

Credit Risk Analysis & Portfolio Insights

Author: Arif Shaikh

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Dataset: Kaggle - Bank Loan Status Dataset

1. Problem Statement

Our client seeks to identify high-risk borrowers to reduce defaults while maintaining portfolio size. Using loan-level data, we engineered a transparent risk score and analyzed portfolio exposure to quantify risk concentration and guide credit decision-making.

2. Key Insights

| Risk Factor | Observation | Impact on Default |
|-----------------------------------|------------------------------------------------------------------------------------------------------------------------------------------------|---------------------------------------------------------------------|
| Credit Score | Default rate decreases as credit score increases. Very poor & poor scores contribute disproportionately to defaults. | Strong predictor of credit risk. |
| Debt-to-Income (DTI) | Borrowers with DTI >35% have the highest default rates. | Primary driver (≈93% contribution). |
| Credit Utilization | High utilization correlates with higher default probability. | Secondary driver (≈11% contribution). |
| Employment Tenure | Tenure <2 years = higher default risk. | Moderate predictor |
| Home Ownership | Renters are riskier than owners. | Minor but relevant driver |
| Risk Segmentation | The portfolio concentrated in Medium (52%) & High (44%) risk bands , contributing most to cumulative defaults. Low-risk loans = 3%. | Indicates portfolio exposure is significant in higher-risk segments |
| Trend Analysis (Window Functions) | Cumulative defaults accumulate faster in Very High / High-risk bands. | Shows actionable segments for monitoring |

Summary: Credit score, DTI, and credit utilization are the **top 3 risk drivers**. Medium & high-risk borrowers account for the majority of defaults, highlighting the portfolio's exposure.

3. SQL Key Outputs

i) Loan-level table with risk metrics (`credit_risk_scoring`)

| | risk_bucket | avg_credit_score | avg_dti | avg_utilization | loans |
|---|-------------|------------------|---------|-----------------|-------|
| ▶ | High Risk | 634 | 18.36 | 0.64 | 166 |
| | Medium Risk | 711 | 17.23 | 0.62 | 2662 |
| | Low Risk | 730 | 12.47 | 0.33 | 2255 |

ii) Analytical Base Table (ABT) (`credit_risk_abt`)

| | total_loans | total_defaults | avg_dti | avg_utilization | total_high_risk |
|---|-------------|----------------|---------|-----------------|-----------------|
| ▶ | 5083 | 981 | 15.16 | 0.49 | 4784 |

iii) Risk Band Table (`credit_risk_bands`)

| | risk_band | loans | portfolio_pct |
|---|-------------|-------|---------------|
| ▶ | Medium Risk | 2662 | 52.37 |
| | High Risk | 2255 | 44.36 |
| | Low Risk | 166 | 3.27 |

iv) Cumulative Default Table (`credit_risk_ranked`)

| | loan_id | risk_score | risk_band | is_default | cumulative_defaults | cumulative_loans | cumulative_default_pct |
|---|--------------------------------------|------------|---------------|------------|---------------------|------------------|------------------------|
| ▶ | 41b50163-6c7d-4887-91b7-4b1e26733c17 | 2.9 | Very Low Risk | 0 | 46 | 169 | 27.22 |
| | cf075856-e58b-4f8d-98b1-a4d4f38cdb57 | 2.9 | Very Low Risk | 1 | 46 | 168 | 27.38 |
| | 82251b66-c544-4d23-b48e-55aac76b1cd8 | 2.9 | Very Low Risk | 0 | 45 | 167 | 26.95 |
| | cee21435-fe7b-4c64-8f92-39d15996a35e | 2.9 | Very Low Risk | 0 | 45 | 166 | 27.11 |
| | cfc59665-2766-4bcc-930a-9c785d44b810 | 2.9 | Very Low Risk | 1 | 45 | 165 | 27.27 |
| | 9edff292-c673-4285-974c-5554d3d5b925 | 2.9 | Very Low Risk | 0 | 44 | 164 | 26.83 |
| | c17c19e8-d1cf-4e90-b6bb-50fa67c81e7c | 2.9 | Very Low Risk | 0 | 44 | 163 | 26.99 |
| | e88ec80e-f8f0-4539-89b9-d7ac4ea84b6e | 2.9 | Very Low Risk | 1 | 44 | 162 | 27.16 |
| | cdac6fd5-4a9c-4103-abfe-c72c184edc0e | 2.9 | Very Low Risk | 0 | 43 | 161 | 26.71 |
| | feb2b313-99e0-4d6a-9777-018205de8b40 | 2.9 | Very Low Risk | 1 | 43 | 160 | 26.88 |

4. Recommended Actions

- Review or reject top 10–20% highest-risk loans.
- Implement enhanced monitoring for Medium Risk loans.
- Maintain standard workflow for Low Risk loans.
- Use DTI and utilization thresholds as early warning triggers.
- Future enhancement: incorporate macroeconomic or behavioral data for predictive modeling.

