

# DA Internship - Final Project Presentation

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19th June - 16th August, 2023

## **OBJECTIVES:**

1. Compare behavior of Performing Loaners (PL) with Non-Performing Loaners (NPL)
2. Identify significant variables that differentiate NPLs from PLs

DT	AVG_12MTHS_FREK_DB	AVG_6MTHS_FREK_CR	FLAG_RESTRU	<del>6to3DB</del>	slopeFDB
ID_DEB	AVG_12MTHS_FREK_CR	AVG_3MTHS_CASA	FLAG_DEFERRED	<del>12to9FCR</del>	interceptFDB
DEBTOR_CATEGORY	AVG_9MTHS_CASA	AVG_3MTHS_DPK	REDFLAG_YELLOW	<del>9to6FCR</del>	12to3CR
PLAFOND_IDR	AVG_9MTHS_DPK	AVG_3MTHS_AMT_DB	REDFLAG_RED	<del>6to3FCR</del>	12to3DB
OS_IDR	AVG_9MTHS_AMT_DB	AVG_3MTHS_AMT_CR	REDFLAG_INFORMASI	<del>12to9FDB</del>	12to3FCR
STRT_PAST_DUE_DT	AVG_9MTHS_AMT_CR	AVG_3MTHS_FREK_DB	FLAG_BLACKLIST	<del>9to6FDB</del>	12to3FDB
HARI_TUNGGAHAN	AVG_9MTHS_FREK_DB	AVG_3MTHS_FREK_CR	percent_used	<del>6to3FDB</del>	
CUST_TYPE_CD	AVG_9MTHS_FREK_CR	SALDO_AVG_CASA	total_flags	slopeCR	
COLLECT_KEY	AVG_6MTHS_CASA	SALDO_AVG_DPK	<del>12to9CR</del>	interceptCR	
AVG_12MTHS_CASA	AVG_6MTHS_DPK	AVG_MUTASI_DB	<del>9to6CR</del>	slopeDB	
AVG_12MTHS_DPK	AVG_6MTHS_AMT_DB	AVG_MUTASI_CR	<del>6to3CR</del>	interceptDB	
AVG_12MTHS_AMT_DB	AVG_6MTHS_AMT_CR	FREK_DB	<del>12to9DB</del>	slopeFCR	
AVG_12MTHS_AMT_CR	AVG_6MTHS_FREK_DB	FREK_CR	<del>9to6DB</del>	interceptFCR	

## STEPS

1. ~~remove unwanted variables~~
2. create new variables
  - a.  $\text{percent\_used} = \text{OS\_IDR} / \text{PLAFOND\_IDR}$
  - b.  $\text{total\_flags} = \text{REDFLAG\_YELLOW} + 3(\text{REDFLAG\_RED}) + 5(\text{REDFLAG\_INFORMASI}) + 7(\text{FLAG\_BLACKLIST})$
  - c.  $12\text{to}9, 9\text{to}6, 6\text{to}3 = (9-12)/12, (6-9)/9, (3-6)/6$
  - d. slope, intercept = derived from linear regression with values from c.
  - e.  $12\text{to}3 = (3-12)/12$
3. ~~remove unwanted variables~~ (use correlation matrix to determine (pairs = over 90% correlation))

DATA PREPARATION

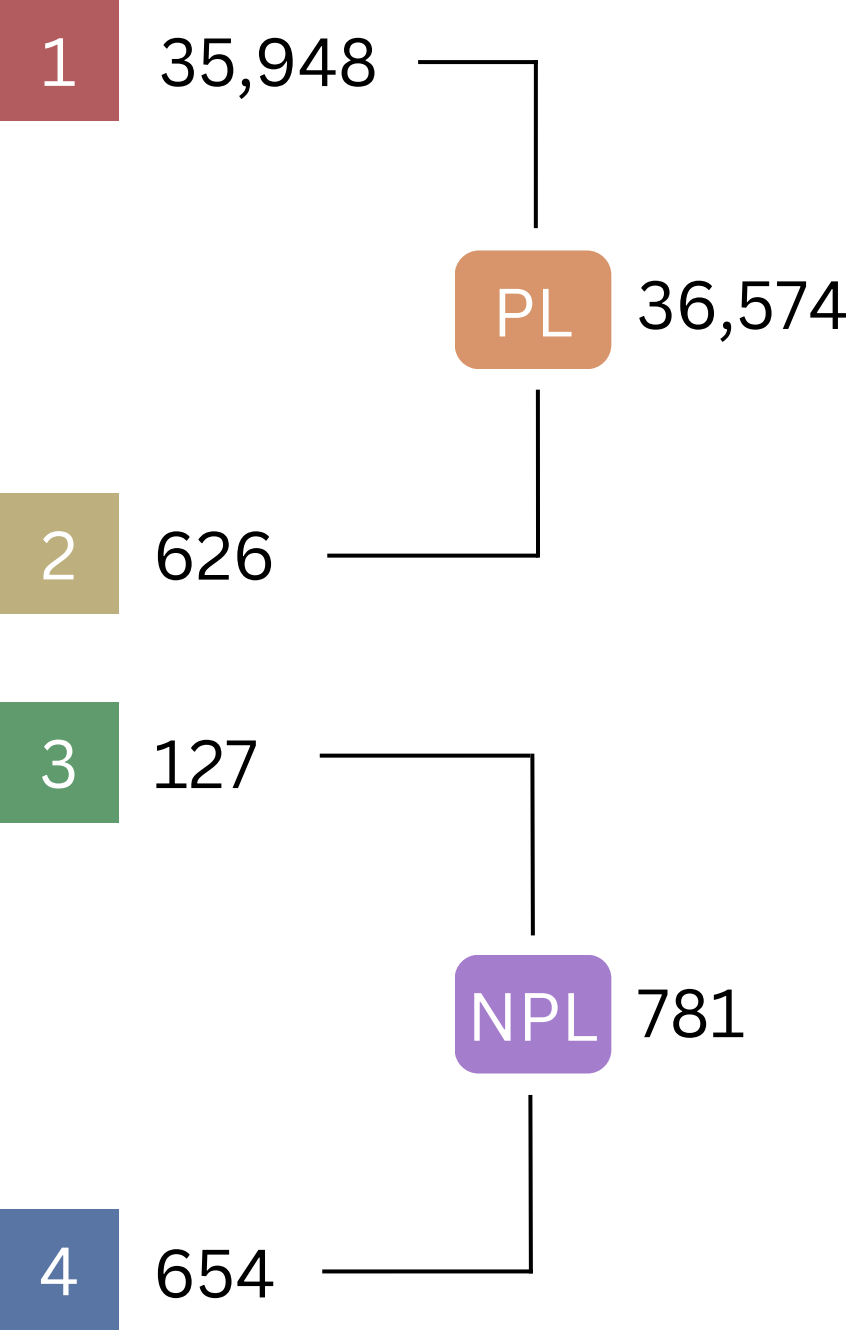
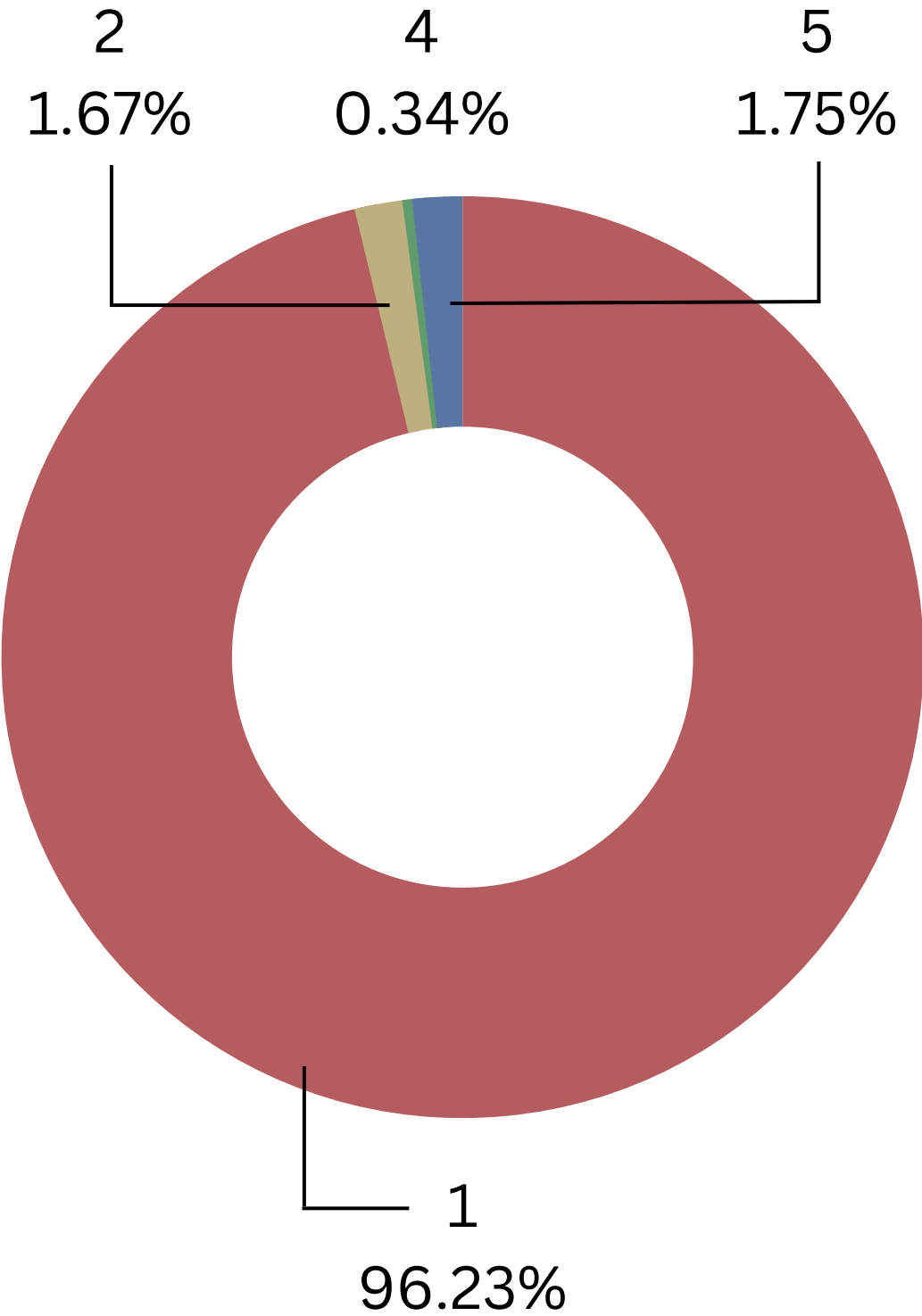
# Loaners Breakdown

