

Model Testing Report

Experiment: model_testing

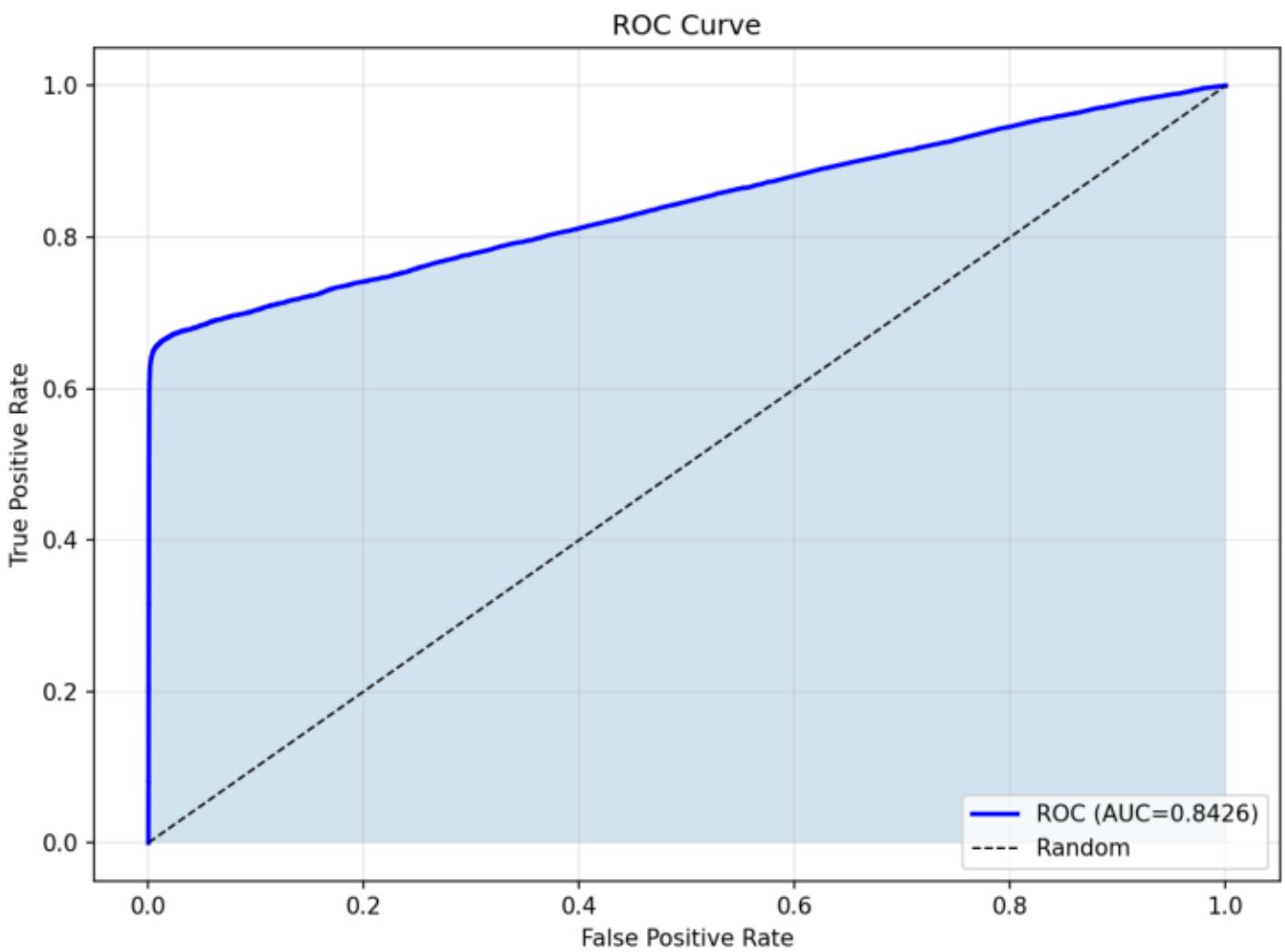
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1) Model Effectiveness

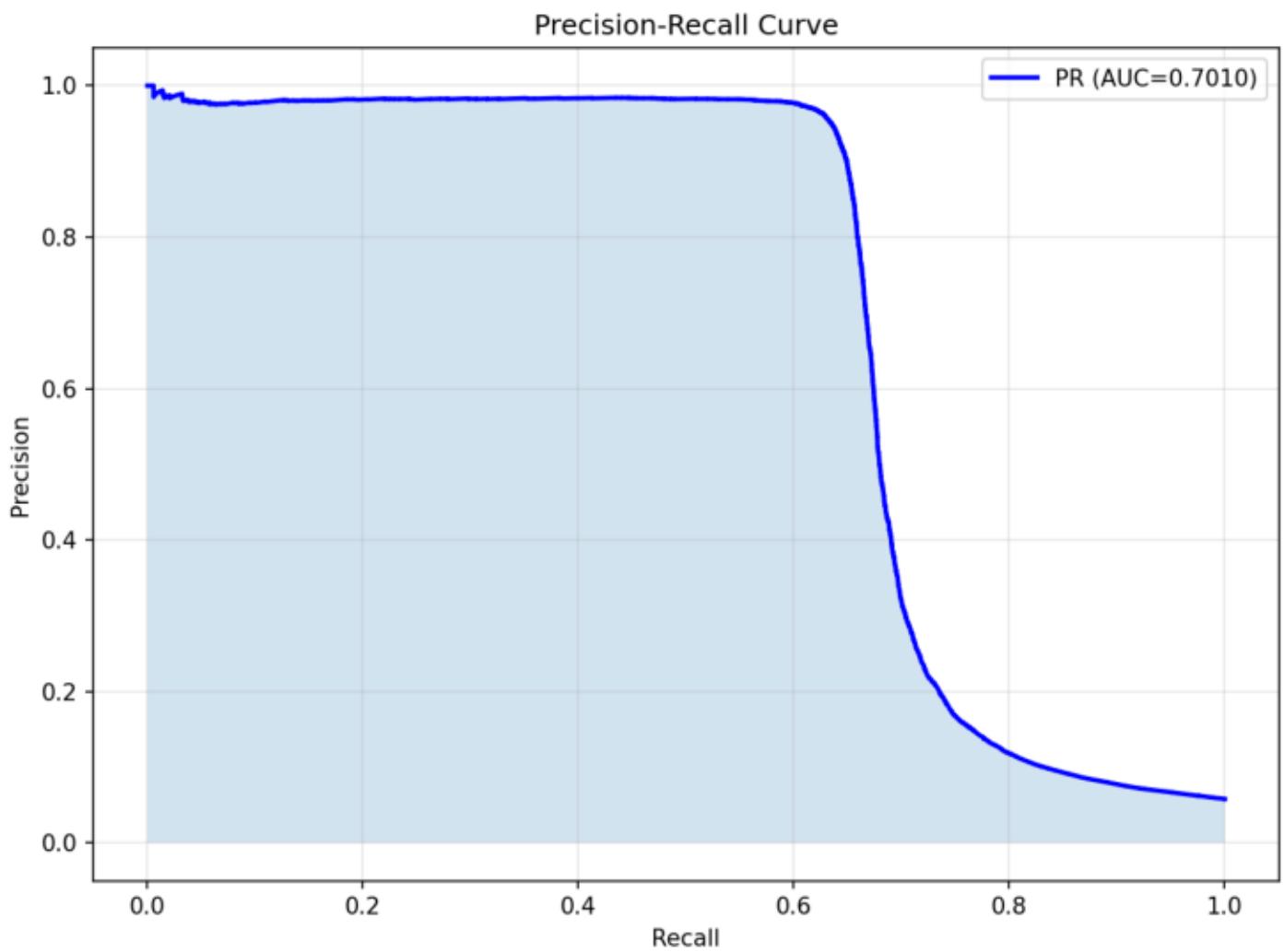
- auc_roc: 0.8426
- auc_pr: 0.7010
- precision: 0.9713
- recall: 0.6140
- f1: 0.7524
- ks_statistic: 0.6503
- ks_threshold: 0.0800
- confusion_matrix.TN: 188273
- confusion_matrix.FP: 209
- confusion_matrix.FN: 4446
- confusion_matrix.TP: 7072
- precision_at_k.10: 1.0000
- precision_at_k.50: 1.0000
- precision_at_k.100: 0.9900
- precision_at_k.200: 0.9850
- precision_at_k.500: 0.9800
- recall_at_k.10: 0.0009
- recall_at_k.50: 0.0043
- recall_at_k.100: 0.0086
- recall_at_k.200: 0.0171
- recall_at_k.500: 0.0425