5. Business Impact

| Metric | Current | Expected Improvement |
|------------------------|------------------|------------------------------------------------------------------|
| Portfolio Default Rate | ~19% | Reduce to <16% (by targeting high-risk borrowers) |
| High-Risk Exposure | 44% of portfolio | Focused monitoring could mitigate ~50–60% of potential losses. |
| Decision-Making Speed | Manual analysis | SQL-based dashboards enable real-time actionable insights |

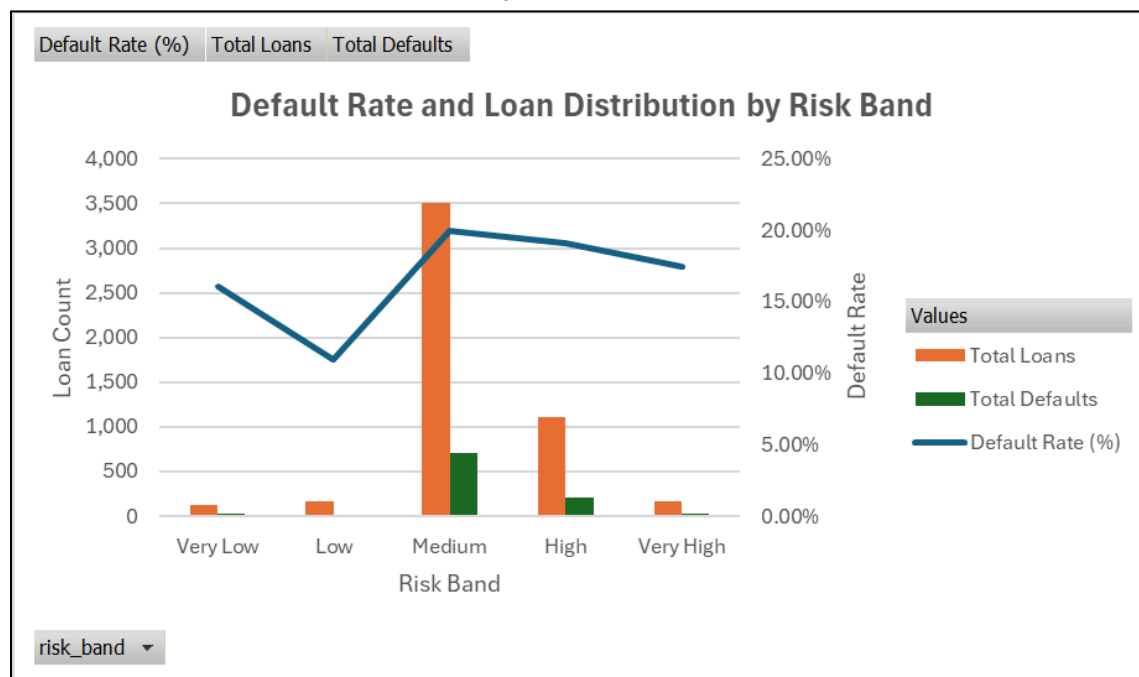
Implementing these measures will **improve risk-adjusted portfolio performance** and strengthen the bank's credit decision process

6. Implementation Plan

- Integrate **risk_score** and risk bands into loan approval workflow.
- Generate monthly risk band reports with cumulative default metrics.
- Review top drivers periodically; recalibrate score as needed.
- Optionally, extend score with macroeconomic trends or behavioral indicators.