- PL

Performing Loaners
- NPL

Non-Performing Loaners

## Collect Key



## Customer Type

Organization: 41%  
Individual: 59%

## Debtor Category

SME: 93.9%  
Kommersial: 6.9%

## Customer Type

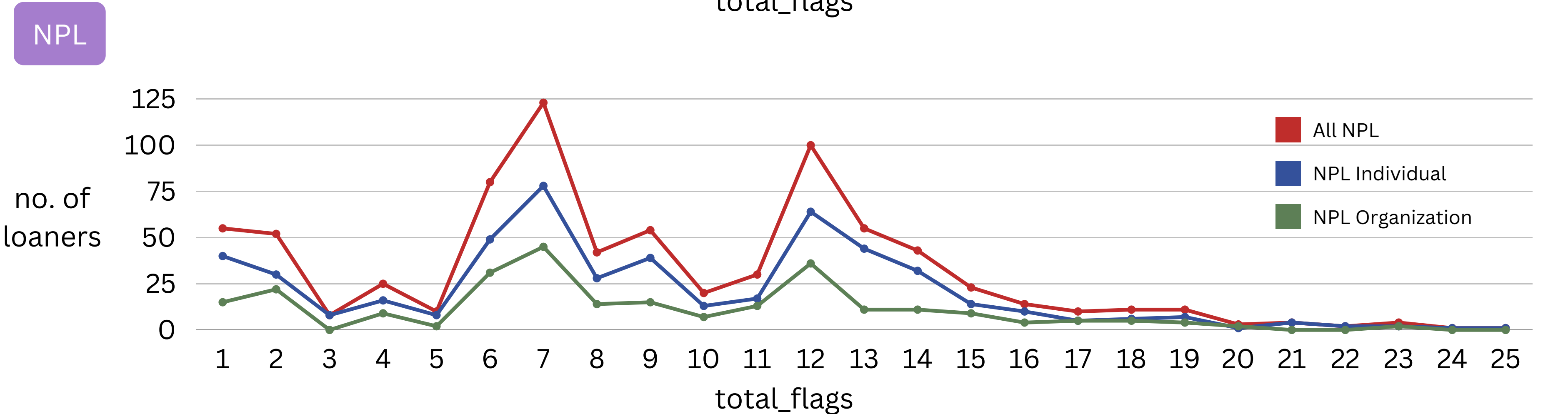
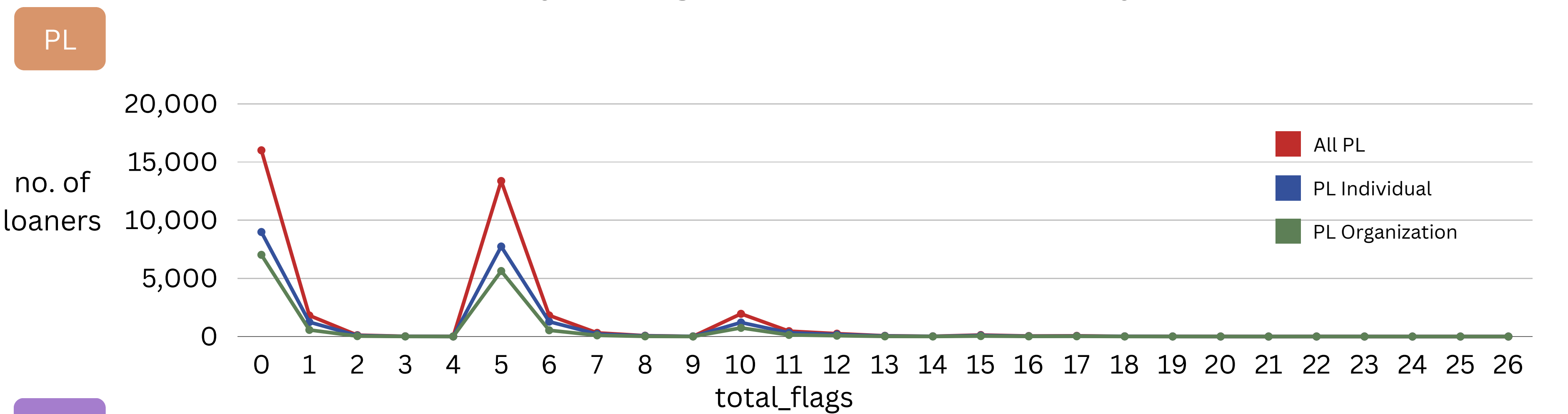
Organization: 33.5%  
Individual: 66.5%

## Debtor Category

SME: 92.8%  
Kommersial: 7.2%

\*not significant

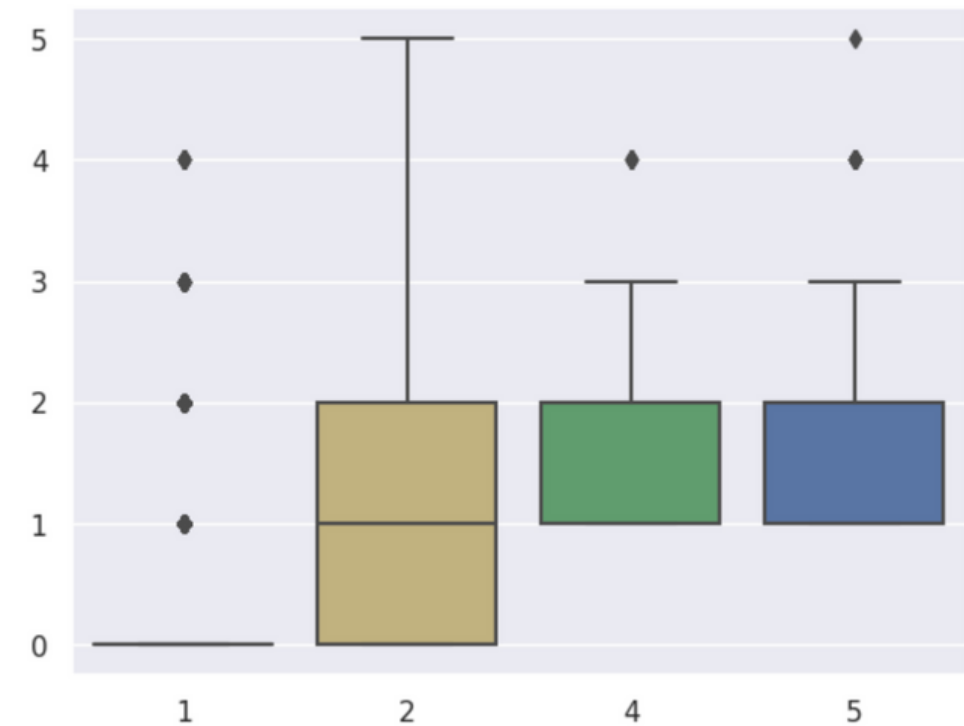
# Frequency of Flags (Compare Customer Type)