1) Model Effectiveness - Explanations

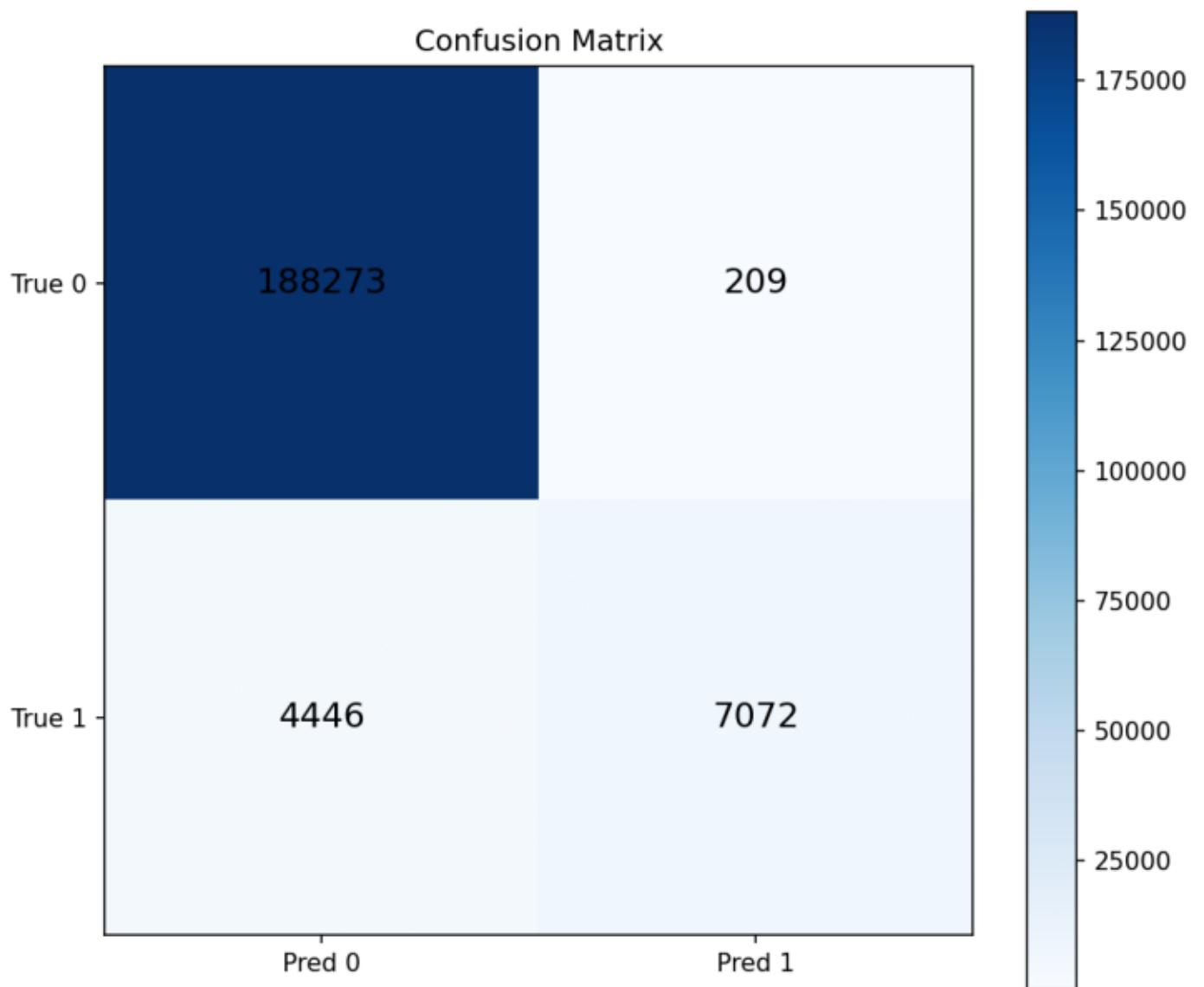
- Positive rate is 5.76%; AUC-ROC is 0.8426 (strong ranking).
- AUC-PR is 0.7010, above the base rate 5.76%, indicating meaningful lift.
- At threshold 0.50, precision=0.9713, recall=0.6140, F1=0.7524 (precision-heavy (fewer positives caught, fewer false alarms)).
- Confusion matrix: TP=7072, FP=209, FN=4446, TN=188273 (FPR=0.11%, FNR=38.60%).
- KS is 0.6503 at threshold 0.08, indicating very strong separation.
- Top-10: precision=1.0000, recall=0.0009 shows the quality of the highest-risk shortlist.
- Top-500: precision=0.9800, recall=0.0425 shows how much coverage you get with a larger queue.

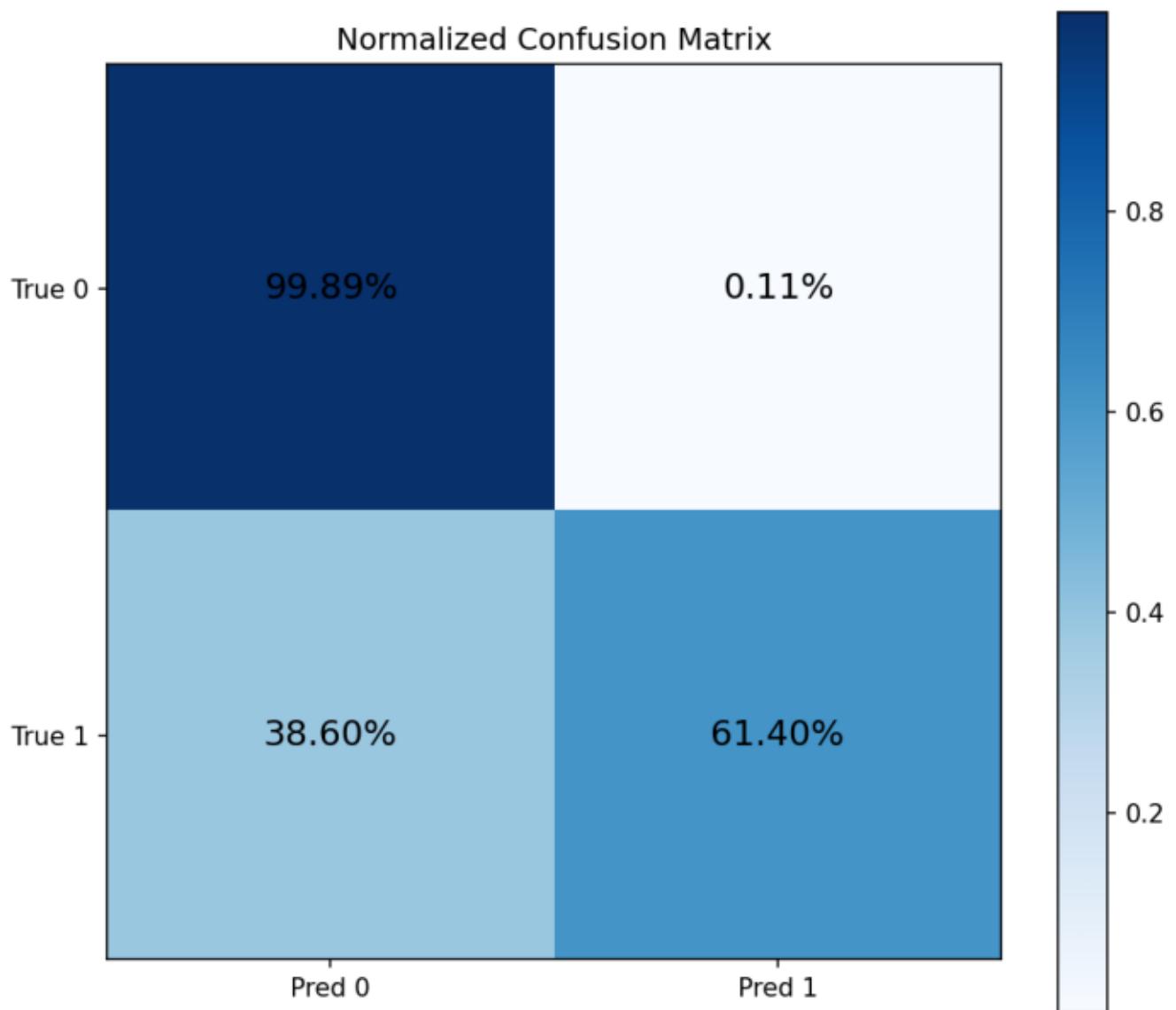


roc_curve: AUC-ROC=0.8426; curve above diagonal indicates strong separation.

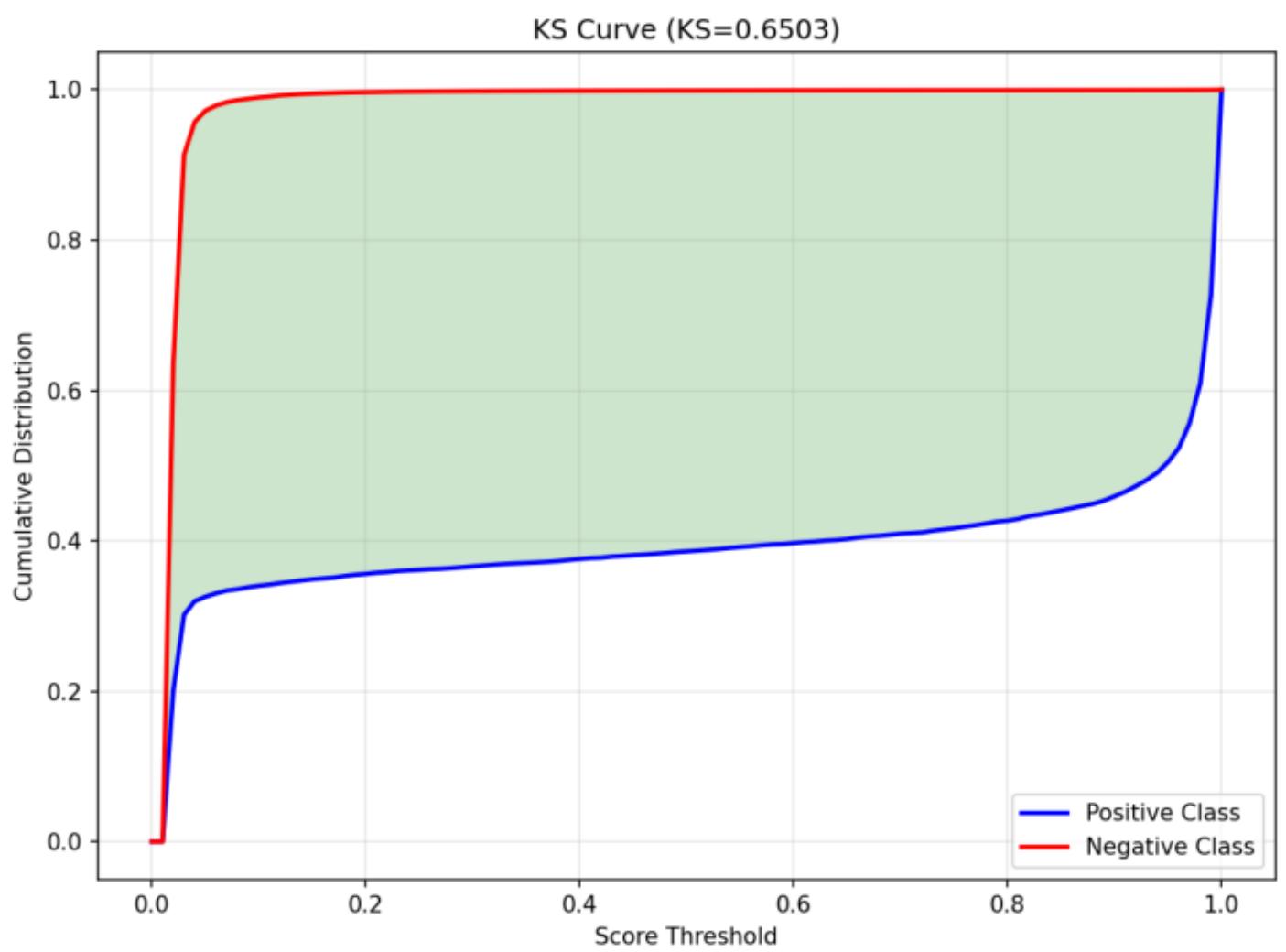


pr_curve: AUC-PR=0.7010 vs baseline 5.76%; higher curve implies better precision at recall.



