

# Anti-money laundering on Bitcoin

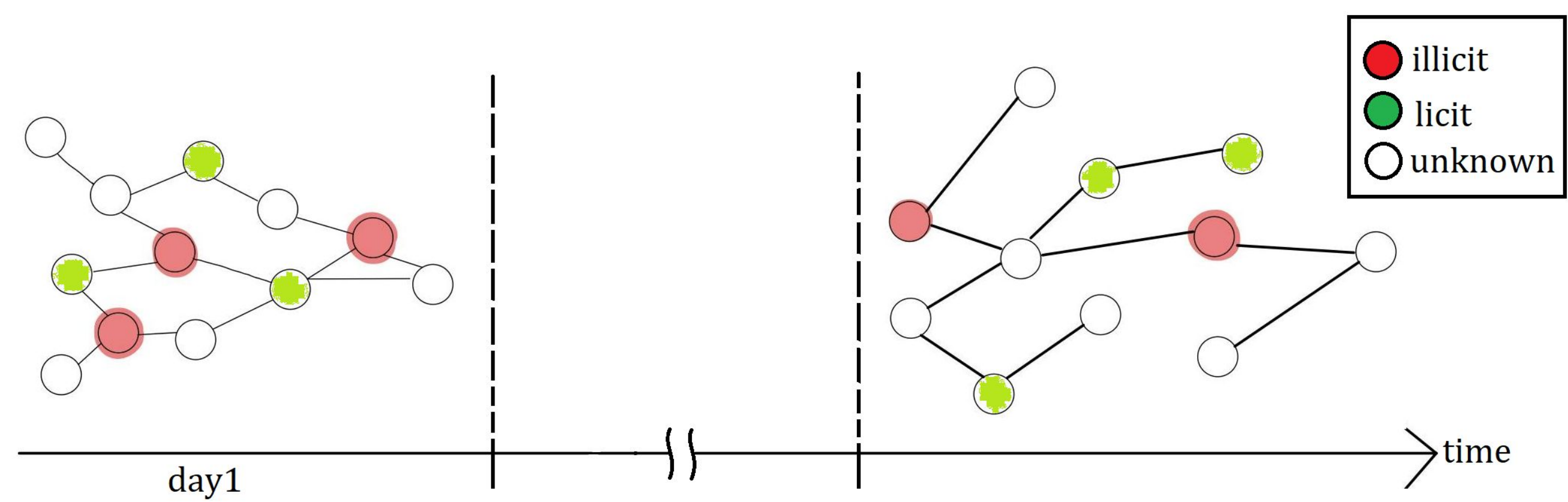
## Data Analysis & Experiment on Various Models

CHUN-YEN LEE 李俊諺; ZHAO-QIAN YUAN 袁肇謙; CHENG-FU WENG 翁丞甫  
109-1 Network & Multimedia Lab Final Project