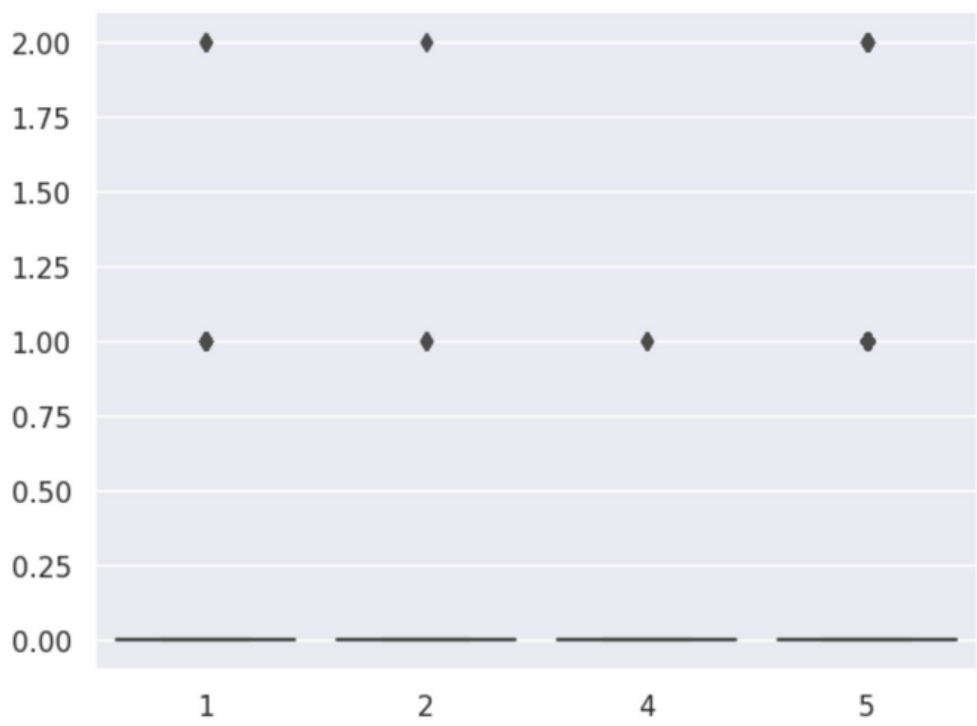
# Loaners Category vs No. of Flags

x-axis: no. of flags  
y-axis: collect\_key

yellow

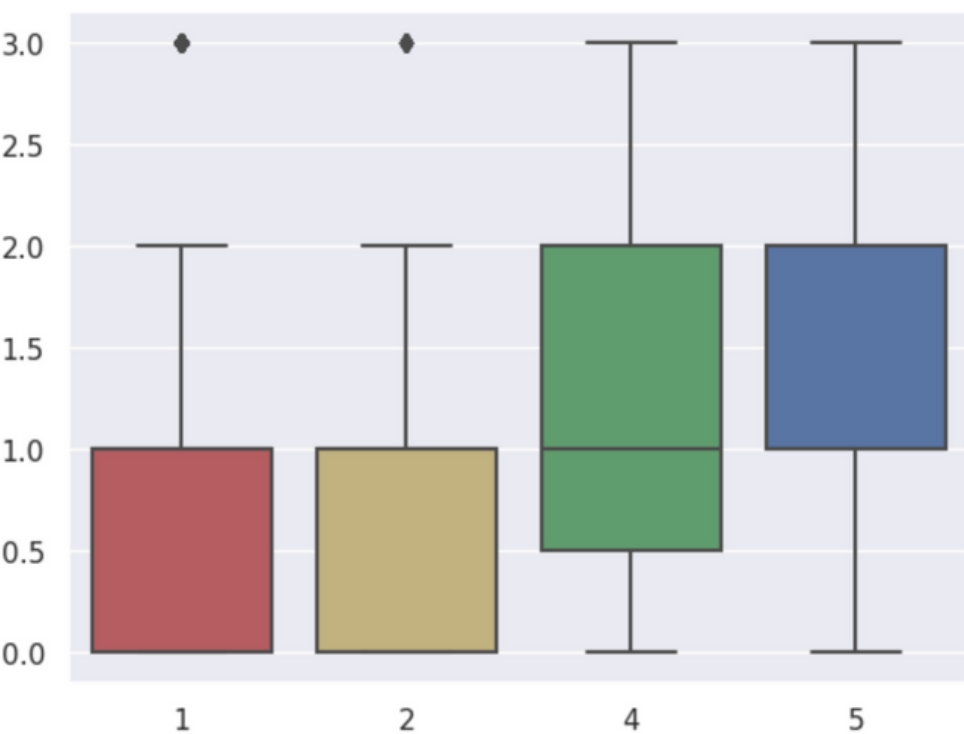


red

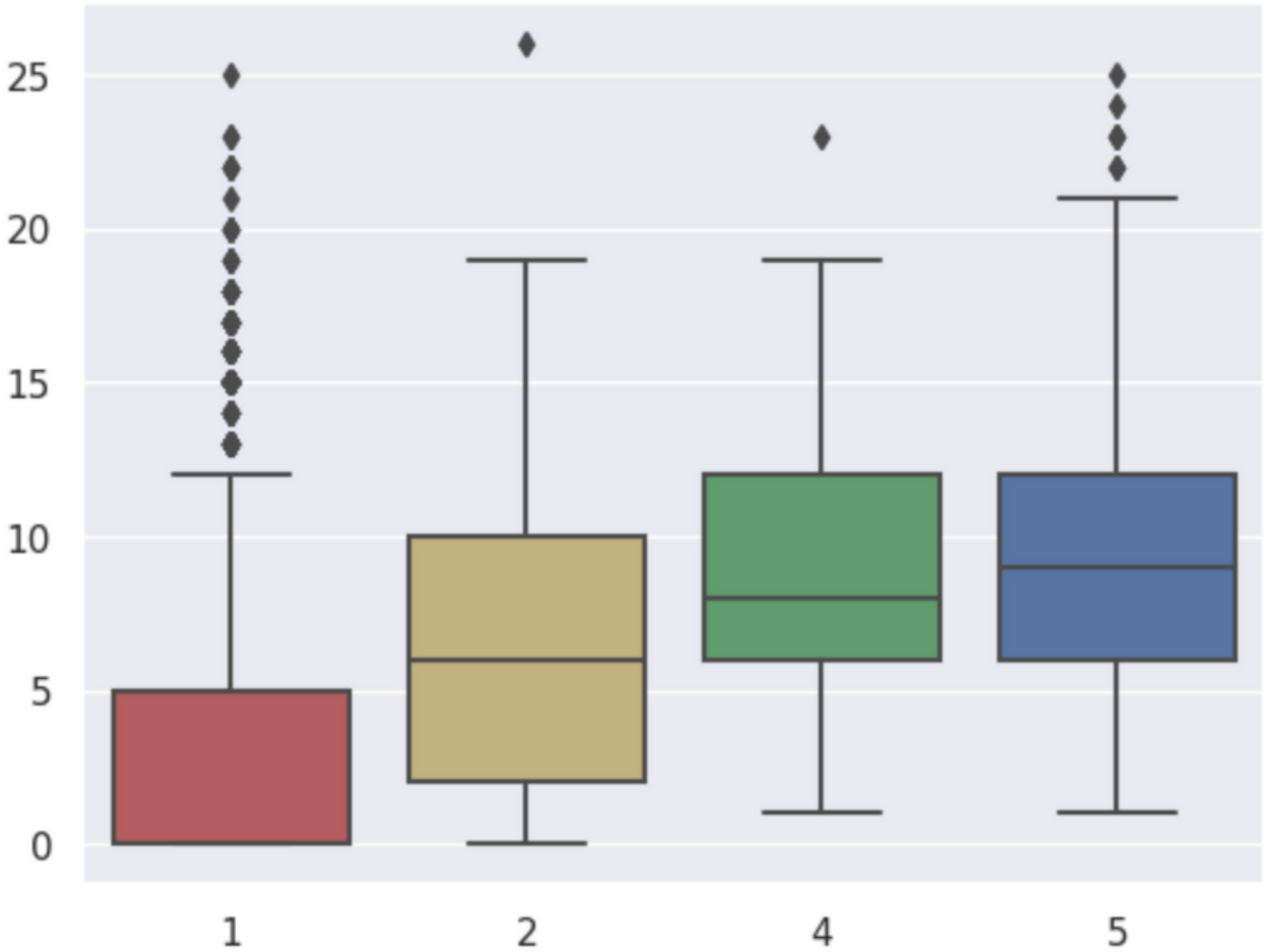
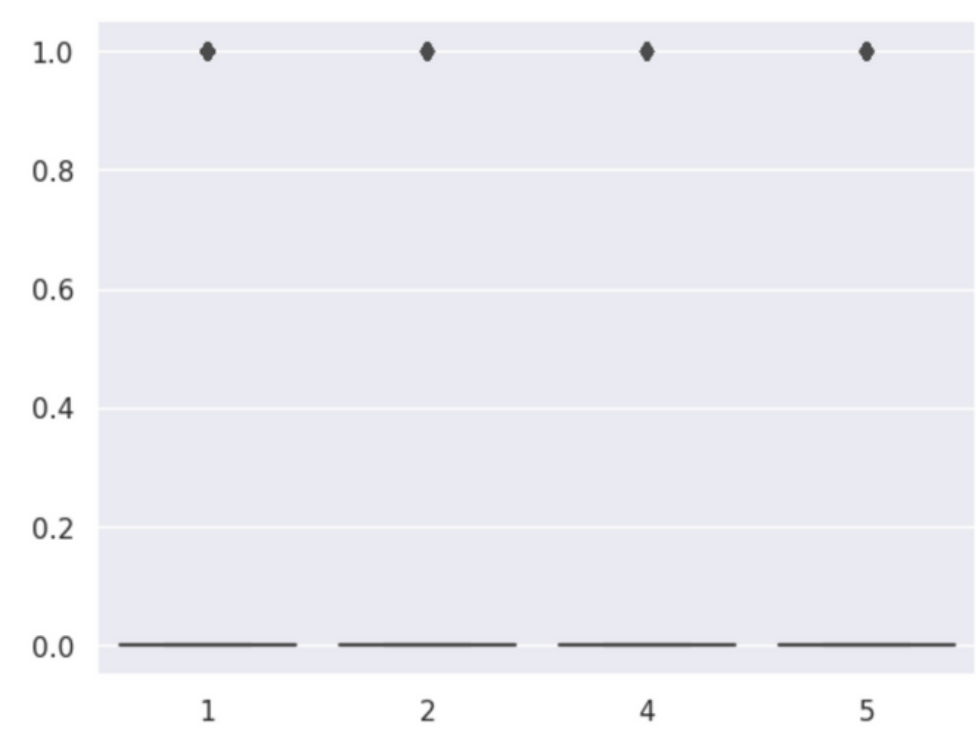


total flags = yellow + 3(red) + 5(informasi) + 7(blacklist)