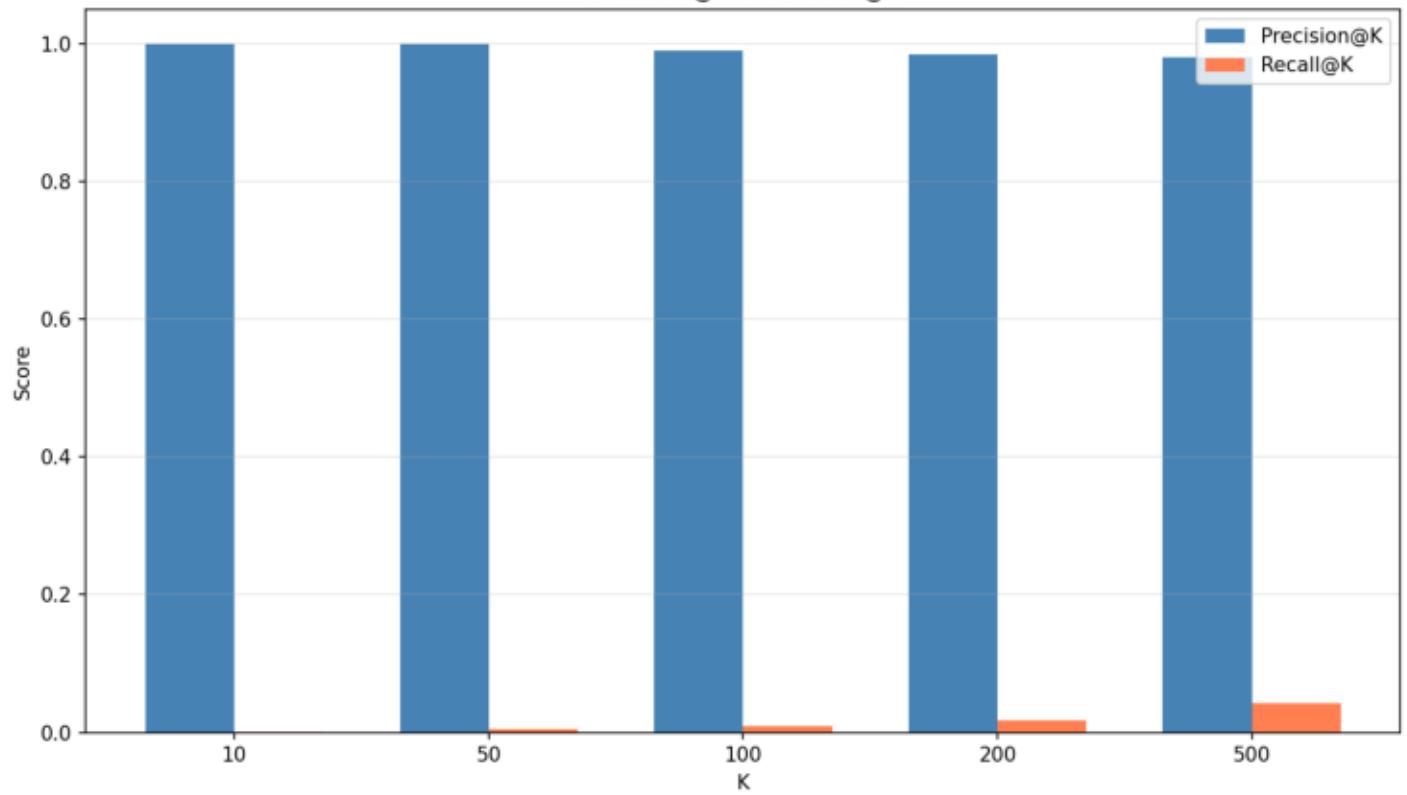


confusion_matrix_norm: Normalized rates: TPR=61.40%, TNR=99.89%.

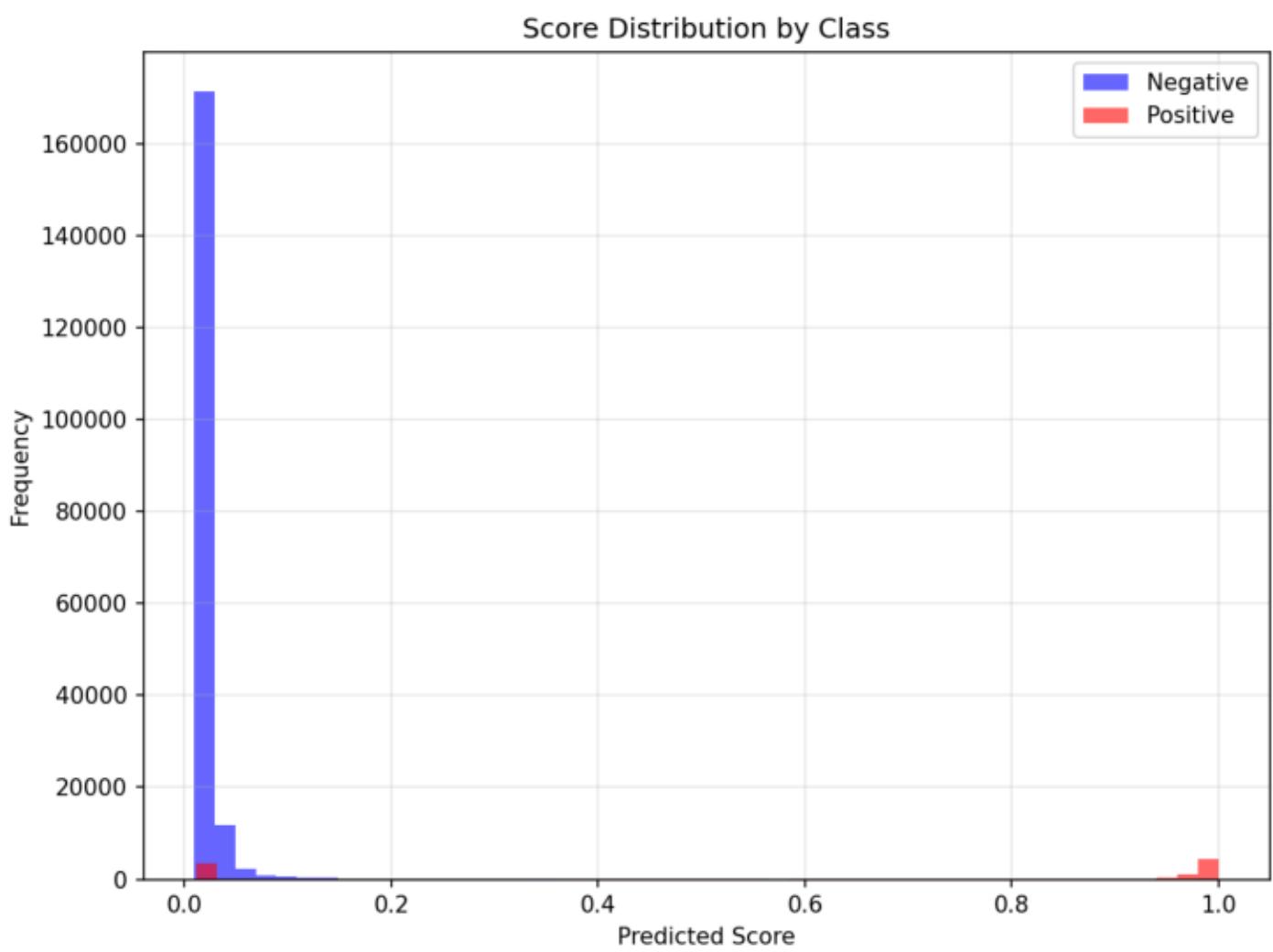


ks_curve: Maximum separation KS=0.6503 at threshold 0.08.

