1.0 1.0 0.0 **202**40**2**.5450**0**3 0 **5**2000**M**aN 0.0 NaN NaN 2.0 5.0 100.0 NaN 95000NjaN 0.0 NaN NaN 5.0 1.0 **802**40**0**.3823**7**9 0 99.0

```
#reduced_credit_df["year"] = reduced_credit_df["date"].astype(str).str[:4]
```

Importing the housing spending data Columns we are keeping:

- userID: unique respondent in-survey identification number used to match respondents between different surveys
- date: data survey was conducted (2013 2024)

reduced\_credit\_df.columns

- qsp3\_1: home appliance purchase during last 4 months (only have data from 2015 and on; house option added only Aug 2016)
- qsp3\_2: electronics, computers, or cell phone purchase during last 4 months (only have data from 2015 and on; house option added only Aug 2016)
- qsp3\_3: furniture purchase during last 4 months (only have data from 2015 and on; house option added only Aug 2016)
- qsp3\_4: home repairs, improvements or renovations purchase during last 4 months (only have data from 2015 and on; house option added only Aug 2016)
- qsp3\_5: cars/vehicle purchase during last 4 months (only have data from 2015 and on; house option added only Aug 2016)
- qsp3\_6: trip/vacation purchase during last 4 months (only have data from 2015 and on; house option added only Aug 2016)
- qsp3\_9: house/apartment purchase during last 4 months (only have data from 2015 and on; house option added only Aug 2016)
- qsp3\_8: no large purchase during last 4 months (only have data from 2015 and on; house option added only Aug 2016)
- qsp5\_1: proportion of current monthly salary on housing
- qsp5\_2: proportion of current monthly salary on utilities
- qsp5\_3: proportion of current monthly salary on food
- qsp5\_4: proportion of current monthly salary on clothing, footwear, personal care
- qsp5\_6: proportion of current monthly salary on transportion
- qsp5\_7: proportion of current monthly salary on medical care
- qsp5\_8: proportion of current monthly salary on education and child care
- qsp5\_9: proportion of current monthly salary on other (gifts, child support, charity, etc.)
- qsp12\_n: if get 10% more income, what they would do (3 for use all to pay debt; 5,6, 7 for use some to pay debt; 1, 2, 4 for use none for debt)
- qsp15new: variability in household income month to month (1 for <5%; 2 for 5%-15%; 3 for >15%)
- k2e: whether there exist a family budget (1 for yes)

```
' hhspending_df = pd.read_excel("potentialDataSetsHW2/
sce_household_spending.xlsx", sheet_name="Data")\nreduced_hhspending_df =
hhspending_df[["userid", "date", \n
"qsp3_1", "qsp3_2", "qsp3_3", "qsp3_4", "qsp3_5", "qsp3_6", "qsp3_9", "qsp3_8",\n
"qsp5_1", "qsp5_2", "qsp5_3", "qsp5_4", "qsp5_5", "qsp5_6", "qsp5_7", "qsp5_8",
"qsp5_9",\n "qsp12n", "qsp15new",
"k2e"]]\nreduced_hhspending_df '
```

```
#reduced_hhspending_df["year"] =
reduced_hhspending_df["date"].astype(str).str[:4]
```

#### Importing Job and Income Data

Columns we want to keep:

- userid: Unique identifier
- date: date survey was conducted (2014 and on)
- 13: annual salary before tax and deductions (including bonuses, overtime, tips/commissions)
- 17dk\_1: never had a paying job (1 for never; empty for having a paying job)

```
' job_df = pd.read_excel("potentialDataSetsHW2/sce_labor.xlsx",
sheet_name="Data")\nreduced_labor_df = job_df[["userid", "date", \n
"l3", "l7dk_1"]]\nreduced_labor_df '
```

```
#reduced_labor_df["year"] = reduced_labor_df["date"].astype(str).str[:4]
```

Importing household status

Columns we want to keep:

- userid
- date

- a1new: current living situation (1 for married; 2 for separated; 3 for divorced; 4 for widowed; 5 for never married)
- b1 1: household income
- b5\_1: Percentage change in HH income last 12 mos. vs. previous year
- i1 6: Disabled

```
' hhfinance_df = pd.read_excel("potentialDataSetsHW2/
sce_household_finance.xlsx", sheet_name="Data")\nreduced_hhfinance_df =
hhfinance_df[["userid", "date", \n
"alnew", "bl_1", "b5_1", "i1_6"]]\nreduced_hhfinance_df '
```

```
#reduced_hhfinance_df["year"] =
reduced_hhfinance_df["date"].astype(str).str[:4]
```

```
""" reduced_credit_df.set_index(["userid", "date"], inplace = True)
reduced_hhspending_df.set_index(["userid", "date"], inplace = True)
reduced_labor_df.set_index(["userid", "date"], inplace = True)
reduced_hhfinance_df.set_index(["userid", "date"], inplace = True) """
```

```
' reduced_credit_df.set_index(["userid", "date"], inplace =
True)\nreduced_hhspending_df.set_index(["userid", "date"], inplace =
True)\nreduced_labor_df.set_index(["userid", "date"], inplace =
True)\nreduced_hhfinance_df.set_index(["userid", "date"], inplace = True) '
```

```
""" joined_df = reduced_credit_df.join(reduced_hhspending_df, how = "inner")
"""
```

```
' joined_df = reduced_credit_df.join(reduced_hhspending_df, how = "inner") '
```

```
' common_index = set(reduced_credit_df.index) &
set(reduced_hhfinance_df.index) \n\nprint(f"Number of common (userid, date):
{len(common_index)}") '
```