informasi



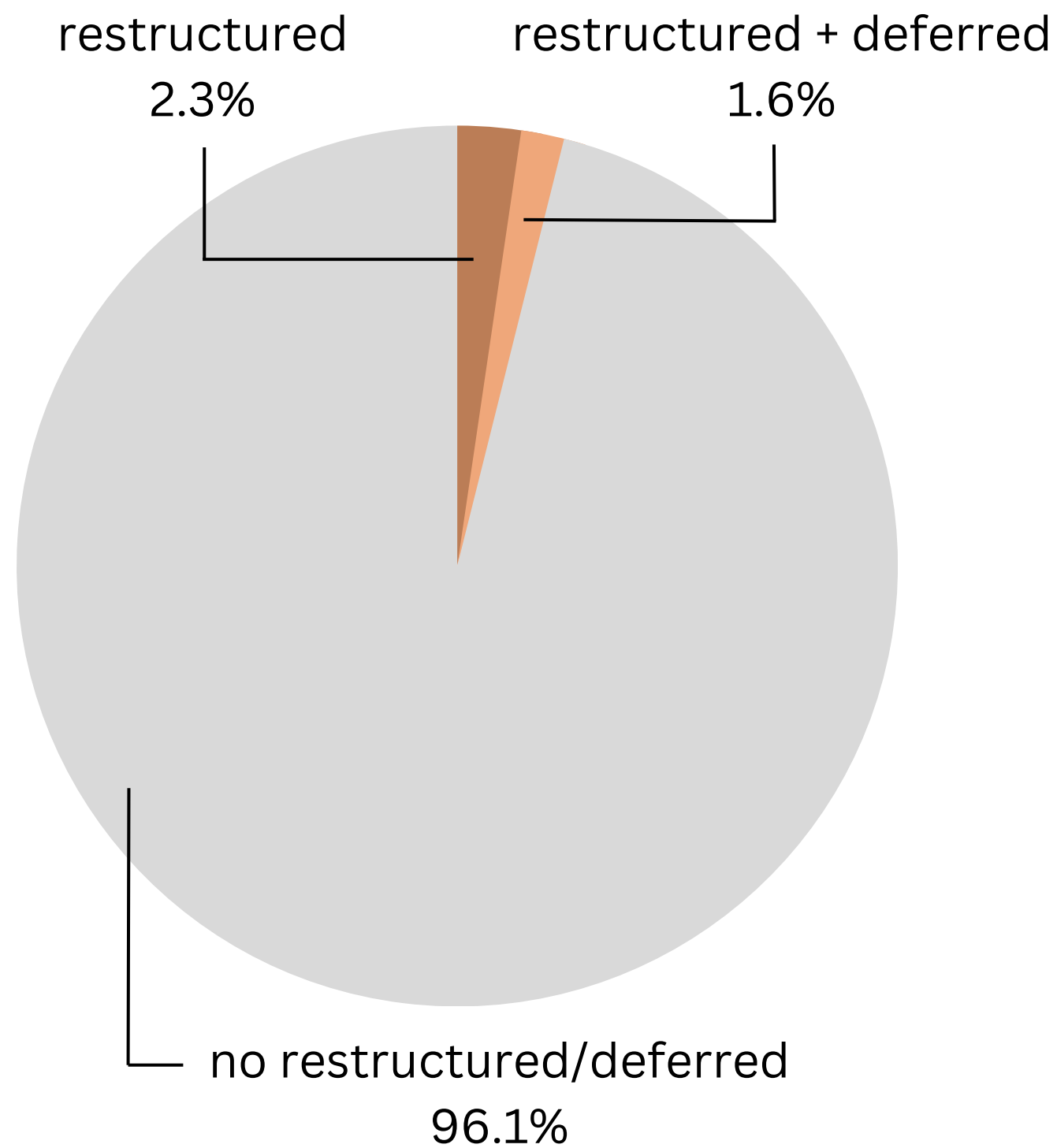
blacklist



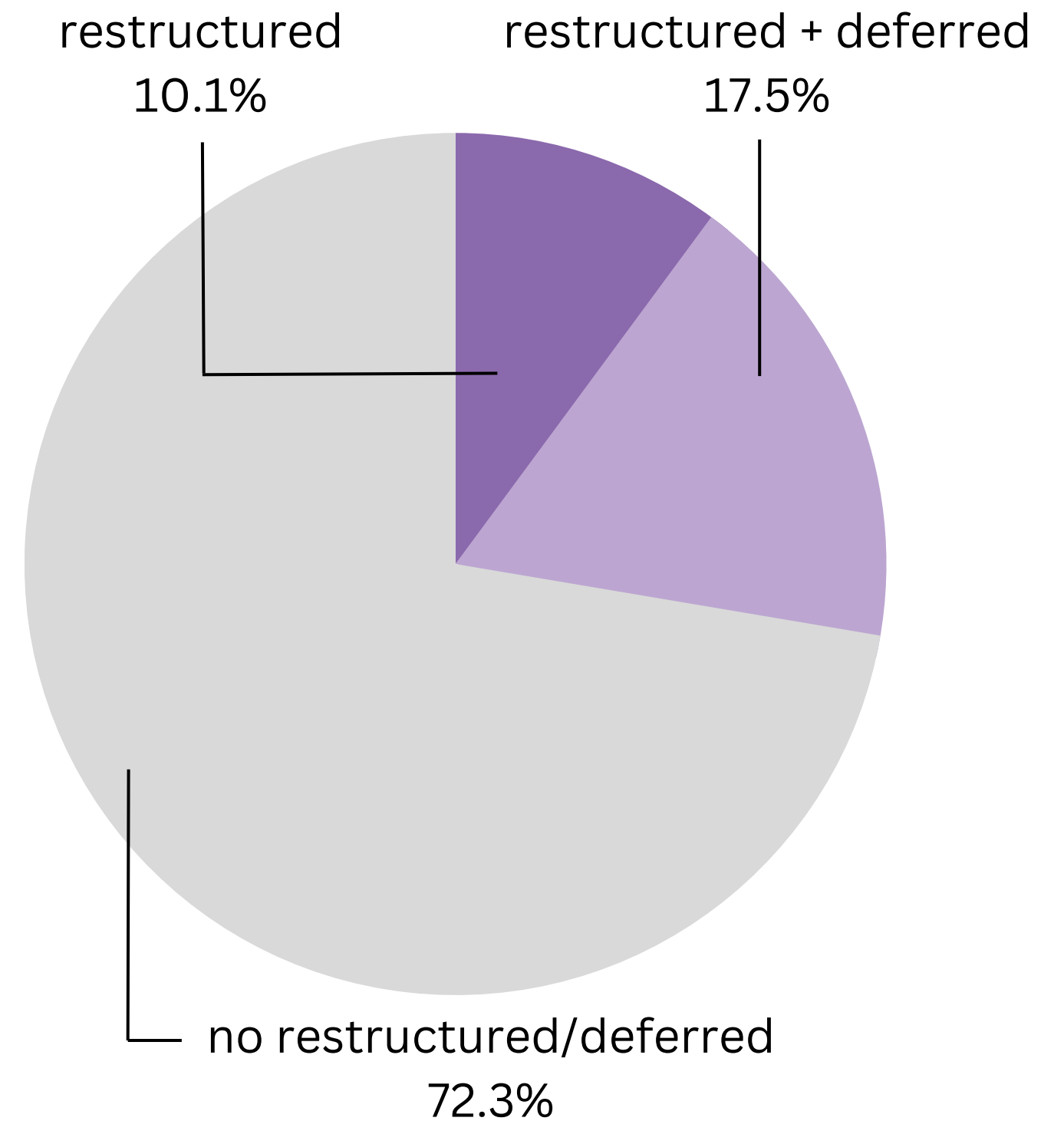
\*yellow, informasi, total significant

## Restructured / Deferred Analysis

PL



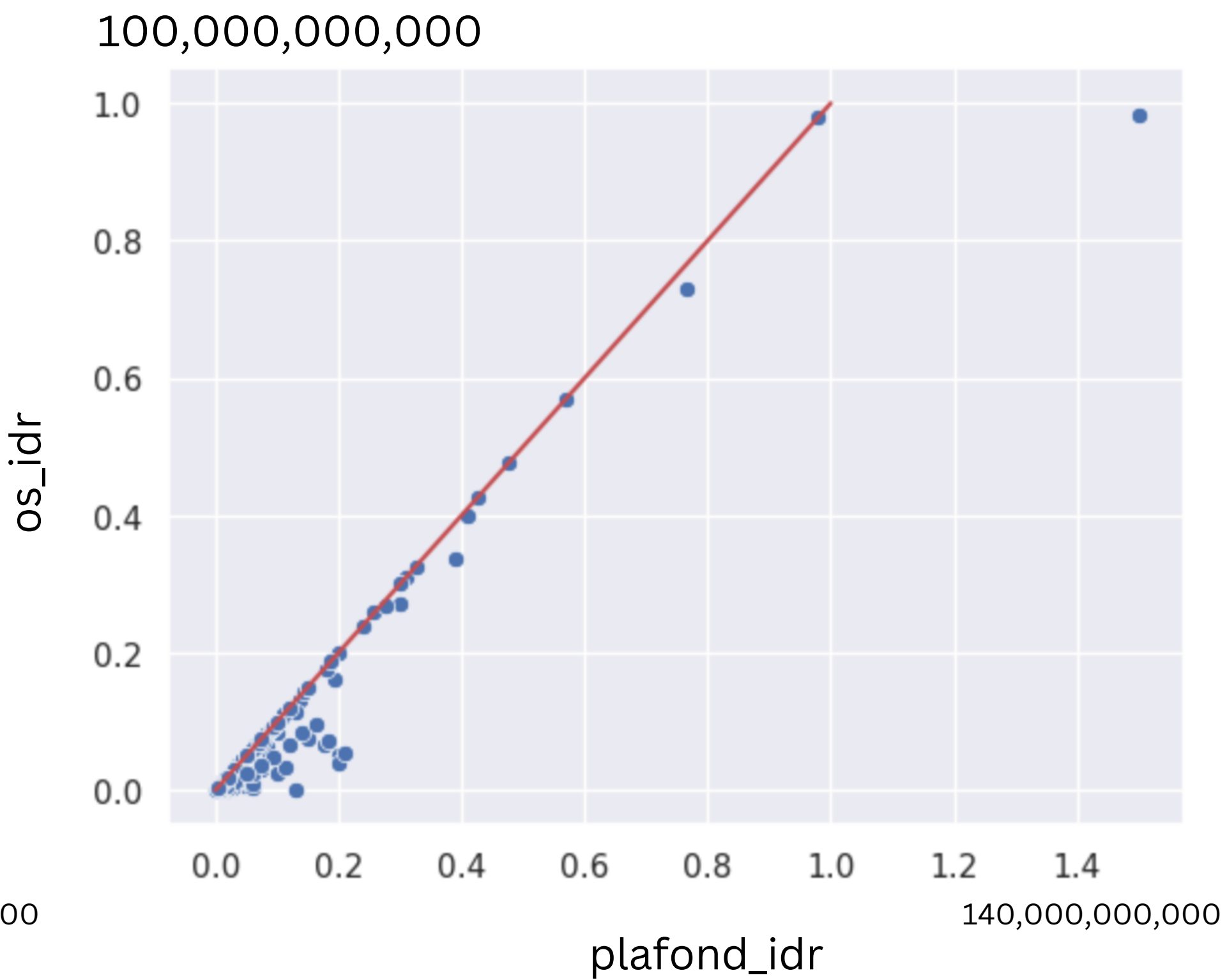
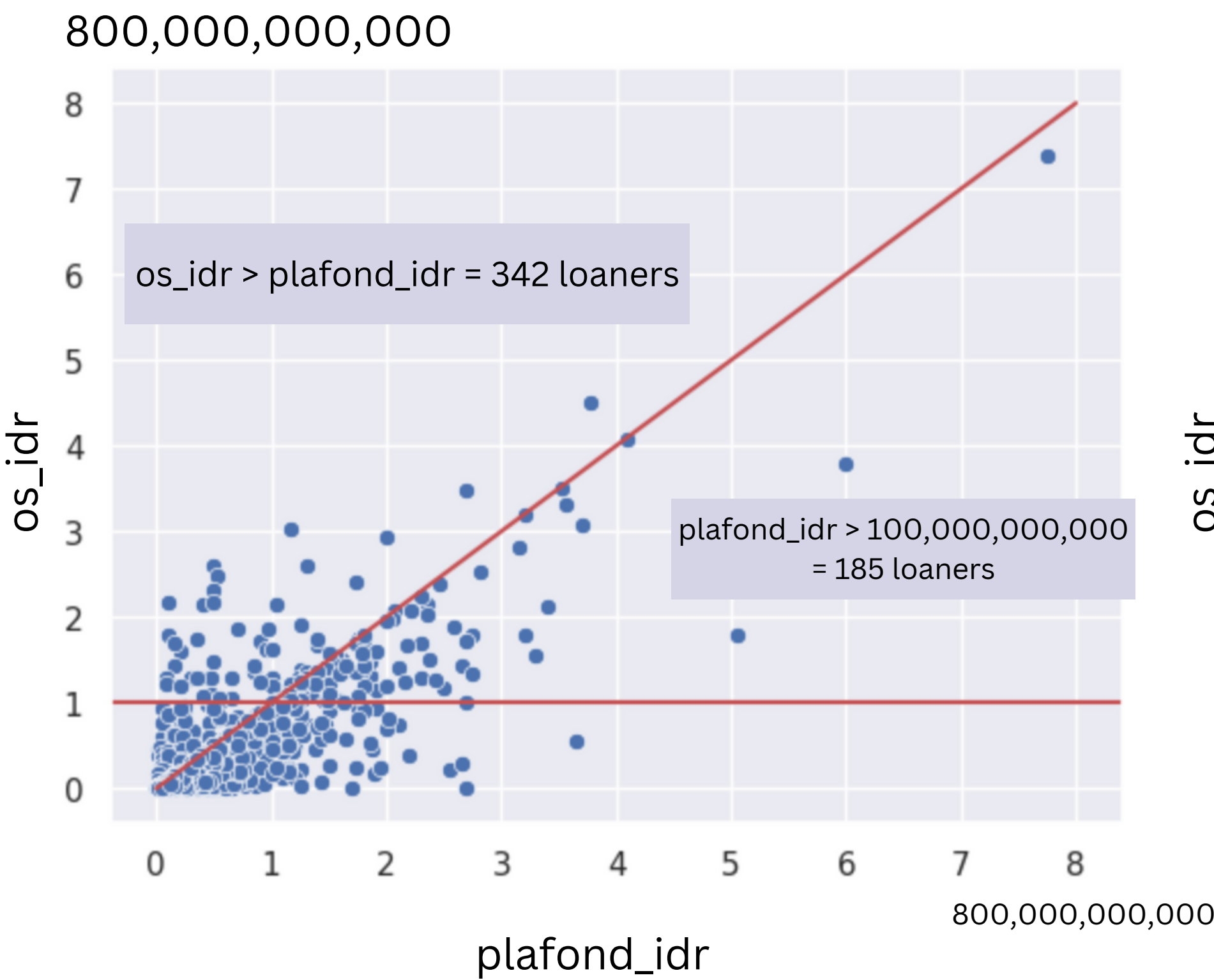
NPL



# Plafond vs OS

PL (36,574 loaners)