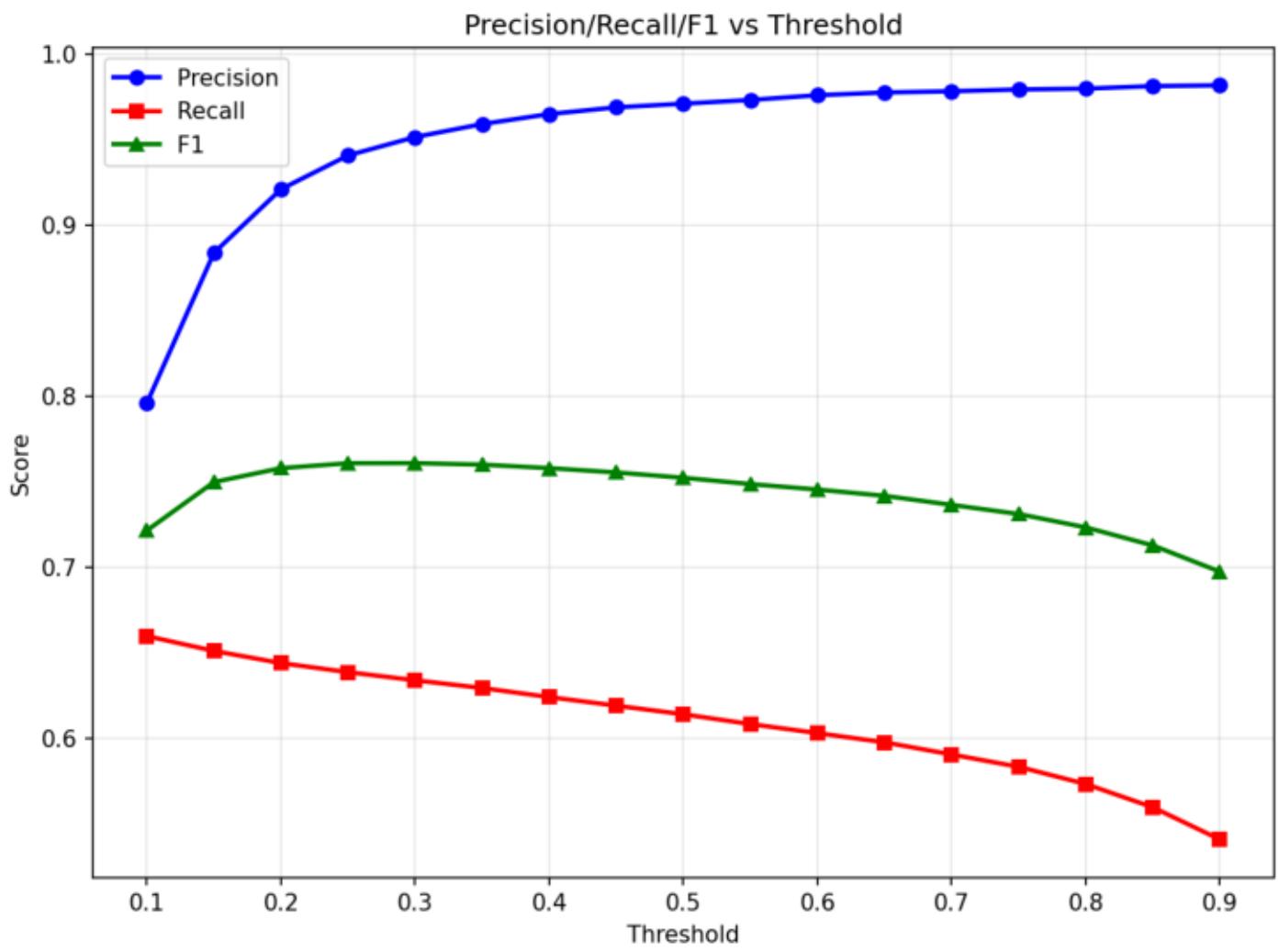
Precision@K and Recall@K



precision_recall_at_k: Bars show precision/recall as you expand the review queue.



score_distribution: Mean score: positive=0.607, negative=0.024 (larger gap is better).



threshold_analysis: Best F1 in tested grid is 0.7610 at threshold 0.30.

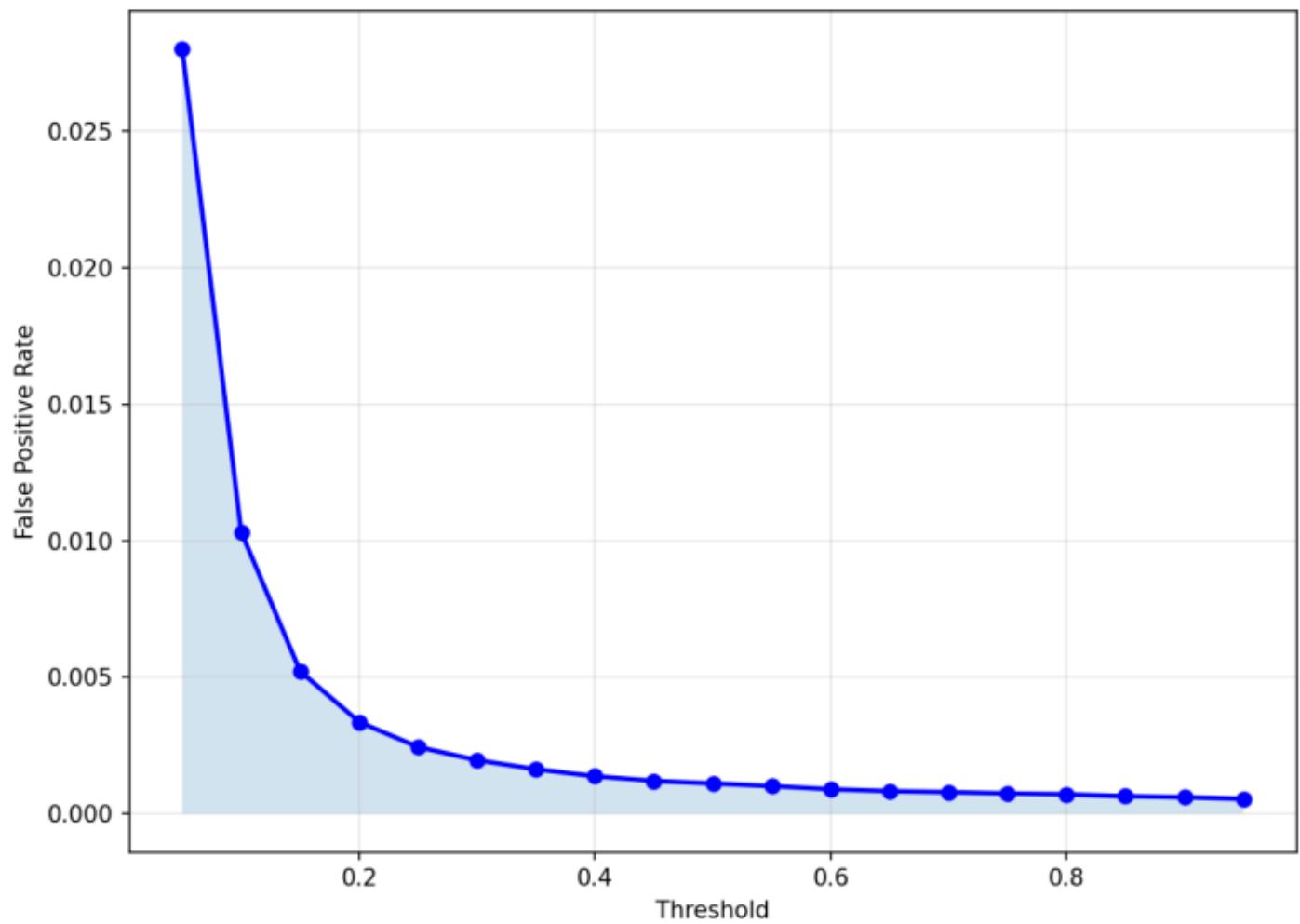
2) Model Efficiency

```
- fpr: 0.0011
- tn: 188273
- fp: 209
- threshold: 0.5000
- fpr_at_thresholds.t_0.05: 0.0280
- fpr_at_thresholds.t_0.10: 0.0103
- fpr_at_thresholds.t_0.15: 0.0052
- fpr_at_thresholds.t_0.20: 0.0034
- fpr_at_thresholds.t_0.25: 0.0024
- fpr_at_thresholds.t_0.30: 0.0020
- fpr_at_thresholds.t_0.35: 0.0016
- fpr_at_thresholds.t_0.40: 0.0014
- fpr_at_thresholds.t_0.45: 0.0012
- fpr_at_thresholds.t_0.50: 0.0011
- fpr_at_thresholds.t_0.55: 0.0010
- fpr_at_thresholds.t_0.60: 0.0009
- fpr_at_thresholds.t_0.65: 0.0008
- fpr_at_thresholds.t_0.70: 0.0008
- fpr_at_thresholds.t_0.75: 0.0007
- fpr_at_thresholds.t_0.80: 0.0007
- fpr_at_thresholds.t_0.85: 0.0006
- fpr_at_thresholds.t_0.90: 0.0006
- fpr_at_thresholds.t_0.95: 0.0005
```

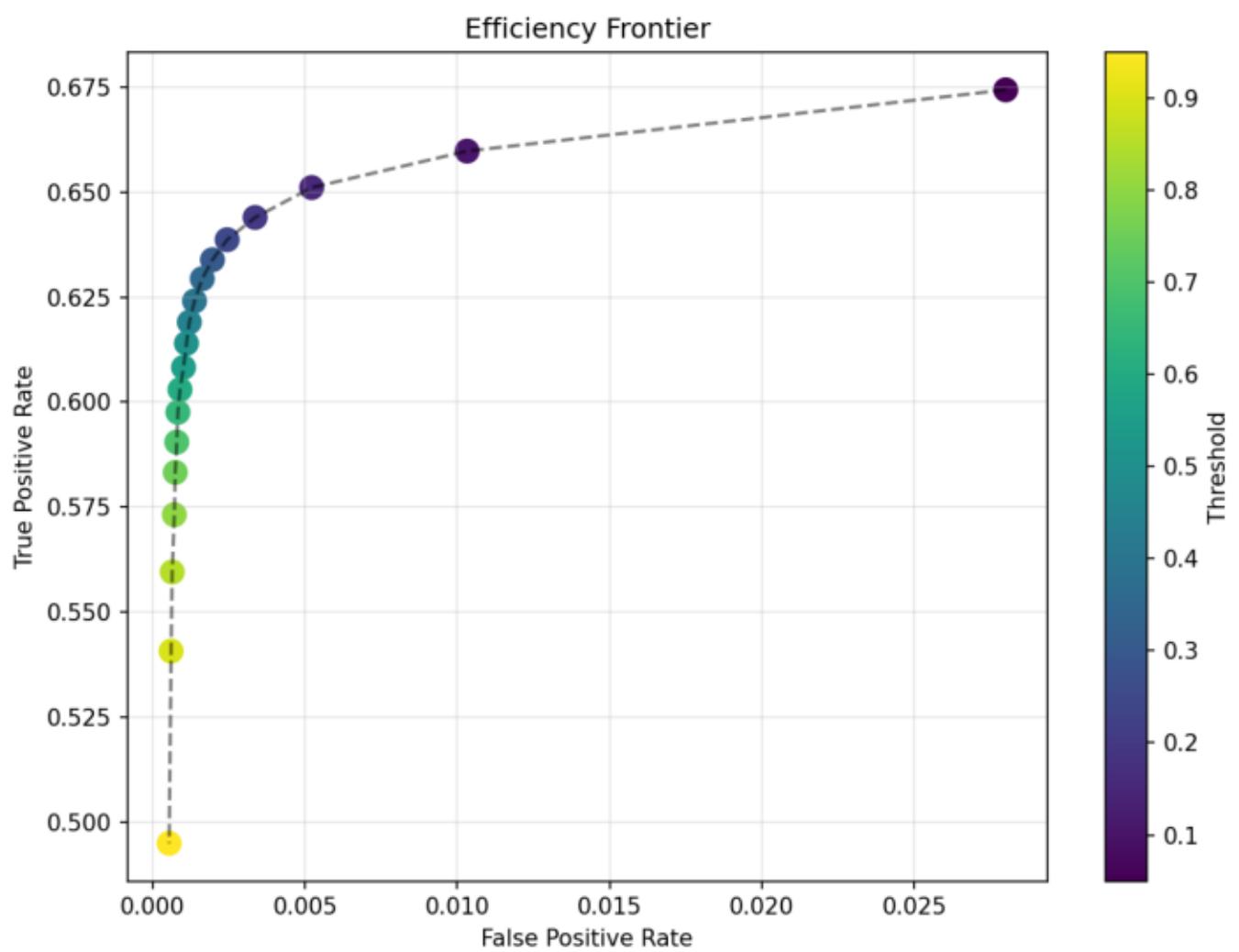
2) Model Efficiency - Explanations

- At threshold 0.50, $FPR=0.0011$ (low); $FP=209$ out of 188482 negatives.
- At the same threshold, $TPR=0.6140$ with $TP=7072$ and $FN=4446$, showing the capture rate of positives.
- A threshold near 0.05 yields $FPR \approx 0.0280$ with $TPR \approx 0.6743$ if you want to target ~5% false positives.