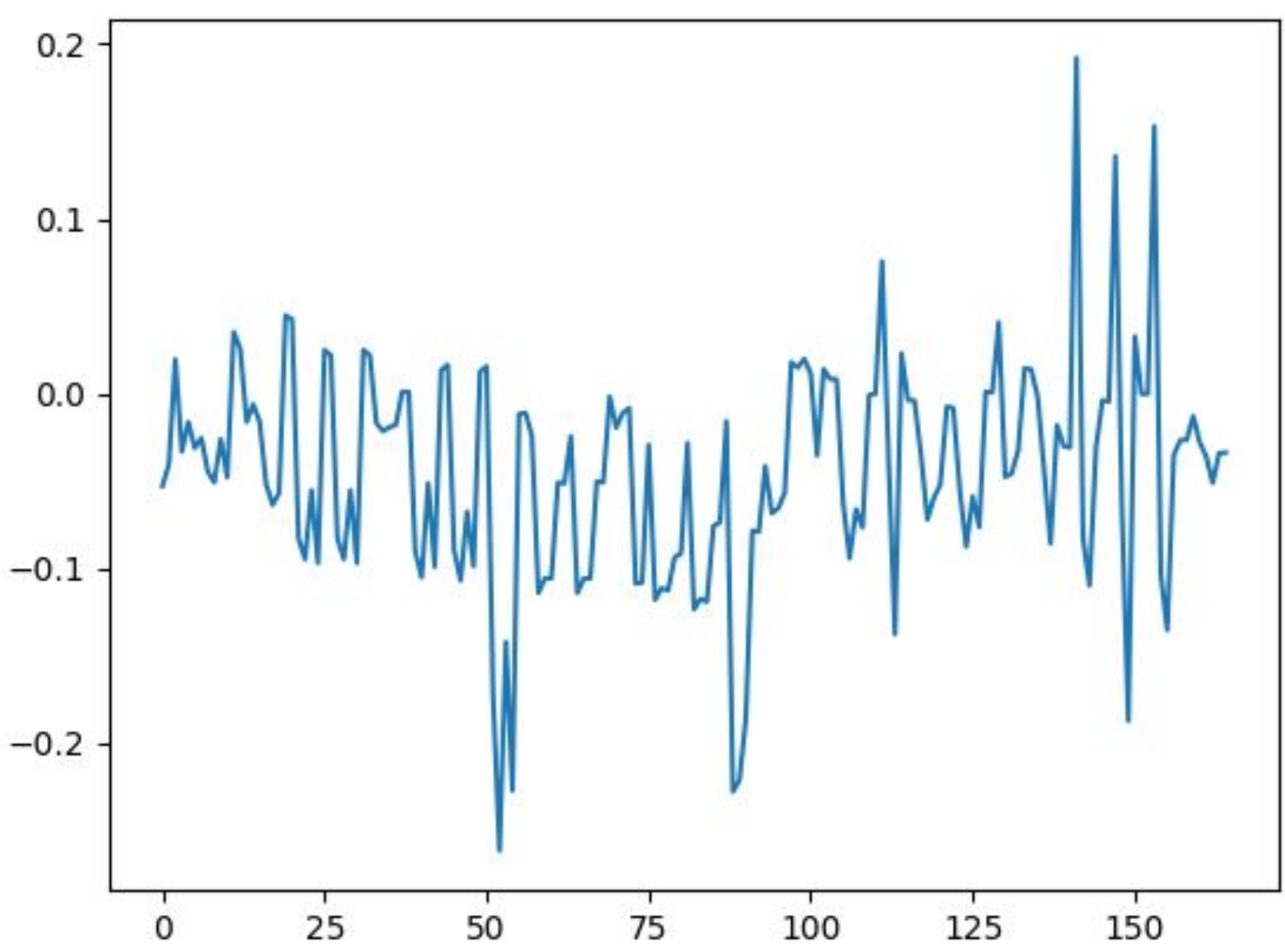
### Introduction

Anti-money laundering (AML) regulations play a critical role in safeguarding financial systems, but bear high costs for institutions and drive financial exclusion for those on the socioeconomic and international margins. We contribute the analysis on Elliptic Data Set, to see if any feature highly correlated to illicit transaction.