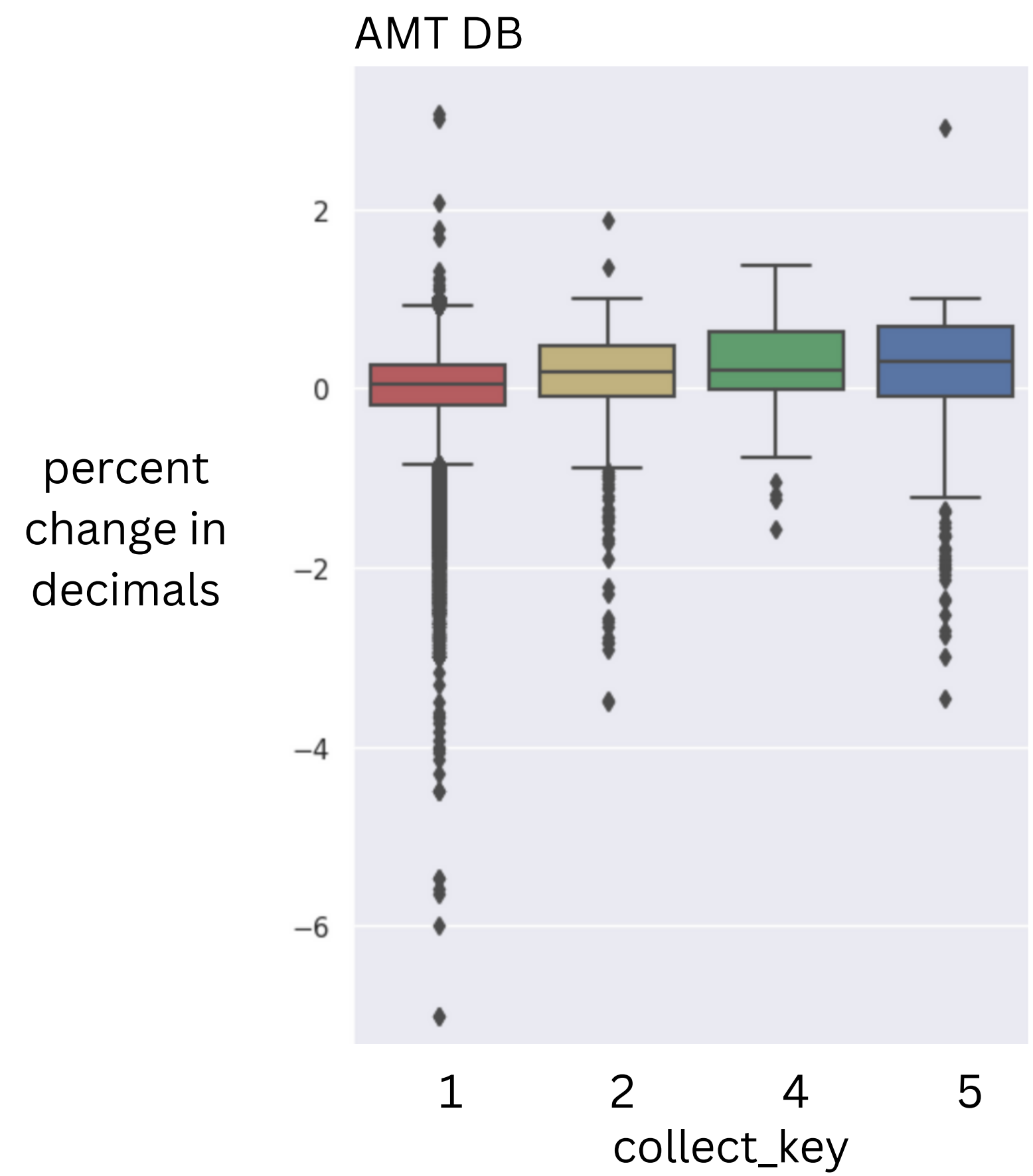
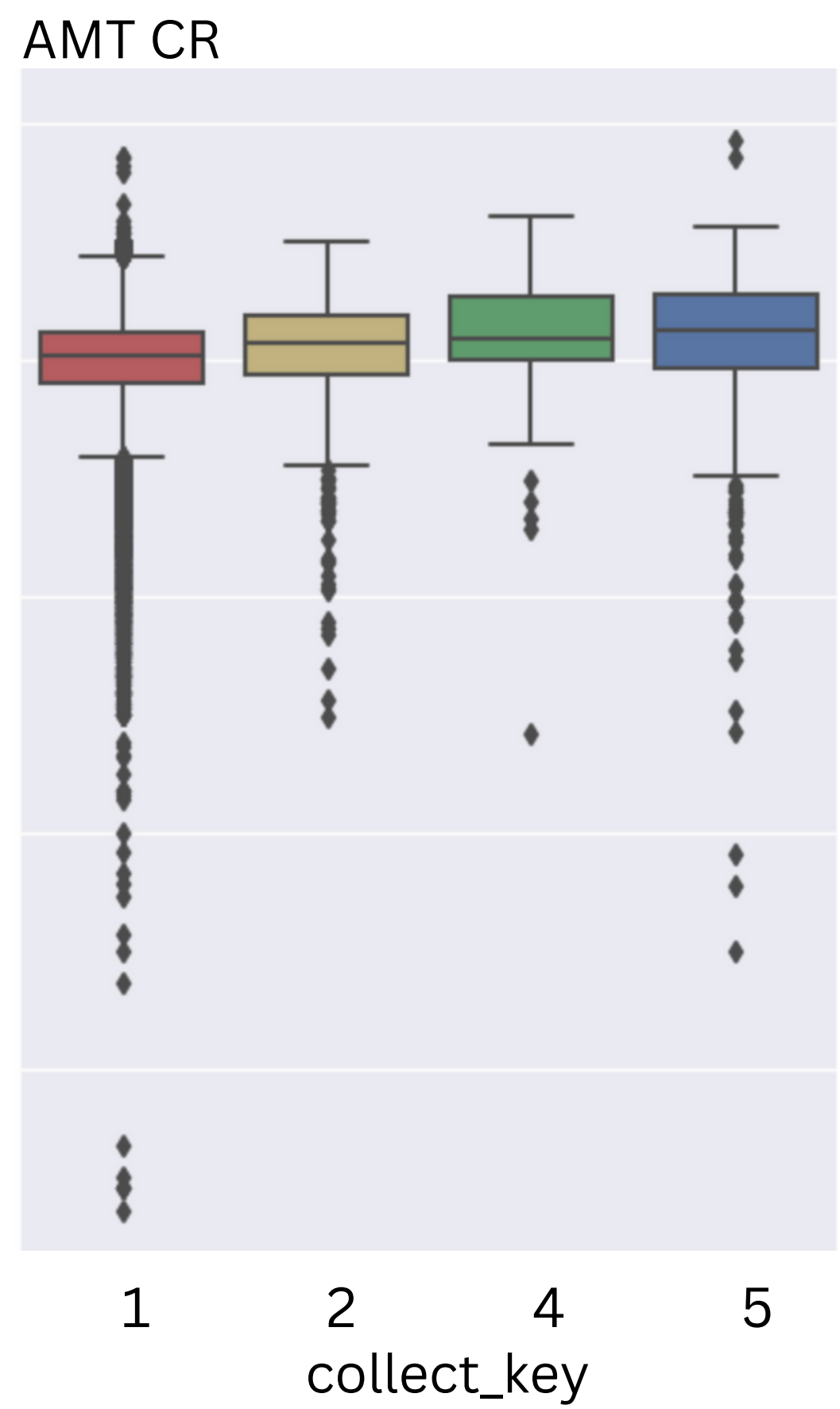
NPL (781 loaners)



\*significant

# Change in Transaction Behavior (transaction amount)

Y-axis =  $\frac{3\text{MTHS} - 12\text{MTHS}}{12\text{MTHS}}$



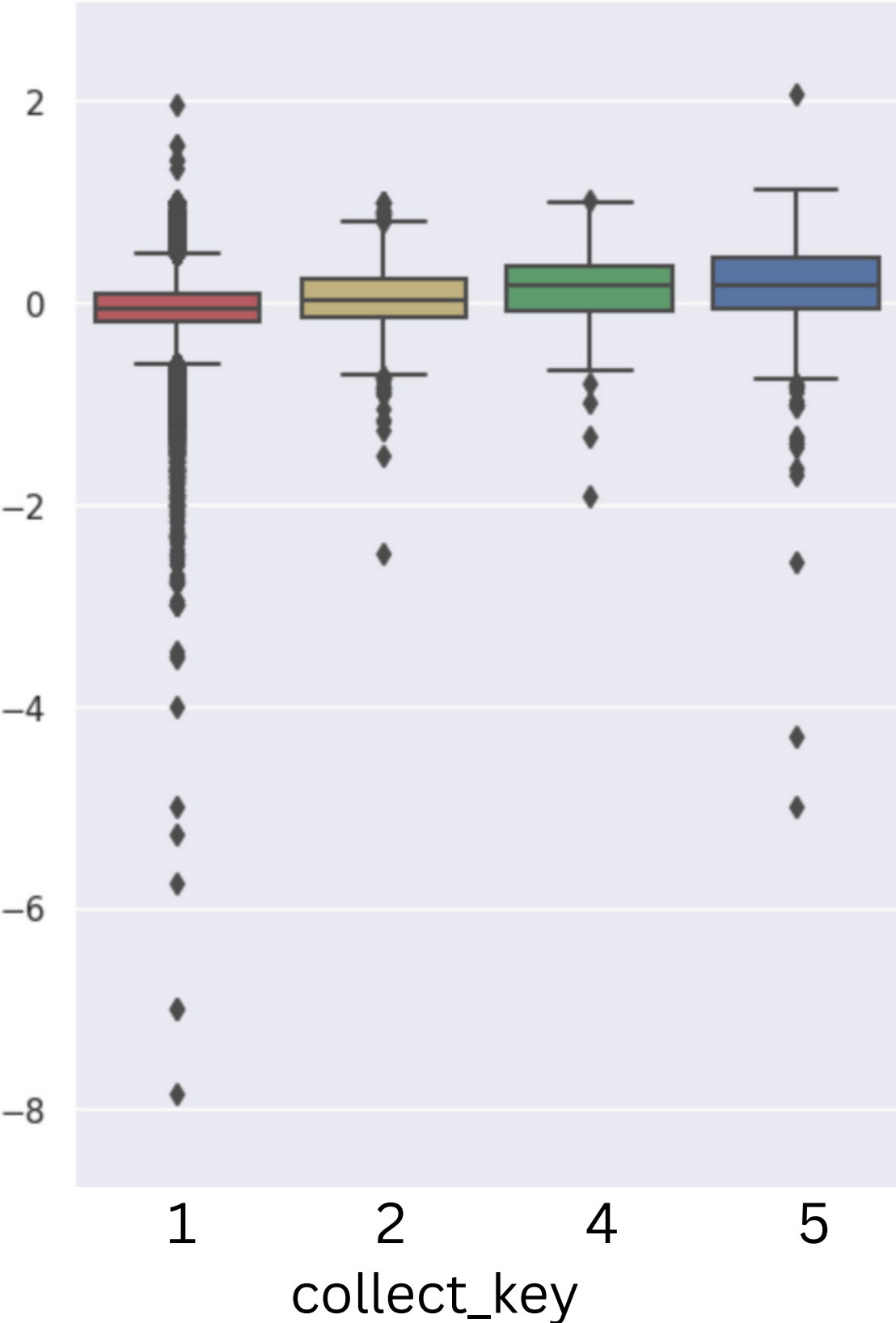


\*not significant

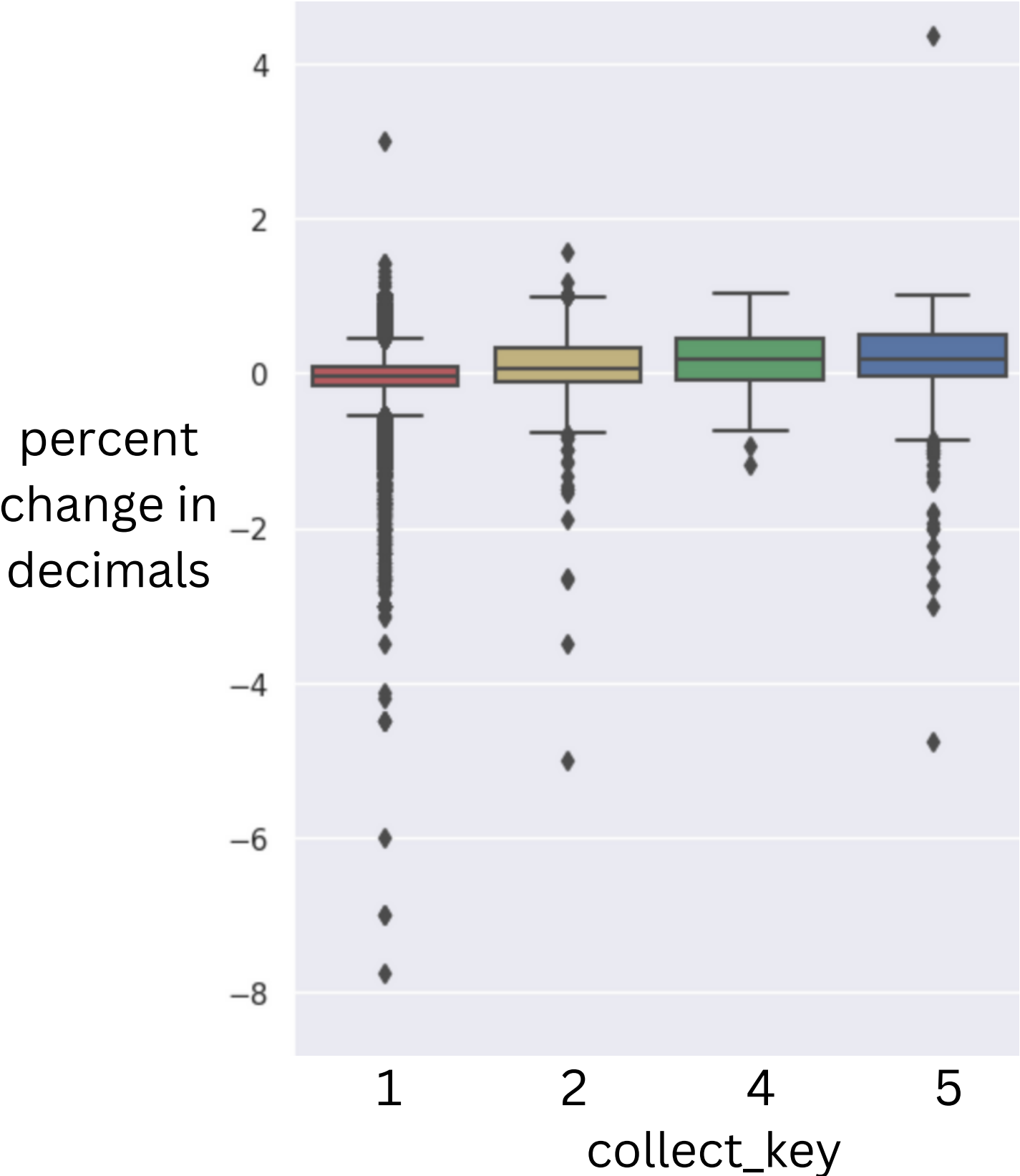
# Change in Transaction Behavior (transaction frequency)