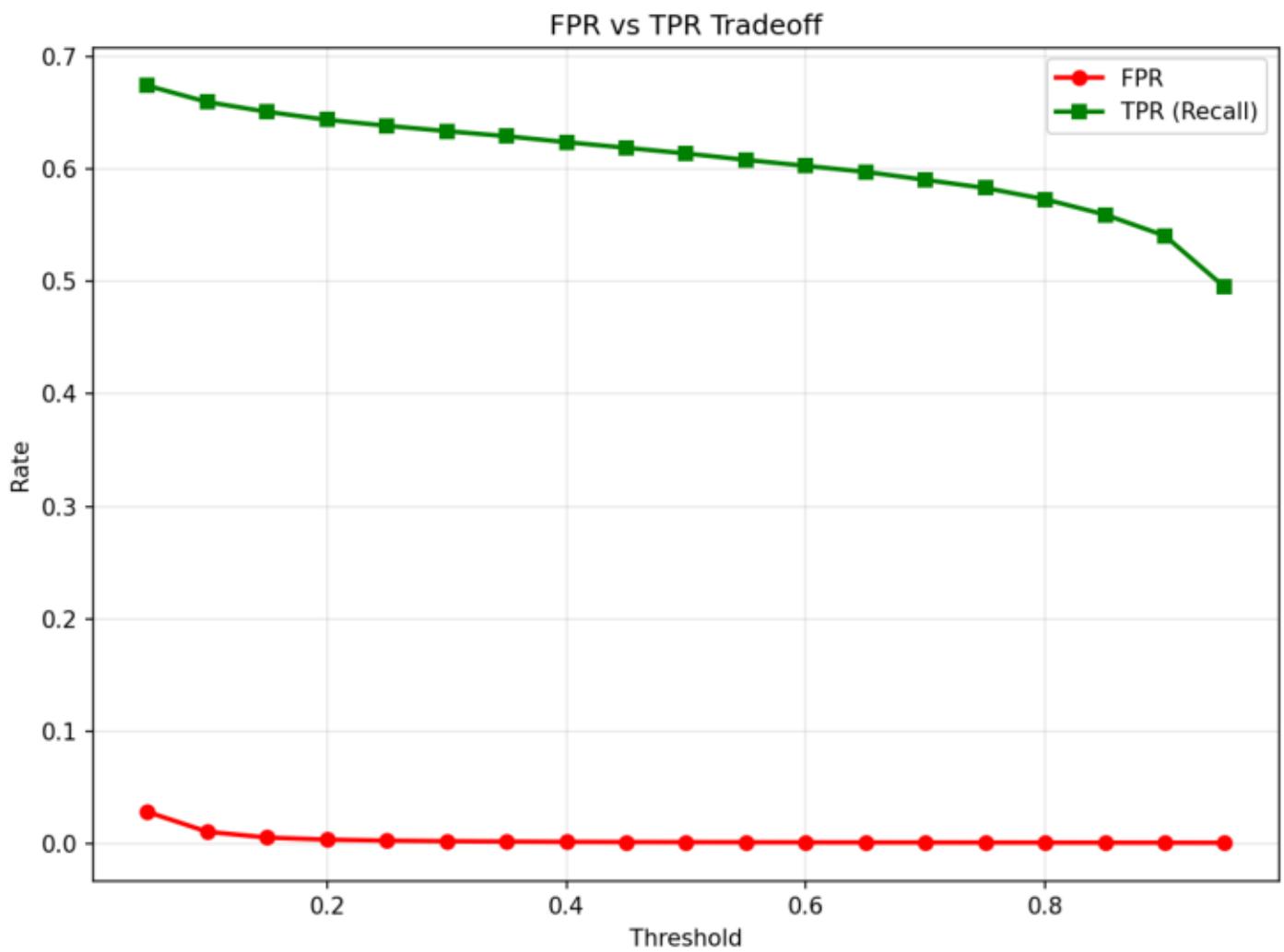
FPR vs Classification Threshold



fpr_vs_threshold: FPR decreases as the threshold increases; use it to pick an operating point.



efficiency_frontier: Each point shows the FPR/TPR tradeoff; move toward the top-left for better efficiency.



fpr_tpr_tradeoff: FPR and TPR curves highlight how recall drops as you reduce false positives.

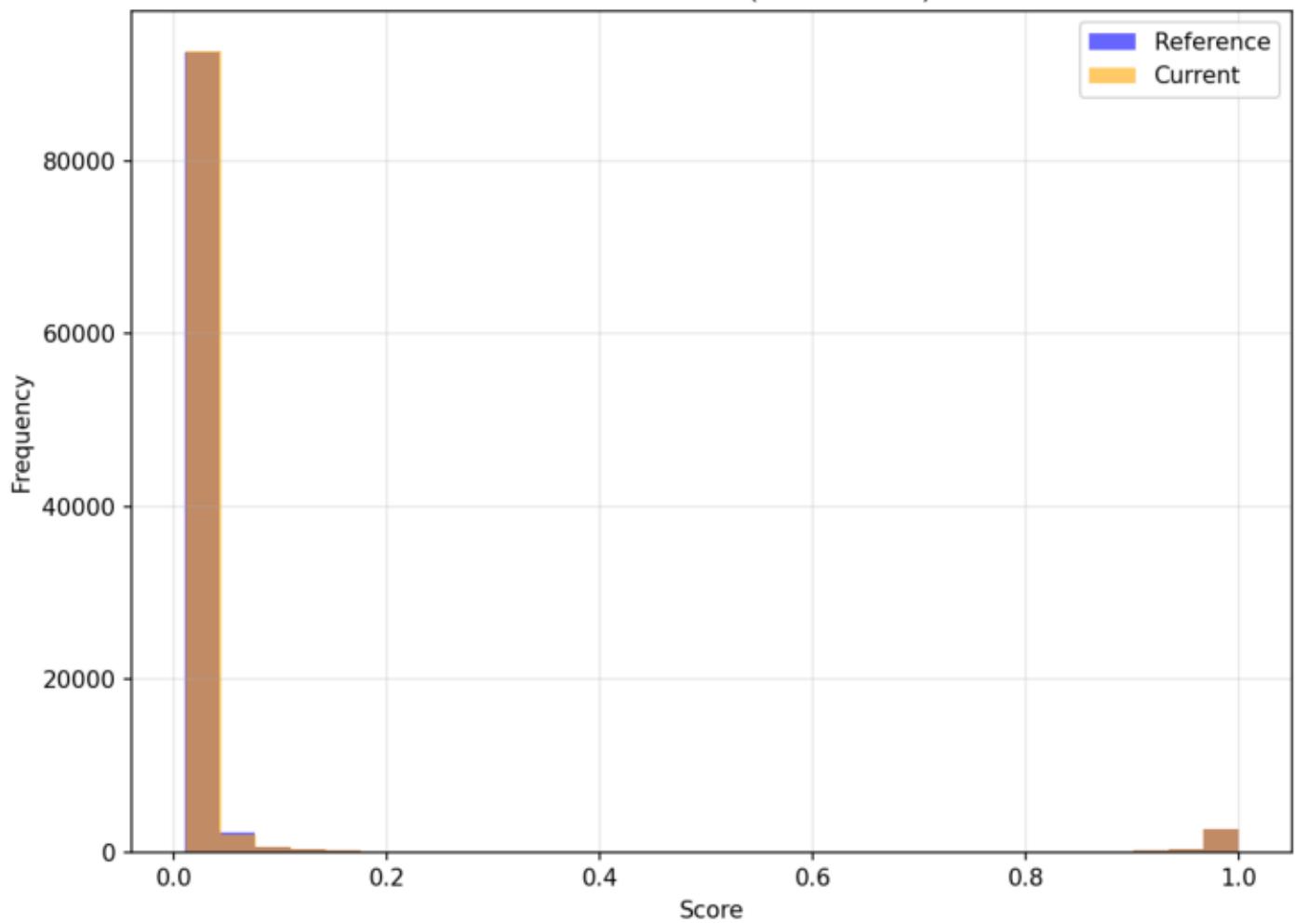
3) Model Stability

- psi: 0.0002
- cv_auc_roc_mean: 0.8426
- cv_auc_roc_std: 0.0047
- cv_auc_pr_mean: 0.7010
- cv_auc_pr_std: 0.0086
- bootstrap_auc_roc_mean: 0.8400
- bootstrap_auc_roc_ci_lower: 0.8338
- bootstrap_auc_roc_ci_upper: 0.8474
- concept_drift_detected: False
- concept_drift_score: 0.0078