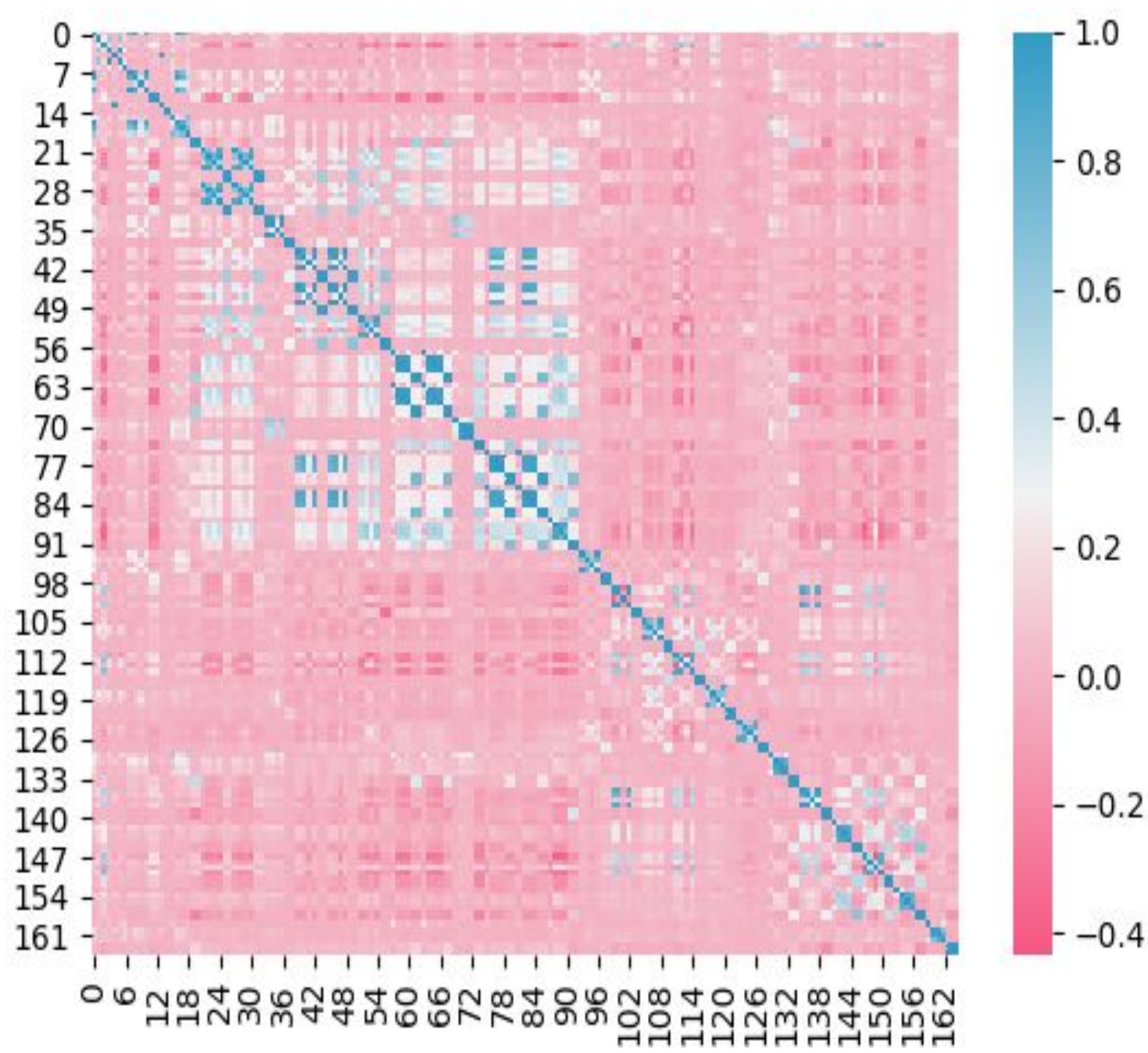
### Data Structure



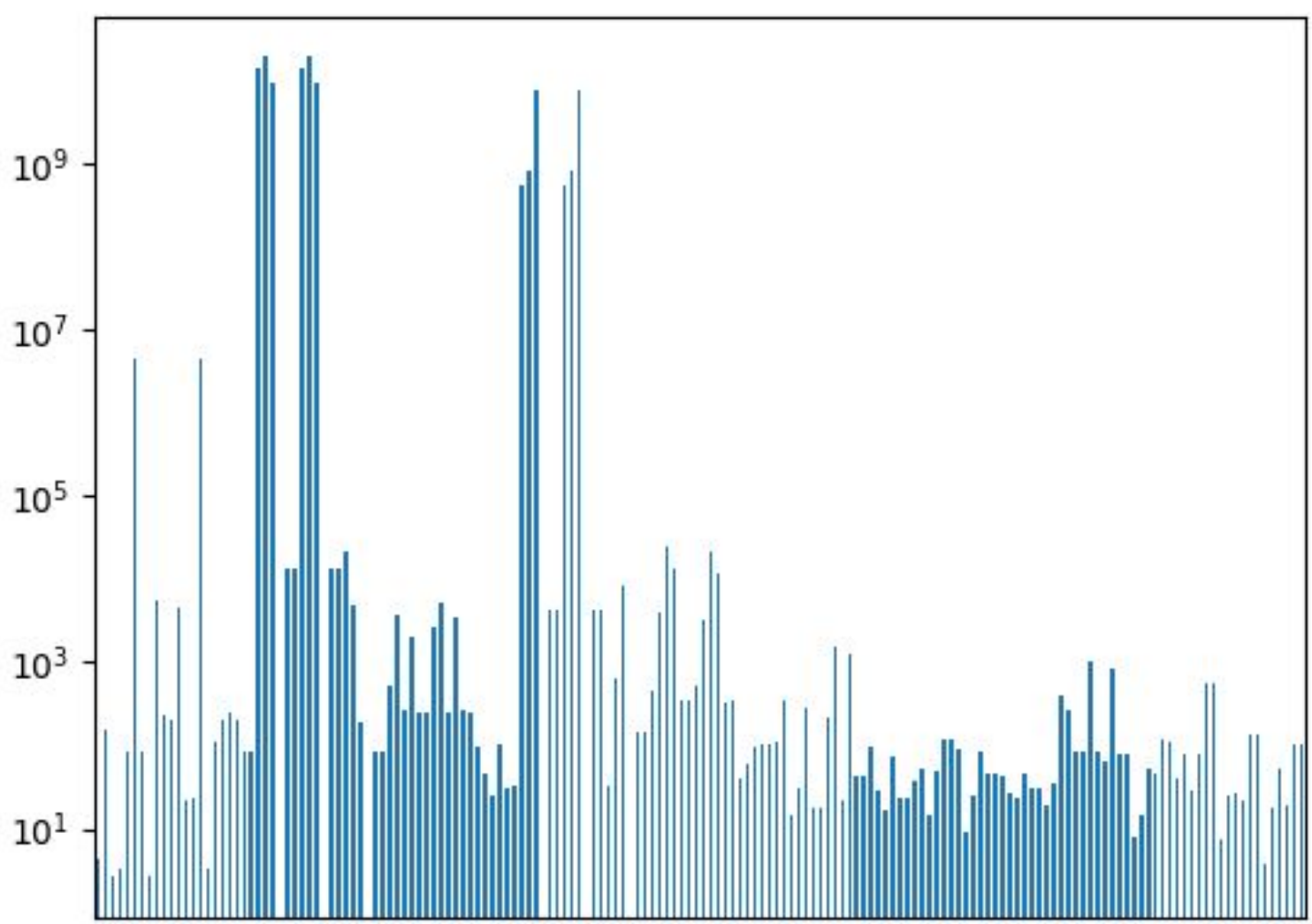
### Feature Correlation with Output



### Highly Correlated Features- correlation matrix



### Highly Correlated Features- VIF



### Experiment Replication

Method	Paper	Self-produced
	$F_1$	$F_1$
Logistic Regression	0.481	0.529
RandomForest	0.788	0.784
MLP	0.653	0.665
degree_2 Logistic Regression		0.540
degree_2 Polynomial Regression		0.447
Linear Regression		0.257
Logistic RegressionCV		0.676
SVM		0.682

### PCA

Method	$F_1$
Logistic Regression	0.477
RandomForest	0.637
MLP	0.689
degree_2 Logistic Regression	0.590
degree_2 Polynomial Regression	0.489
Linear Regression	0.265
Logistic RegressionCV	0.673
SVM	0.714

### Feature selection

Method	PCA	correlation
	$F_1$	$F_1$
Logistic Regression	0.502	0.492
RandomForest	0.613	0.779
MLP	0.667	0.283
degree_2 Logistic Regression	0.616	nan
degree_2 Polynomial Regression	0.533	0.640
Linear Regression	0.267	0.156
Logistic RegressionCV	0.660	0.496
SVM	0.714	0.672