Y-axis =  $\frac{3\text{MTHS} - 12\text{MTHS}}{12\text{MTHS}}$

FREK CR



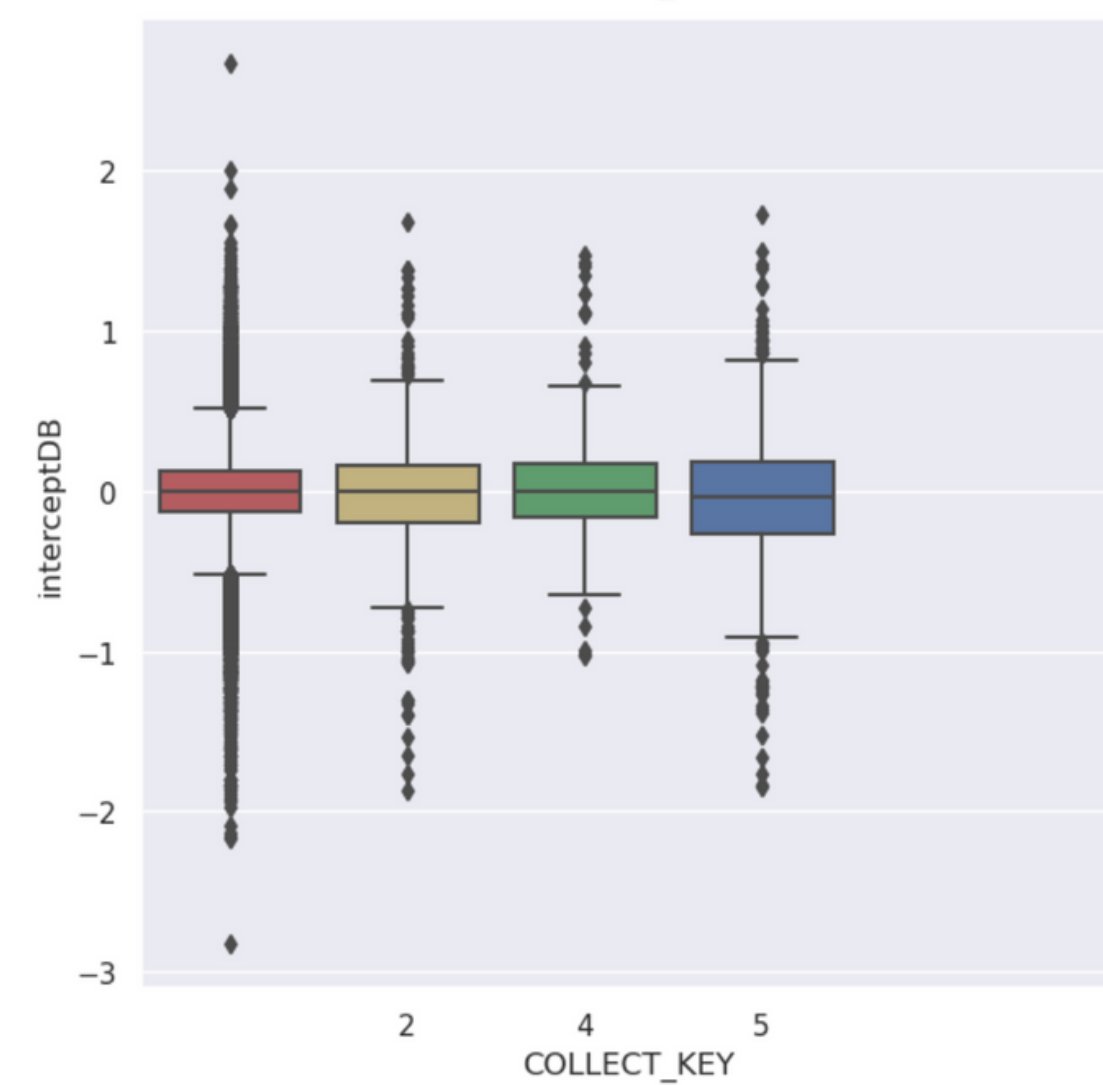
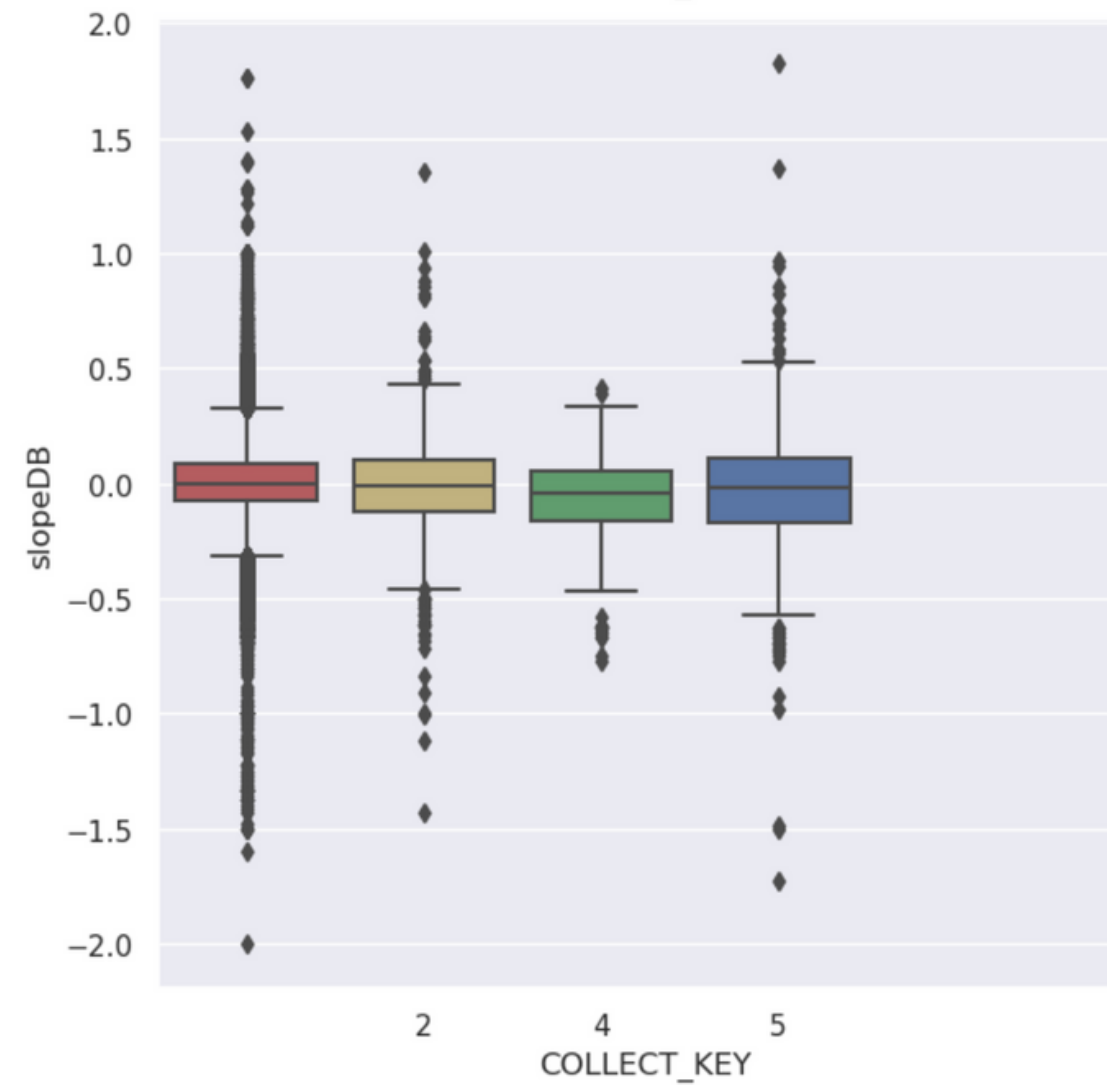
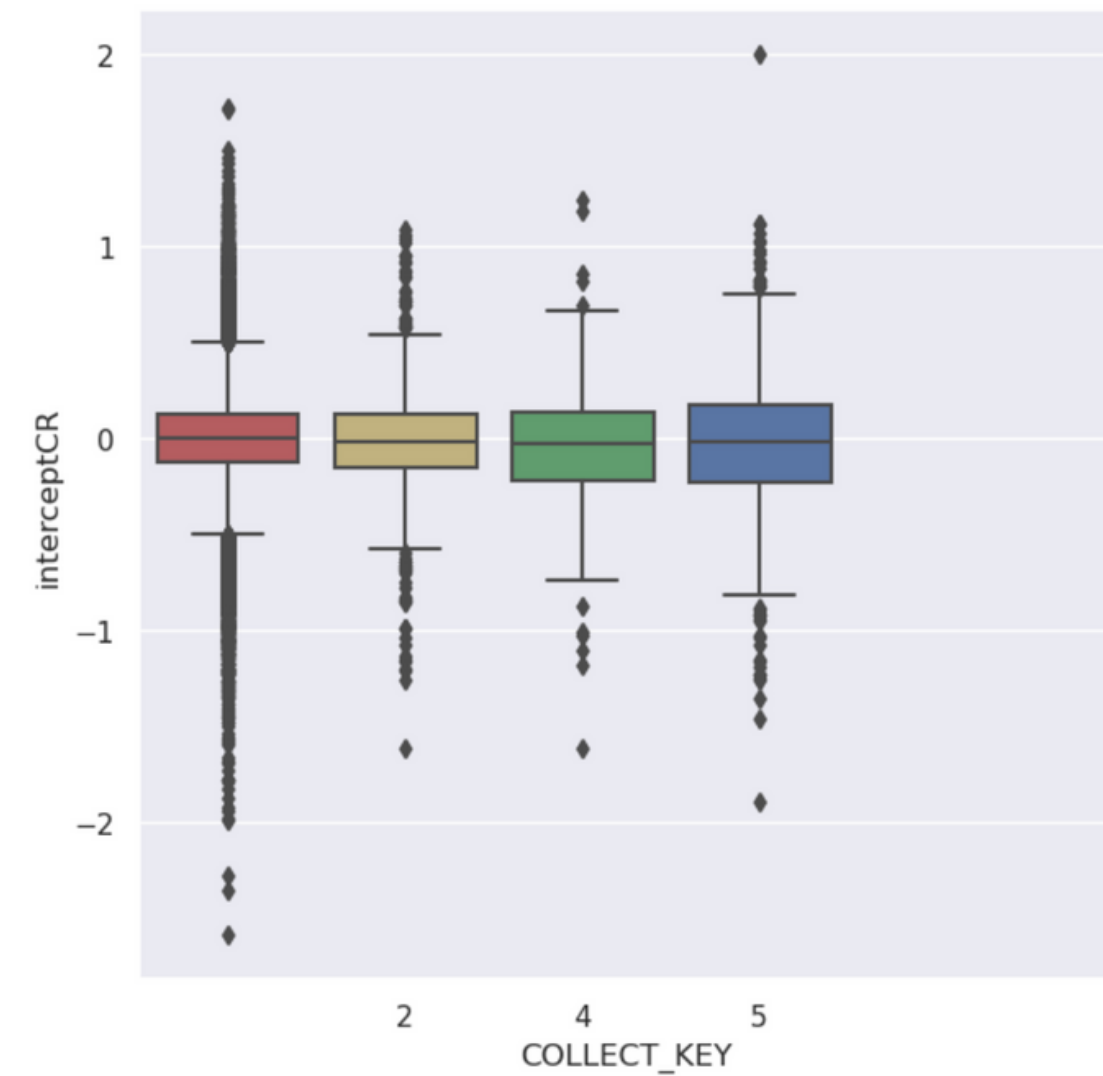
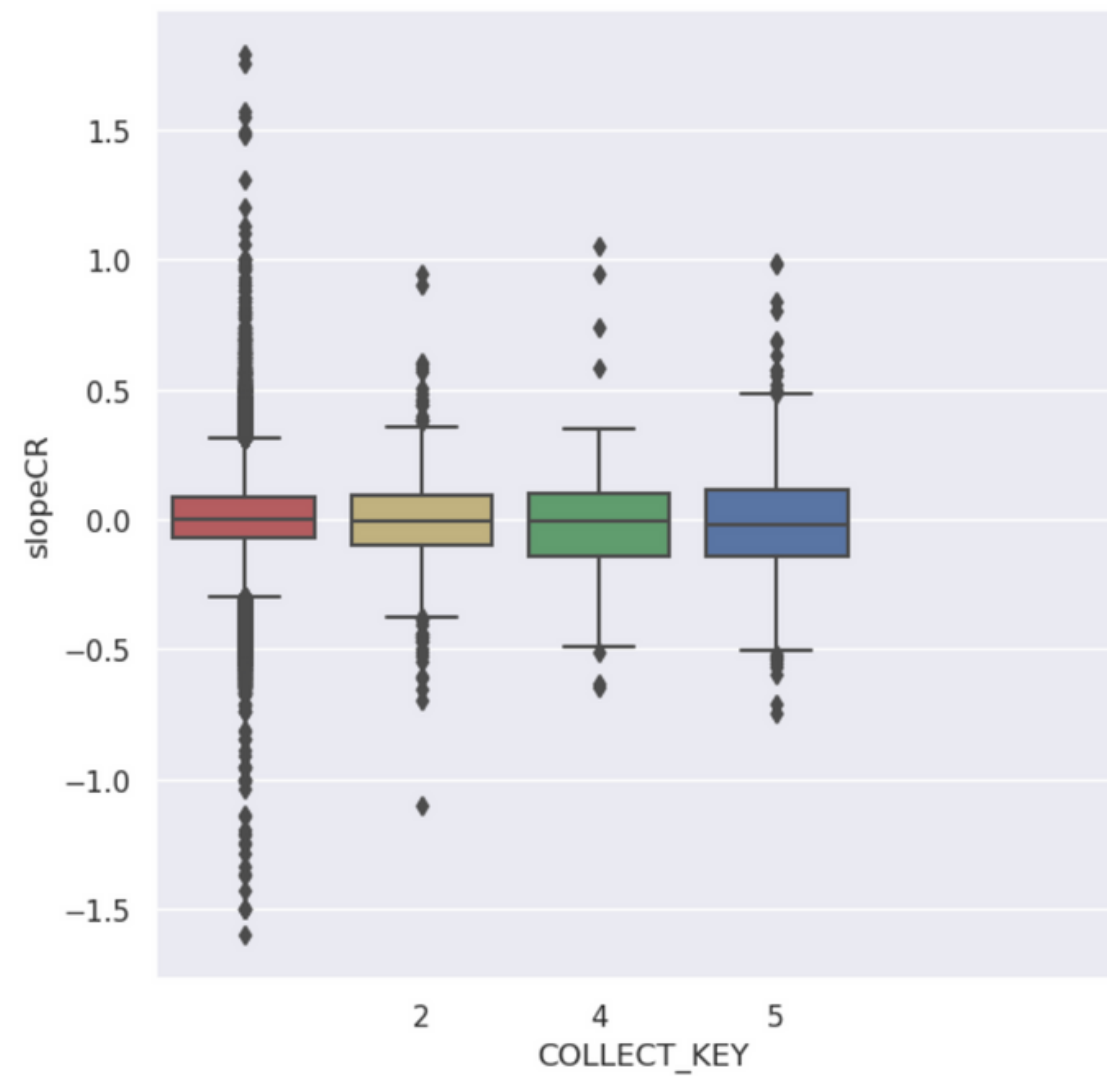
FREK DB