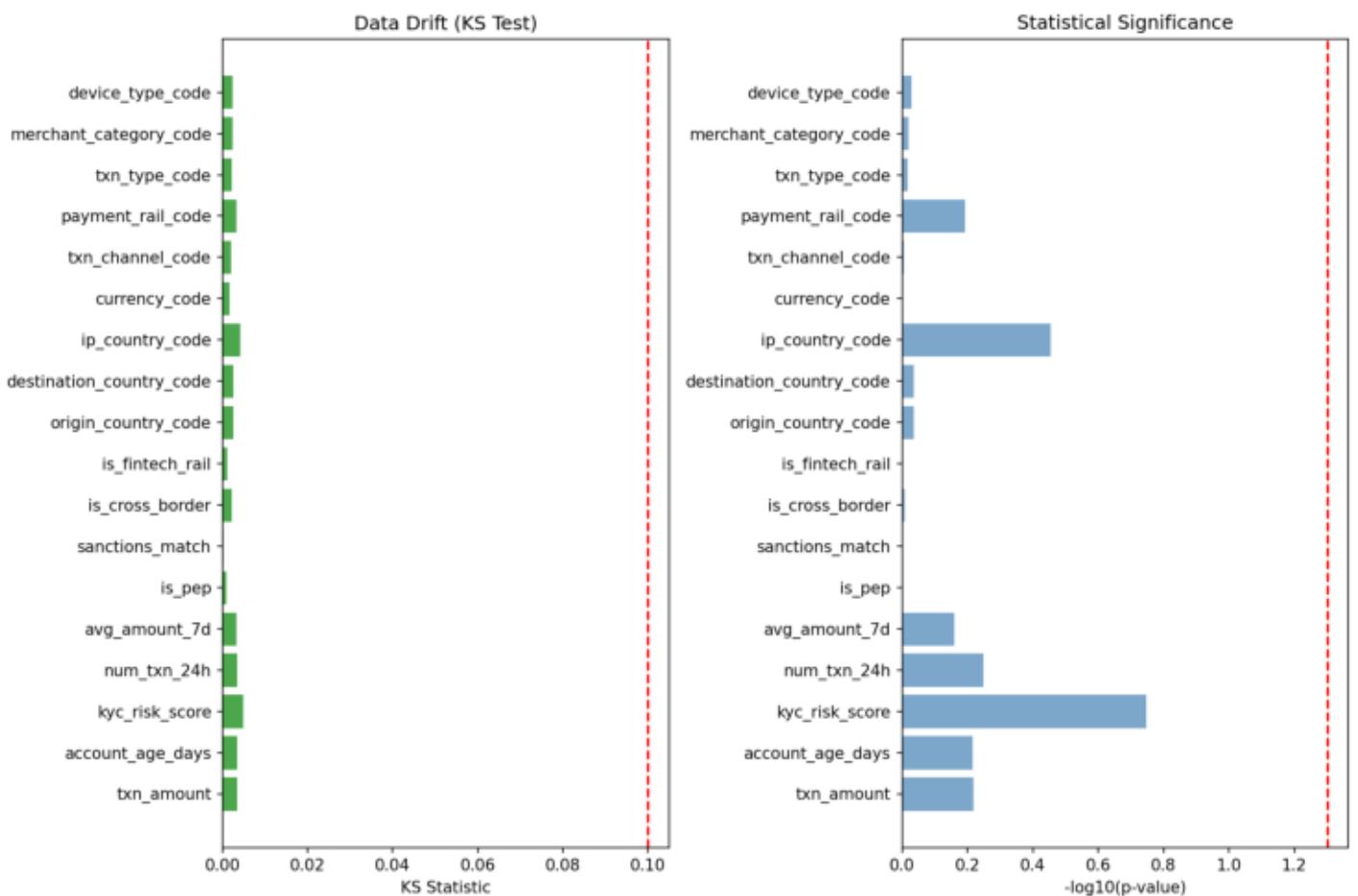
3) Model Stability - Explanations

- PSI is 0.0002, indicating negligible score distribution shift between reference and current data.
- CV AUC-ROC is 0.8426 ± 0.0047 (stable across folds).
- CV AUC-PR is 0.7010 ± 0.0086 (stable across folds).
- Bootstrap AUC-ROC mean is 0.8400 with 95% CI [0.8338, 0.8474] (width 0.0136).
- No concept drift detected (score 0.0078), suggesting stable performance over time.
- Data drift flagged 0/18 numeric features; top drift signals: ['kyc_risk_score', 'ip_country_code', 'num_txn_24h'].

Score Distribution (PSI=0.0002)

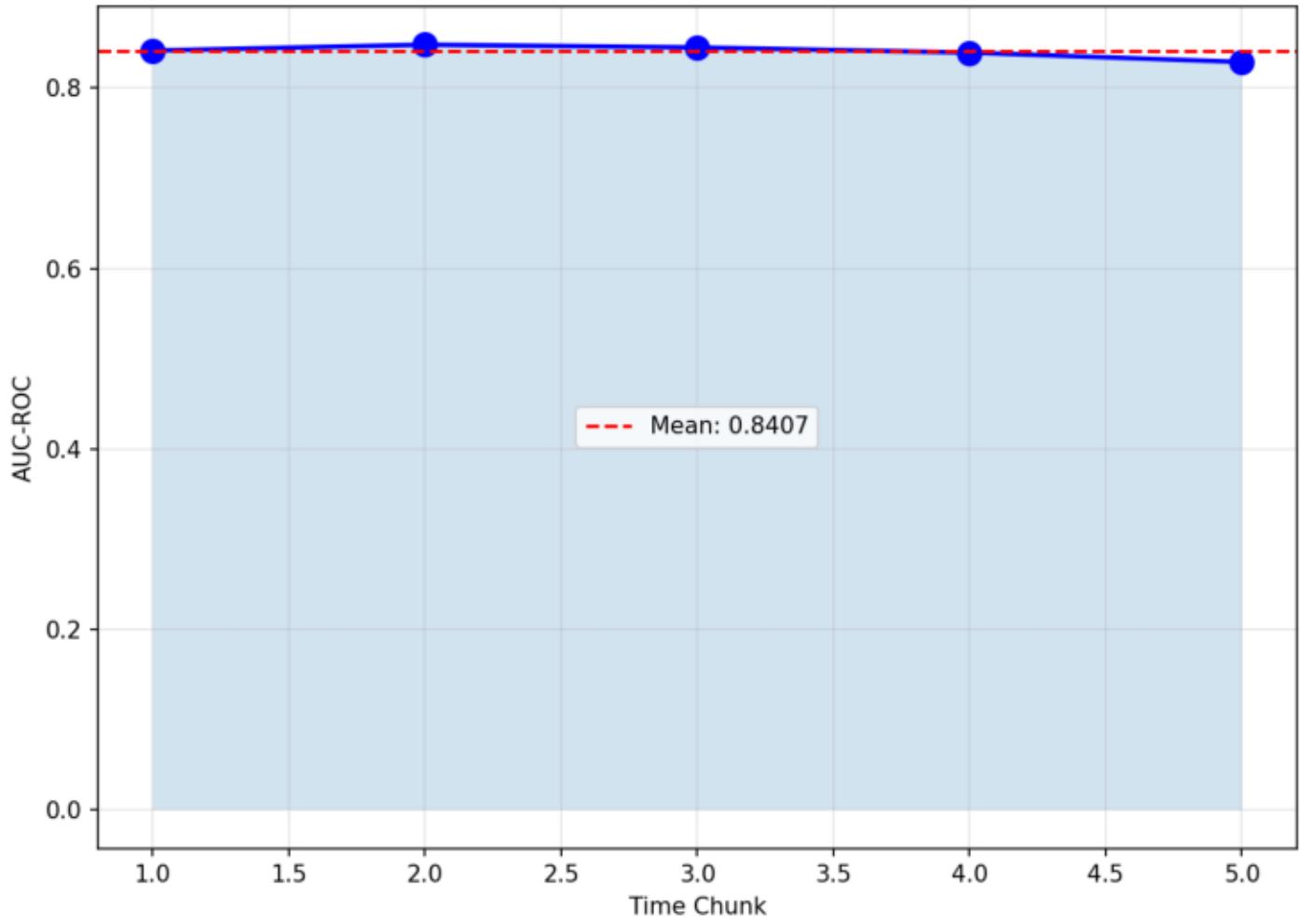


psi_distribution: Score distributions overlap consistent with PSI=0.0002.

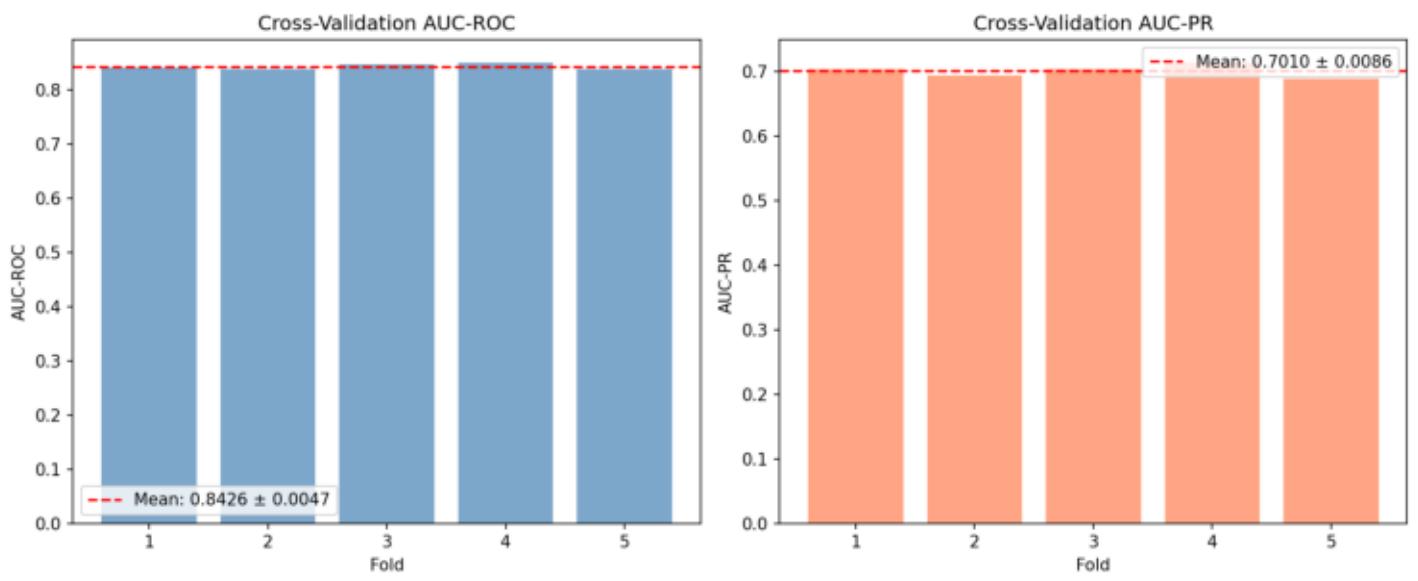


data_drift_heatmap: 0/18 features show drift; red bars highlight flagged features.

