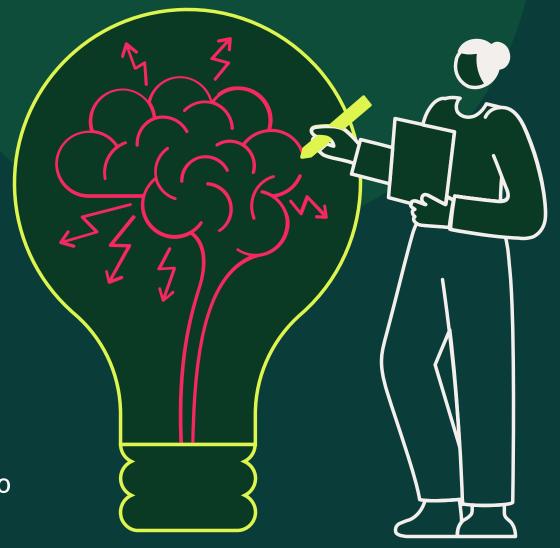


Business Framing Document

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Agenda



Business – Description,
Objectives and
Background



Success Criteria



Data



Risks and Contingencies



Stretch Goals and Plan

Business Problem And Objectives





Prepayment vs. Default Risks: Freddie Mac must balance the competing risks of early loan payoffs and defaults, which directly affect financial stability.



Economic Sensitivity: Borrower behaviors vary widely with economic changes, requiring adaptable forecasting models.

OBJECTIVES

- Monthly Transition Rate Tracking: Predict monthly mortgage status changes and minimize forecast errors.
- Error Minimization: Compare predicted vs. actual rates using root mean square error.
- Period Analysis: Evaluate performance across pre-COVID, COVID, and post-COVID phases.

Business Background

Stakeholders: Freddie Mac provides liquidity to mortgage lenders, allowing them to issue loans to borrowers, while investors rely on Freddie Mac's mortgage-backed securities for returns, and regulatory bodies oversee these processes to ensure stability and affordability in the housing market.

Resources Available:

- 1. Freddie Mac Loan-level Dataset https://www.freddiemac.com/research/datasets/sf-loanlevel-dataset
- 2. Freddie Mac House Price Index Data http://www.freddiemac.com/research/indices/house-price-index.page

Problem History

- <u>California's Housing Volatility:</u> Sharp price and foreclosure shifts from tech changes, regulations, and economic cycles demand adaptable models.
- Prepayment vs. Default Risks: High property values drive prepayments; downturns increase defaults, impacting Freddie Mac's stability.
- Crises Reveal Vulnerabilities: Events like 2008 and COVID-19 exposed risks of sharp price drops and increased foreclosures in California, showing the need for resilient, tailored models.

Success Criteria

Minimizing Forecasting Errors

The primary success metric is reducing the monthly average RMSE between the predicted and actual transition rates of mortgages. Smaller errors indicate a more reliable model.

Pre-COVID, COVID, and Post-COVID Period Analysis Success also involves demonstrating how the model performs across these periods, ensuring the model adapts to changing conditions.

Data Requirements

1. Freddie Mac Loan-Level Dataset

Contains detailed loan-level data for mortgages purchased by Freddie Mac. Key Variables shown on the right.

2. Freddie Mac House Price Index

Provides house price indices at national, state, and MSA levels.

Key Variables: Year, Month, Geo_Name, Index_NSA, Index_SA

3. Federal Reserve Economic Data

Provides the data for the underlying economic drivers throughout the years.

Key Variables: GDP, unemployment rates, inflation rates, etc.

Sl No	Column Name	
1	CREDIT SCORE	
2	FIRST PAYMENT DATE	
3	FIRST TIME HOMEBUYER FLAG	
4	MATURITY DATE	
5	METROPOLITAN STATISTICAL AREA (MSA) OR METROPOLITAN DIVISION	
6	MORTGAGE INSURANCE PERCENTAGE (MI %)	
7	NUMBER OF UNITS	
8	OCCUPANCY STATUS	
9	ORIGINAL COMBINED LOAN-TO-VALUE (CLTV)	
10	ORIGINAL DEBT-TO-INCOME (DTI) RATIO	
11	ORIGINAL UPB	
12	ORIGINAL LOAN-TO-VALUE (LTV)	
13	ORIGINAL INTEREST RATE	
14	CHANNEL	
15	PREPAYMENT PENALTY MORTGAGE (PPM) FLAG	
16	AMORTIZATION TYPE	
17	PROPERTY STATE	
18	PROPERTY TYPE	
19	POSTAL CODE	
20	LOAN SEQUENCE NUMBER	
21	LOAN PURPOSE	
22	ORIGINAL LOAN TERM	
23	NUMBER OF BORROWERS	
24	SELLER NAME	
25	SERVICER NAME	
26	SUPER CONFORMING FLAG	
27	PRE-RELIEF REFINANCE LOAN SEQUENCE NUMBER	
28	PROGRAM INDICATOR	
29	RELIEF REFINANCE INDICATOR	
30	PROPERTY VALUATION METHOD	
31	INTEREST ONLY INDICATOR (I/O INDICATOR)	
32	MI CANCELLATION INDICATOR	

Assumptions

- 1. Historical loan-level data is complete and accurately represents loan performance and status transitions.
- 2. Borrower behavior remains relatively consistent within each defined period.
- 3. External economic factors influencing borrower decisions are adequately captured in the dataset.
- 4. Data quality is sufficient for analysis, with minimal missing values and inconsistencies.

Limitations

- 1. Changes in borrower behavior and economic conditions during COVID-19 may not be fully reflected in the data.
- 2. The impact of forbearance programs may not be fully reflected in the dataset.
- 3. Unforeseen external factors like regulatory changes that occurred during the specified periods may not be captured in the dataset.

Risks and Contingencies

Data Quality Issues	Missing or incomplete loan performance data may lead to incorrect predictions. Mitigation: Implement data cleaning and validation techniques to handle gaps in the dataset.
Model Overfitting	If the model fits the training data too well, it may not generalize to the testing period (COVID and post-COVID). Mitigation: Use cross-validation and regularization techniques and find new variables that can improve prediction for these time periods.
Effects of Covid-19	The effect of COVID-19 on the economy may not be significantly represented in the data, potentially leading to inadequate model predictions for the COVID period. Mitigation: Analyze the data to identify any gaps related to the pandemic's impact and consider integrating external data sources to better capture these effects.
Regulatory Changes	Policy changes during the pandemic, both short and long term, would have affected borrower behavior. Mitigation: Stay informed about regulatory developments during this period and make changes to the model accordingly.

Stretch Goals

Alternative Data Sources

Explore using Google Trends or financial news sentiment data to improve predictions, as this could capture public sentiment about economic conditions or housing markets.

Advanced Machine Learning Models

Investigate deep learning approaches to improve accuracy.

Real-Time Forecasting

Implementing a model that updates predictions in realtime as new performance data becomes available could provide immediate insights to stakeholders.

Long Term Plan



Thank you