12to3CR						12to3DB					
key		1	2	4	5	key		1	2	4	5
statistic	count	35948	626	127	654	statistic	count	35948	626	127	654
	mean	-0.02079	0.062522	0.176803	0.13867		mean	-0.019688	0.109433	0.265653	0.208409
	std	0.482124	0.550824	0.555789	0.690041		std	0.548806	0.711982	0.523027	0.749481
	min	-7.194555	-3.009541	-3.164119	-5		min	-16.2	-7.524561	-1.574157	-3.469467
	25%	-0.189766	-0.131165	-0.000072	-0.070857		25%	-0.187506	-0.088648	-0.003747	-0.09247
	50%	0.031207	0.143814	0.186785	0.244202		50%	0.041446	0.184205	0.211193	0.300917
	75%	0.23595	0.382989	0.528213	0.552754		75%	0.254321	0.466906	0.628686	0.687958
	max	1.720222	1	1.20793	1.862868		max	3.063641	1.887009	1.367719	7.683775
12to3FCR						12to3FDB					
key		1	2	4	5	key		1	2	4	5
statistic	count	35948	626	127	654	statistic	count	35948	626	127	654
	mean	-0.069162	0.036091	0.129803	0.151356		mean	-0.057772	0.070088	0.203766	0.185612
	std	0.32927	0.370483	0.431738	0.512084		std	0.358161	0.526085	0.425269	0.595897
	min	-7.854167	-2.479029	-1.909091	-5		min	-16.2	-5	-1.185792	-4.76
	25%	-0.192575	-0.145089	-0.07009	-0.049405		25%	-0.171007	-0.123193	-0.084326	-0.043226
	50%	-0.054093	0.030865	0.167286	0.168448		50%	-0.04	0.049745	0.176471	0.185108
	75%	0.078534	0.238092	0.356471	0.4375		75%	0.079086	0.315508	0.44535	0.505422
	max	1.969388	1	1.017857	2.08		max	3	1.56134	1.027513	4.375

\*not significant

Slope + Intercept of  
amount and frequency  
(CR + DB)