Concept Drift Analysis (Drift: Not Detected)

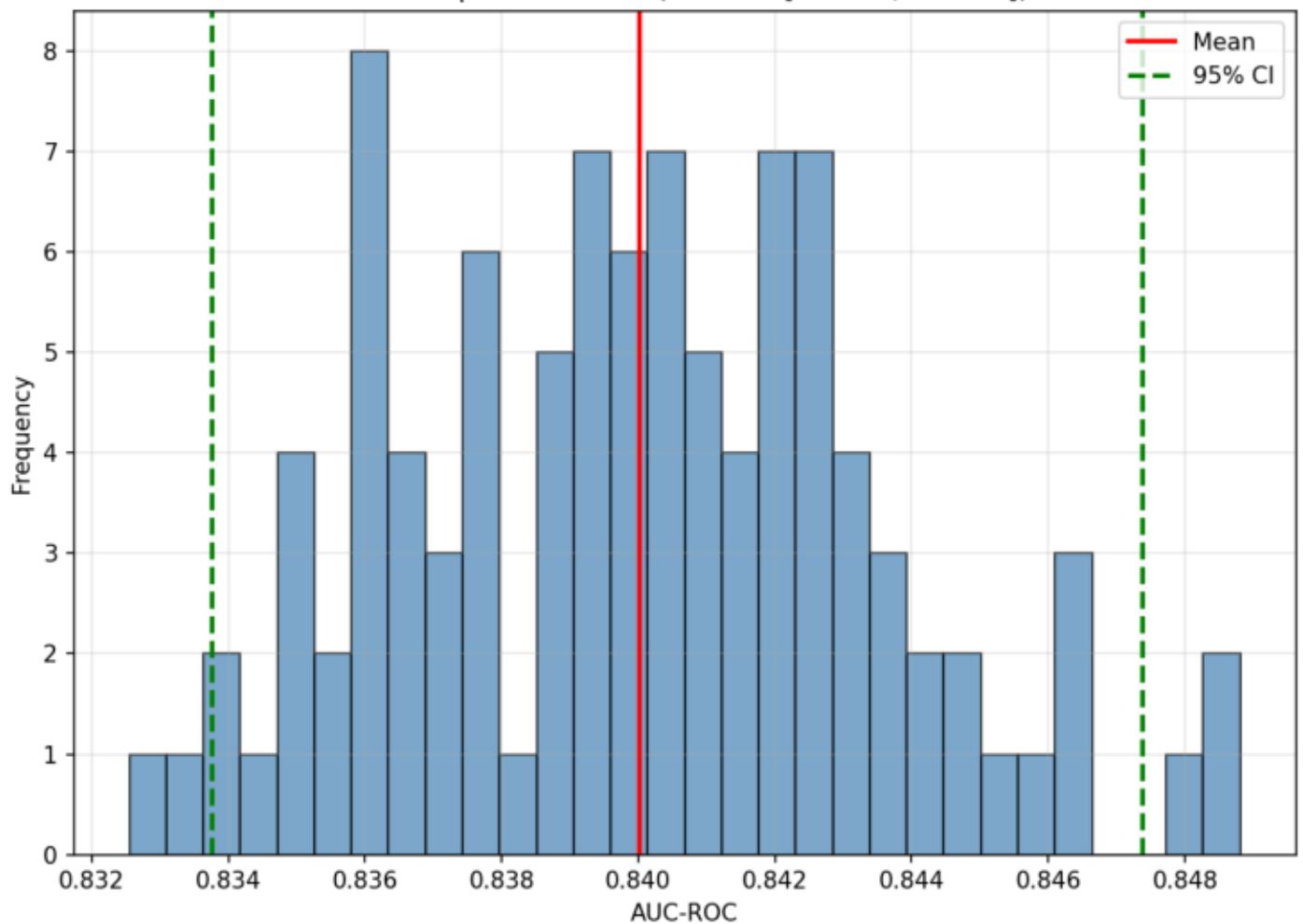


concept_drift: AUC across time chunks shows whether performance is stable over time.

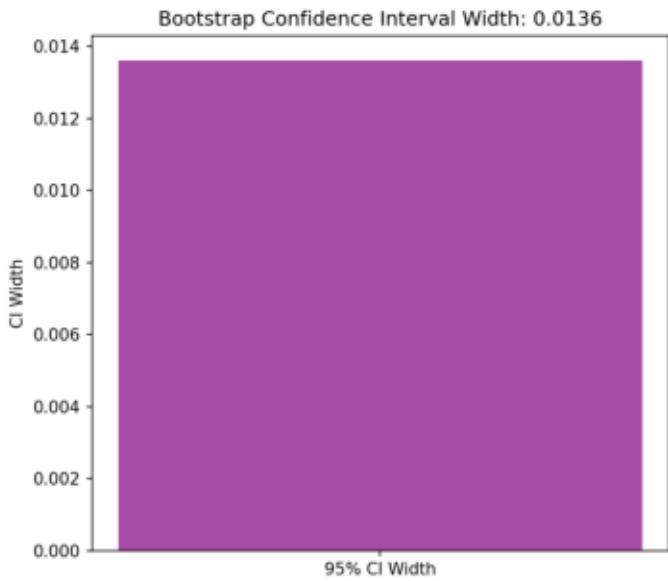
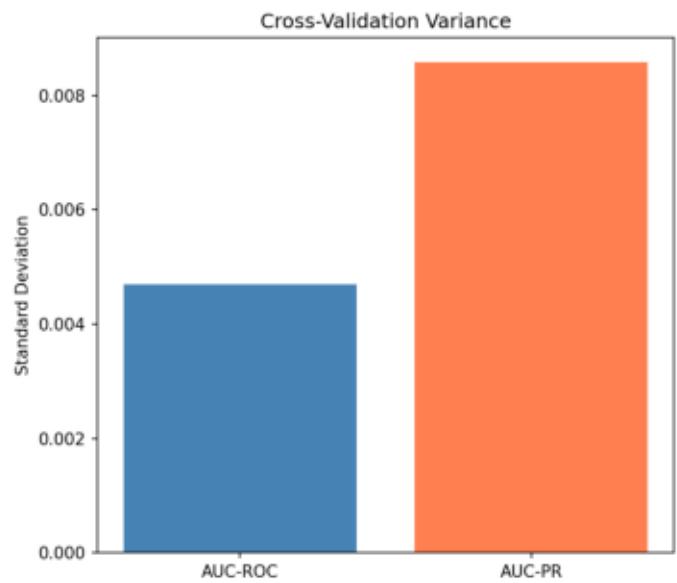
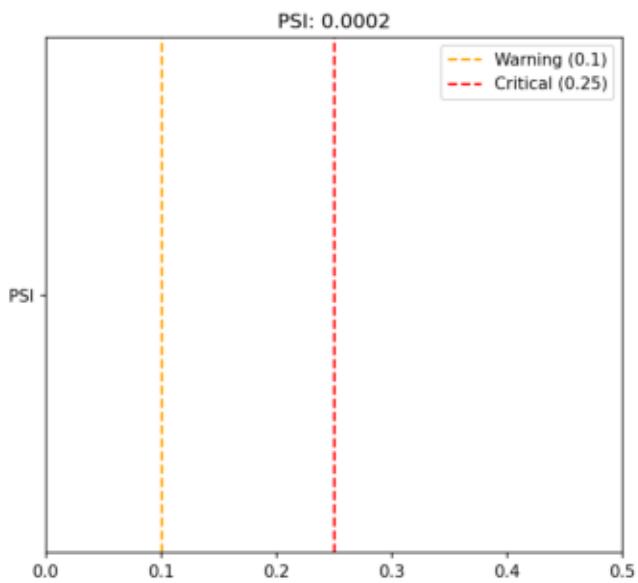


cv_results: Fold scores cluster around ROC 0.8426 and PR 0.7010.

Bootstrap Distribution (95% CI: [0.8338, 0.8474])



bootstrap_distribution: Bootstrap CI [0.8338, 0.8474] reflects performance stability.



STABILITY SUMMARY

PSI: 0.0002 (OK)

CV AUC-ROC: 0.8426 ± 0.0047
CV AUC-PR: 0.7010 ± 0.0086

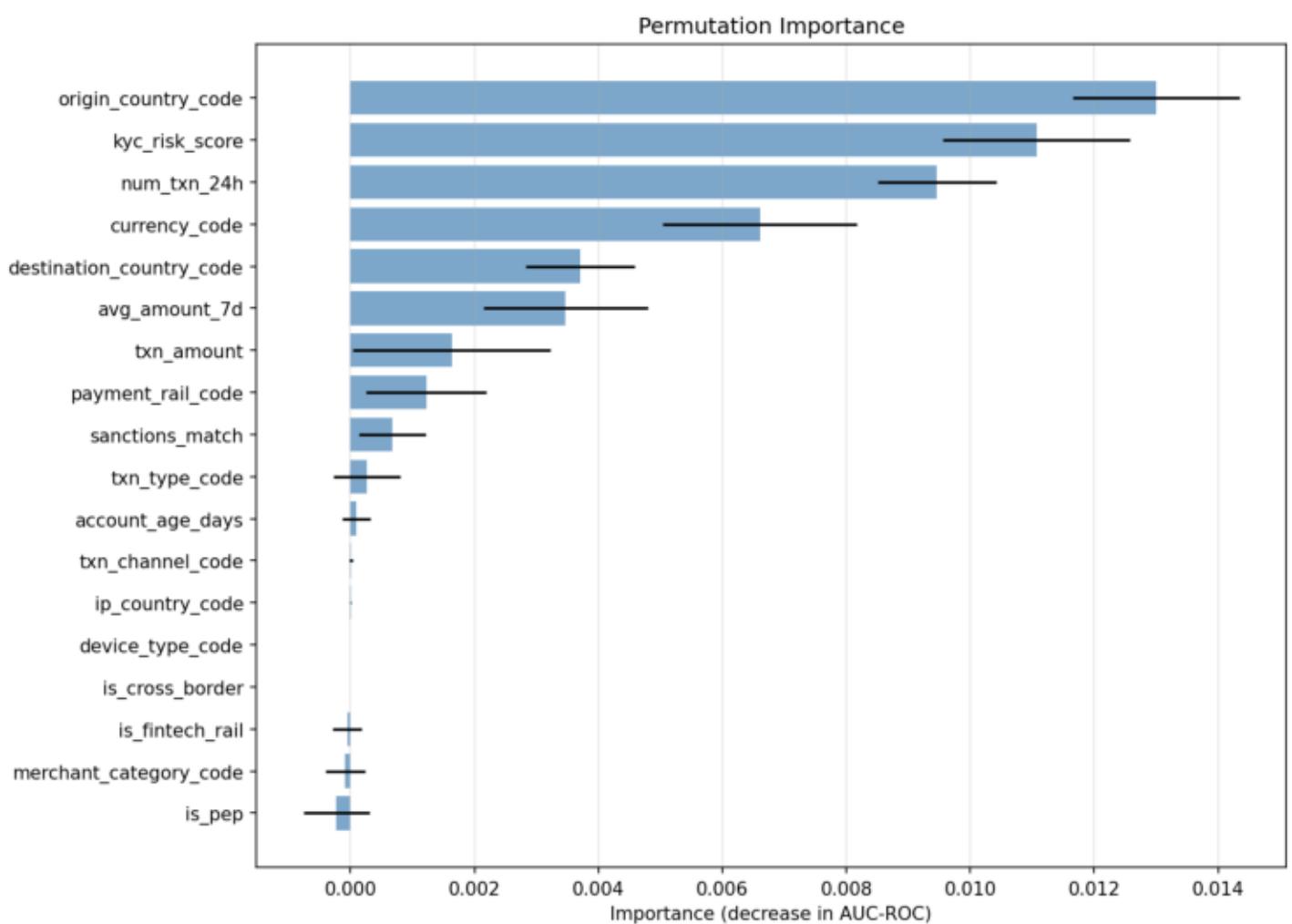
Bootstrap 95% CI: [0.8338, 0.8474]