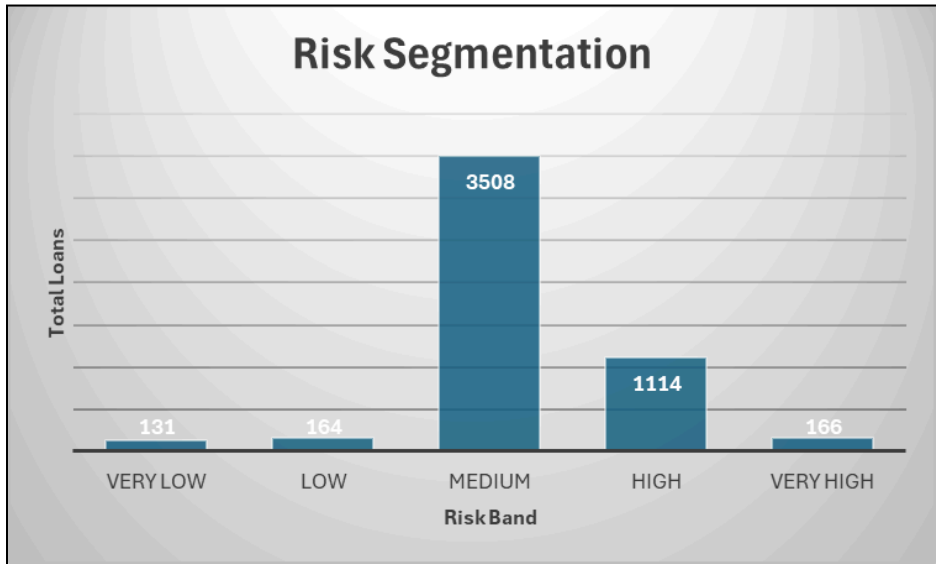
7. Visuals

i) Default Rate and Loan Distribution by Risk Band



The chart shows Default Rate (%), Total Loans, and Total Defaults across risk bands from Very Low to Very High. Most loans are concentrated in the Medium risk band, which also accounts for a large portion of total defaults. Default rates generally increase with risk, confirming that the risk scoring system aligns with actual borrower behavior. The slight decline in the Very High band is due to a smaller number of loans, which introduces some volatility at the extreme tail.

ii) Risk Segmentation



The portfolio is dominated by **Medium-risk borrowers** (3,508 loans), while **High** and **Very High-risk** loans total 1,280, representing significant default exposure. **Low** and **Very Low-risk** loans are rare (295 loans), showing that most of the portfolio carries moderate to high risk. Focused monitoring of these segments is essential for managing defaults.

iii) Top 5 Riskiest Borrowers

| Result Grid | | | | | | | Filter Rows: | Export: | Wrap Cell Content: | Fetch rows: |
|-------------|--------------------------------------|--------------------------------------|------------|-------|--------------------|--|--------------|---------|--------------------|-------------|
| | loan_id | customer_id | risk_score | dti | credit_utilization | | | | | |
| | f9ad44d2-ec3d-4998-813d-2c11f2c69334 | 9abdb8b9-9ccf-4a27-89d8-240628ebc3f2 | 4.3 | 29.4 | 0.77 | | | | | |
| | 2945b719-a387-494e-9c49-b4e2b500e135 | d4f59a9f-98a7-476b-bc80-32cb7a7a94ee | 4.3 | 7.01 | 0.78 | | | | | |
| | c52938aa-5c99-4e2f-9745-1c6b768cab17 | d65be7af-e9fa-41e0-b653-0c81c69d85f8 | 4.3 | 16.2 | 0.79 | | | | | |
| | 1cf81c05-be9f-4d19-9a35-6c8d11c56eb5 | 31b9cc39-5856-4fc9-b6d9-b1692bedcf57 | 4.3 | 28.6 | 0.83 | | | | | |
| | 05f70be6-5930-4536-8091-538e5dc2a48f | e0b51878-e75e-4283-8d43-8d5be06b69ec | 4.3 | 12.96 | 0.85 | | | | | |

The top 5 riskiest borrowers all have a **risk score of 4.3**, indicating extremely high default potential. Their **Debt-to-Income ratios range from 7% to 29%**, and **credit utilization** is between 0.77–0.85, showing that both high debt relative to income and high credit usage contribute to elevated risk. These borrowers highlight the types of accounts that require **immediate monitoring or intervention**.

8. Conclusion

The analysis demonstrates that Credit Score, DTI, and Credit Utilization are the strongest drivers of default risk. Medium and High-risk borrowers account for the majority of portfolio exposure and cumulative defaults. Implementing targeted monitoring, early warning triggers, and SQL-driven dashboards will reduce default rates, improve risk-adjusted performance, and support more informed, data-driven credit decisions. Future enhancements could include predictive modeling with macroeconomic and behavioral data to further optimize portfolio management.