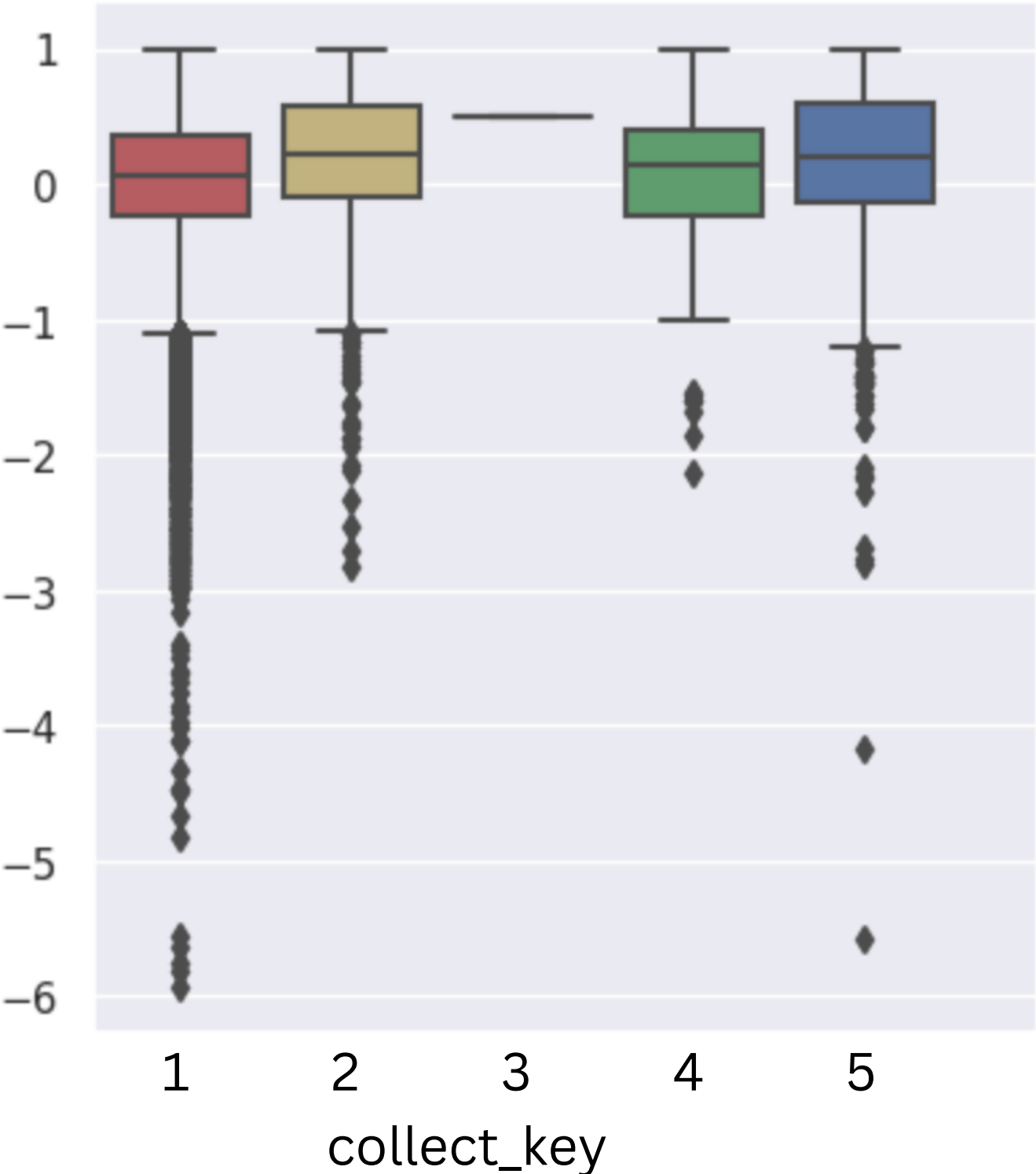


extra

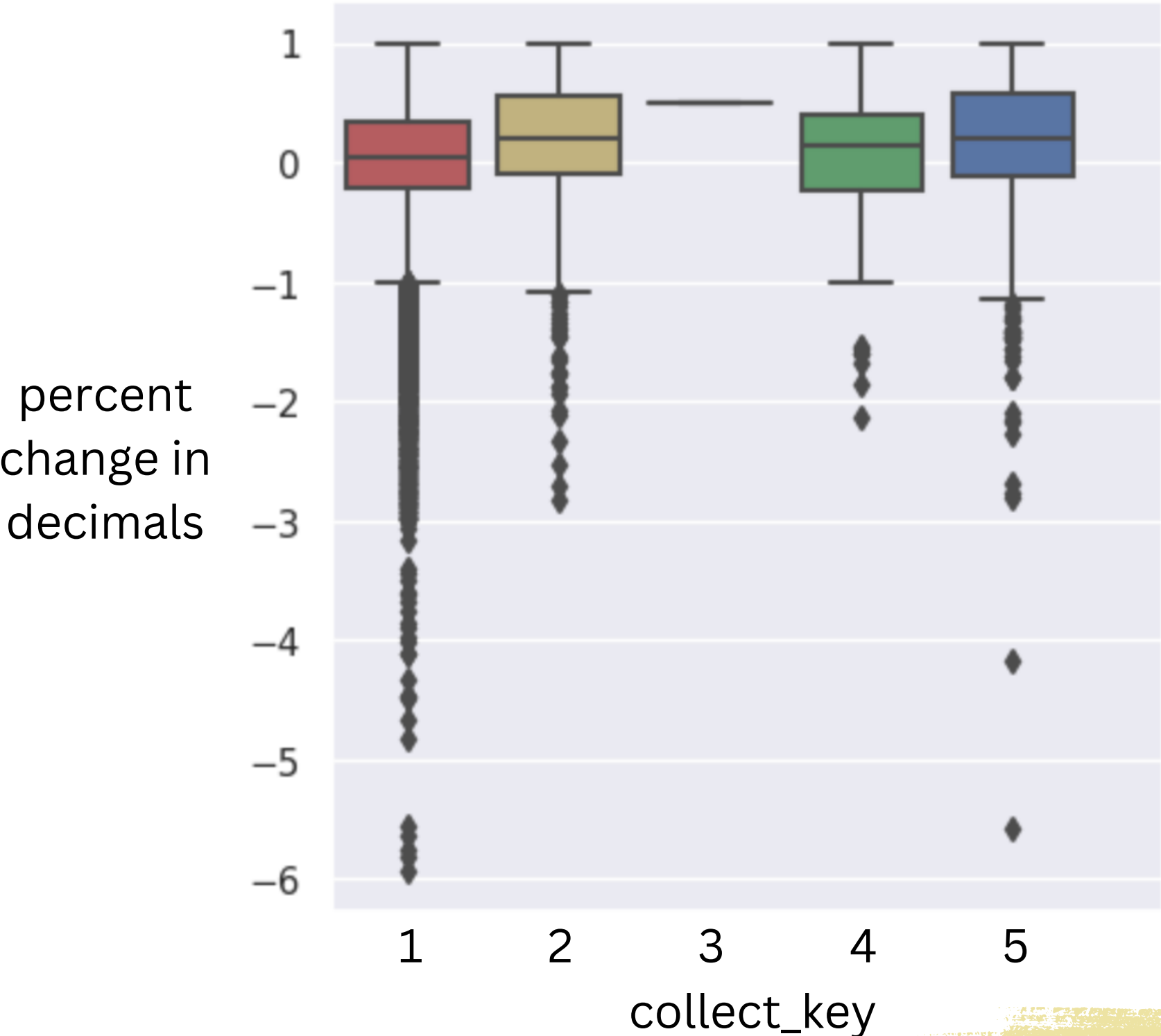
# Additional Analysis

Y-axis =  $\frac{3\text{MTHS} - 12\text{MTHS}}{12\text{MTHS}}$

CASA



DPK

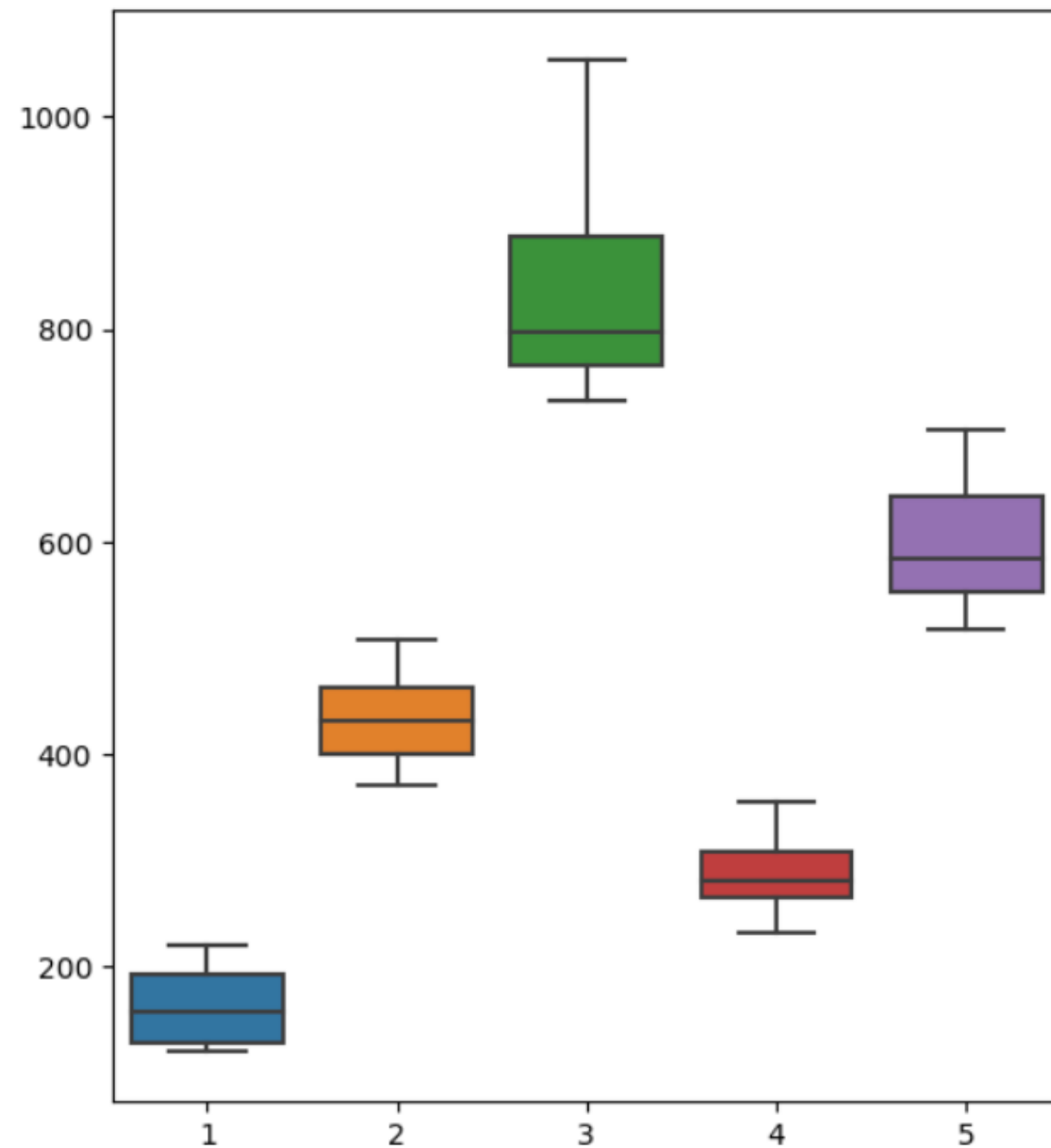


\*not significant

# Additional K-means Analysis

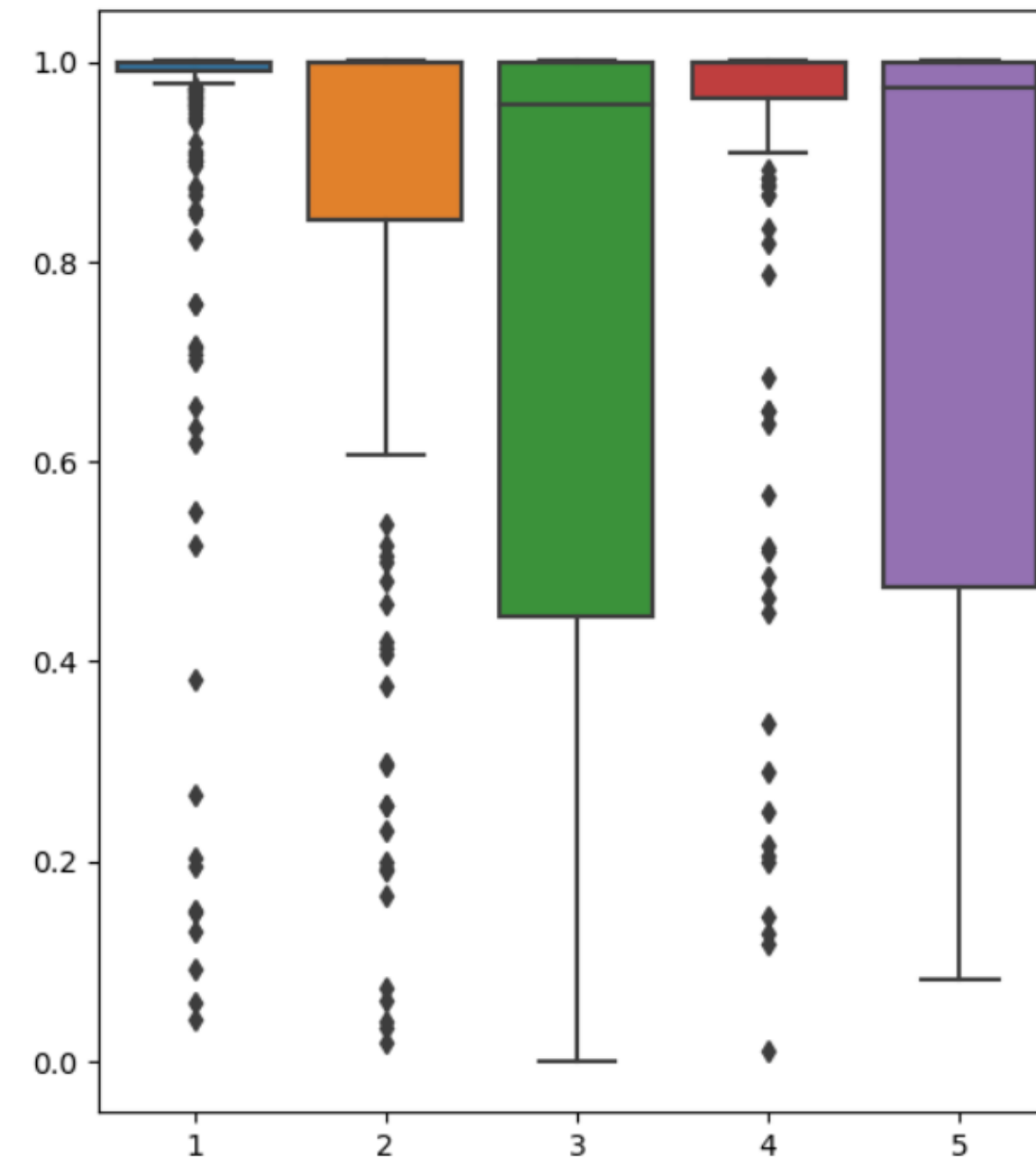
NPL

## Hari Tunggalan



categories made by k-means

## Percent Used



categories made by k-means

\*loaners who used more borrowed money tend to have less arrears duration

\*significant

# Variable Summary

ID_DEB	total_flags
DEBTOR_CATEGORY	slopeCR
PLAFOND_IDR	interceptCR
OS_IDR	slopeDB
HARI_TUNGGAHAN	interceptDB
CUST_TYPE_CD	slopeFCR
COLLECT_KEY	interceptFCR
FLAG_RESTRU	lopeFDB
FLAG_DEFERRED	interceptFDB
REDFLAG_YELLOW	12to3CR
REDFLAG_RED	12to3DB
REDFLAG_INFORMASI	12to3FCR
FLAG_BLACKLIST	12to3FDB
percent_used	

# Logistic Regression

linear\_model.LogisticRegression(solver='saga', max\_iter=300)

Confusion Matrix (all loaners)		Predicted Key			
		1	2	4	5
Actual Key	1	7189	0	0	0
	2	129	0	0	0
	4	0	0	2	16
	5	0	0	2	133

PL: 98.23% accuracy  
NPL: 86.92% accuracy