4) Model Interpretability

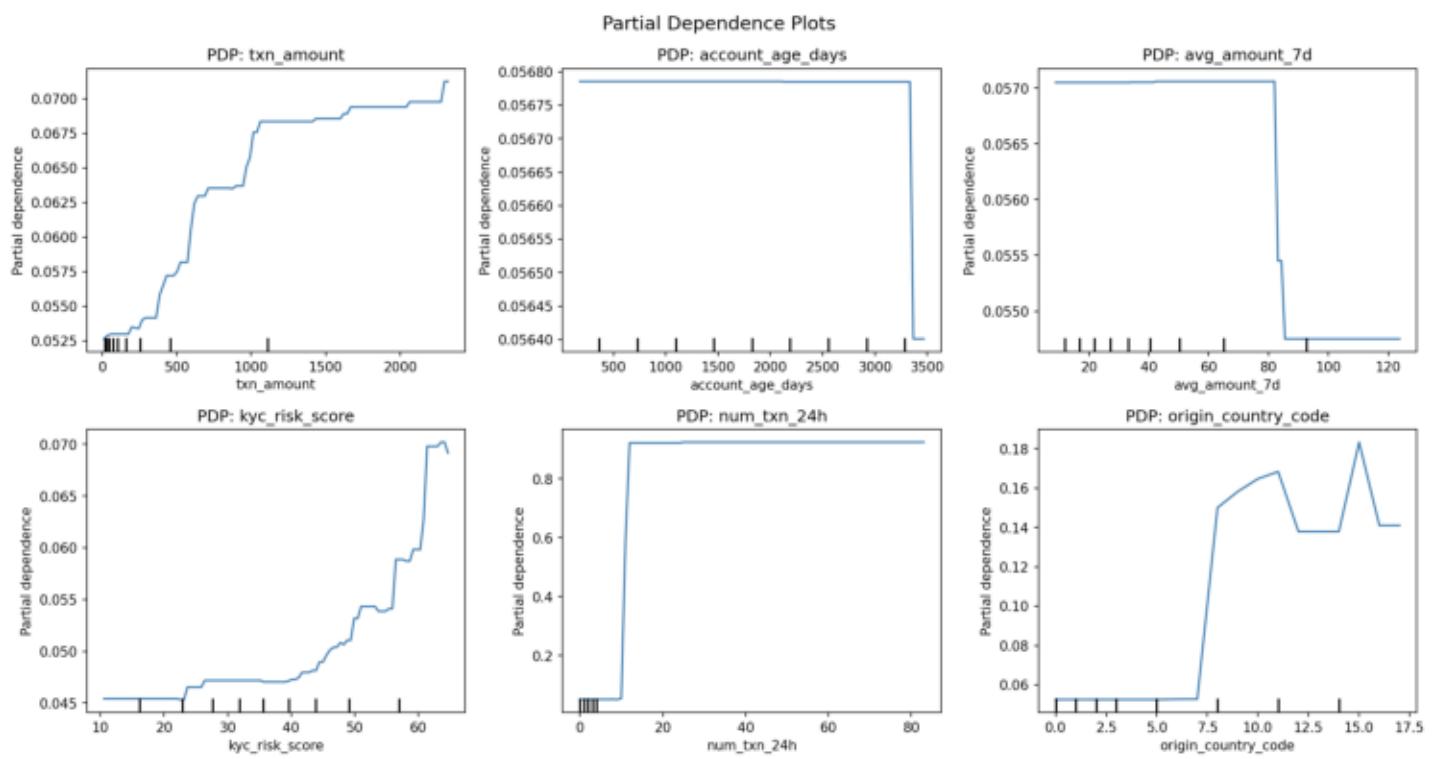
```
- model_type: tree
- methods_used: ['permutation', 'pdp', 'ice']
- shap_status: unavailable
- shap_reason: No module named 'shap'
- perm_top_features: ['origin_country_code', 'kyc_risk_score', 'num_txn_24h', 'currency_code', 'destination_country_code']
- pdp_features: ['txn_amount', 'account_age_days', 'avg_amount_7d', 'kyc_risk_score', 'num_txn_24h']...
- ice_features: ['txn_amount', 'account_age_days', 'avg_amount_7d', 'kyc_risk_score']
```

4) Model Interpretability - Explanations

- Permutation importance highlights: ['origin_country_code', 'kyc_risk_score', 'num_txn_24h', 'currency_code', 'destination_country_code'].
- SHAP plots not generated because SHAP is unavailable (No module named 'shap').
- PDP plots show average effects for: ['txn_amount', 'account_age_days', 'avg_amount_7d', 'kyc_risk_score', 'num_txn_24h', 'origin_country_code'].
- ICE plots show individual-level effects for: ['txn_amount', 'account_age_days', 'avg_amount_7d', 'kyc_risk_score'].

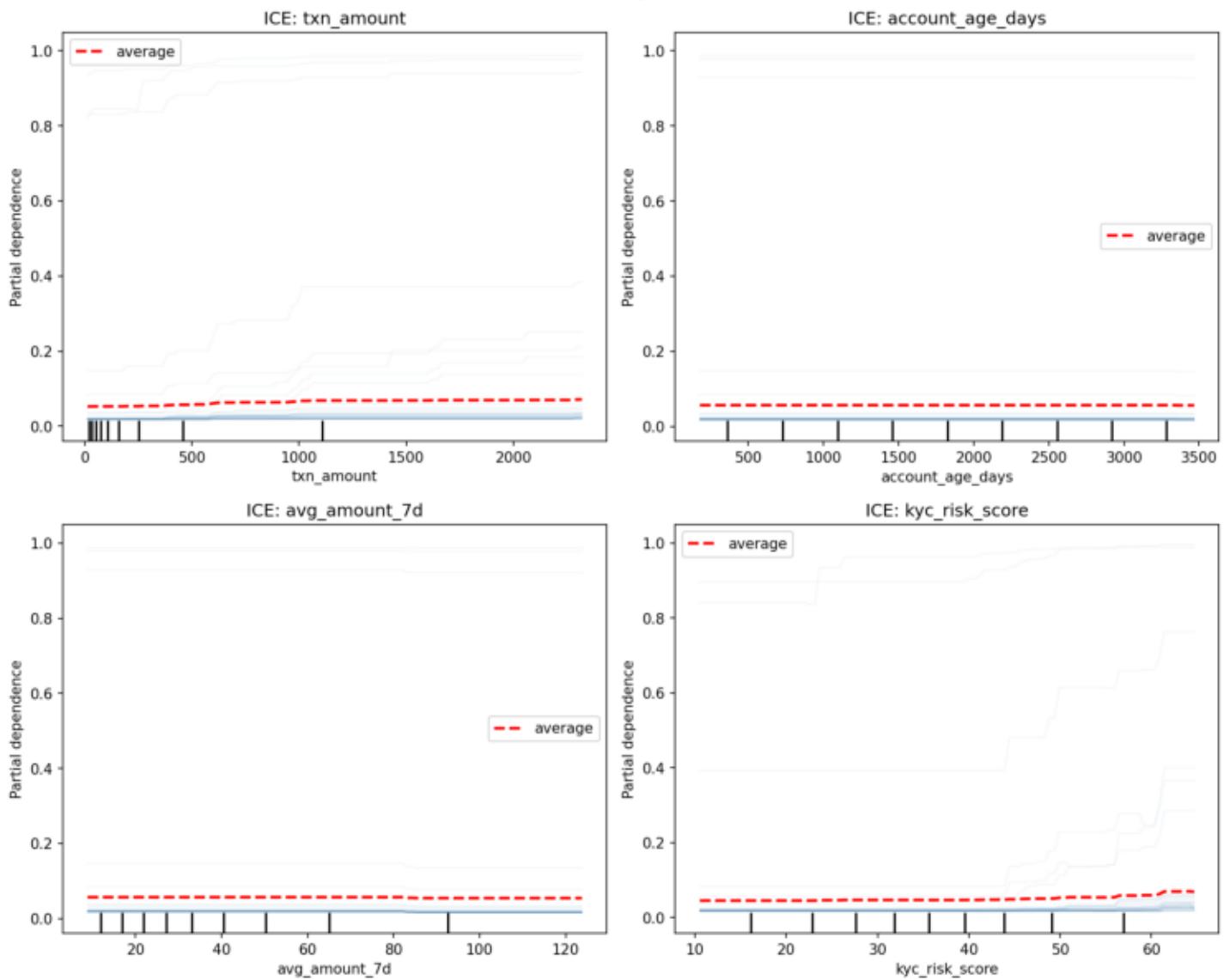


permutation_importance: Bars show how much AUC drops when each feature is permuted.



pdp: PDP curves show average model response as a feature varies.

Individual Conditional Expectation (ICE) Plots



ice: ICE curves show per-instance responses as a feature varies.