Exploring Prosper Loan (P2P) by Min Lu

Introduction:

Prosper is a P2P marketplace providing person-to-person lending utilizing alisting and bidding process to get competitive rates for loans.

In this project, I will use R and data analyst skills to explore relationships between multiple variables.

Explore Dataset

```
## 'data.frame':
                113937 obs. of 81 variables:
                                    : Factor w/ 113066 levels
## $ ListingKey
"00003546482094282EF90E5",..: 7180 7193 6647 6669 6686 6689 6699 6706 6687
                                    : int 193129 1209647 81716 658116
## $ ListingNumber
909464 1074836 750899 768193 1023355 1023355 ...
## $ ListingCreationDate : Factor w/ 113064 levels "2005-11-
09 20:44:28.847000000",..: 14184 111894 6429 64760 85967 100310 72556 74019
97834 97834 ...
## $ CreditGrade
                                     : Factor w/ 9 levels
"","A","AA","B",..: 5 1 8 1 1 1 1 1 1 1 ...
## $ Term
                                     : int 36 36 36 36 36 60 36 36 36
36 ...
## $ LoanStatus
                               : Factor w/ 12 levels
"Cancelled", "Chargedoff", ..: 3 4 3 4 4 4 4 4 4 4 ...
## $ ClosedDate
                                    : Factor w/ 2803 levels "", "2005-11-
25 00:00:00",..: 1138 1 1263 1 1 1 1 1 1 1 ...
## $ BorrowerAPR
                                    : num 0.165 0.12 0.283 0.125
0.246 ...
                                    : num 0.158 0.092 0.275 0.0974
## $ BorrowerRate
0.2085 ...
                                    : num 0.138 0.082 0.24 0.0874
## $ LenderYield
0.1985 ...
## $ EstimatedEffectiveYield : num NA 0.0796 NA 0.0849
0.1832 ...
## $ EstimatedLoss
                             : num NA 0.0249 NA 0.0249
0.0925 ...
                                   : num NA 0.0547 NA 0.06 0.0907 ...
## $ EstimatedReturn
"","A","AA","B",..: 1 2 1 2 6 4 7 5 3 3 ...
## $ ProsperScore
                                     : num NA 7 NA 9 4 10 2 4 9 11 ...
## $ ListingCategory..numeric. : int 0 2 0 16 2 1 1 2 7 7 ...
## $ BorrowerState
                                     : Factor w/ 52 levels
"","AK","AL","AR",..: 7 7 12 12 25 34 18 6 16 16 ...
## $ Occupation
                                    : Factor w/ 68 levels
"", "Accountant/CPA", ...: 37 43 37 52 21 43 50 29 24 24 ...
```

```
## $ EmploymentStatus : Factor w/ 9 levels
"","Employed",..: 9 2 4 2 2 2 2 2 2 2 ...
\#\# $ EmploymentStatusDuration : int 2 44 NA 113 44 82 172 103 269
269 ...
                               : Factor w/ 2 levels "False", "True":
## $ IsBorrowerHomeowner
2 1 1 2 2 2 1 1 2 2 ...
                                 : Factor w/ 2 levels "False", "True":
## $ CurrentlyInGroup
2 1 2 1 1 1 1 1 1 1 ...
## $ GroupKey
                                          : Factor w/ 707 levels
"","00343376901312423168731",..: 1 1 335 1 1 1 1 1 1 1 ...
\#\# $ DateCreditPulled : Factor w/ 112992 levels "2005-11-
09 00:30:04.487000000",...: 14347 111883 6446 64724 85857 100382 72500 73937
97888 97888 ...
## $ CreditScoreRangeLower
                                          : int 640 680 480 800 680 740 680
700 820 820 ...
## $ CreditScoreRangeUpper : int 659 699 499 819 699 759 699
719 839 839 ...
## $ FirstRecordedCreditLine : Factor w/ 11586 levels "","1947-
08-24 00:00:00",..: 8639 6617 8927 2247 9498 497 8265 7685 5543 5543 ...
## $ CurrentCreditLines
                              : int 5 14 NA 5 19 21 10 6 17
17 ...
## $ OpenCreditLines : int 4 14 NA 5 19 17 7 6 16 16 ... 
## $ TotalCreditLinespast7years : int 12 29 3 29 49 49 20 10 32
## $ OpenRevolvingAccounts : int 1 13 0 7 6 13 6 5 12 12 ...
## $ OpenRevolvingMonthlyPayment : num 24 389 0 115 220 1410 214 101
219 219 ...
## $ InquiriesLast6Months : int 3 3 0 0 1 0 0 3 1 1 ...
## $ TotalInquiries
                                          : num 3 5 1 1 9 2 0 16 6 6 ...
## $ CurrentDelinquencies : int 2 0 1 4 0 0 0 0 0 0 ...
## $ AmountDelinquent : num 472 0 NA 10056 0 ...
## $ AmountDelinquent : num 472 0 NA 10056 0 ...

## $ DelinquenciesLast7Years : int 4 0 0 14 0 0 0 0 0 0 ...

## $ PublicRecordsLast10Years : int 0 1 0 0 0 0 0 1 0 0 ...

## $ PublicRecordsLast12Months : int 0 0 NA 0 0 0 0 0 0 0 ...

## $ RevolvingCreditBalance : num 0 3989 NA 1444 6193 ...

## $ BankcardUtilization : num 0 0.21 NA 0.04 0.81 0.39 0.72
0.13 0.11 0.11 ...
## $ AvailableBankcardCredit : num 1500 10266 NA 30754 695 ...
                                          : num 11 29 NA 26 39 47 16 10 29
## $ TotalTrades
29 ...
## $ TradesNeverDelinquent..percentage. : num 0.81 1 NA 0.76 0.95 1 0.68
0.8 1 1 ...
## $ TradesOpenedLast6Months : num 0 2 NA 0 2 0 0 0 1 1 ...
## $ DebtToIncomeRatio
                                          : num 0.17 0.18 0.06 0.15 0.26 0.36
0.27 0.24 0.25 0.25 ...
                                          : Factor w/ 8 levels "$0","$1-
## $ IncomeRange
24,999",..: 4 5 7 4 3 3 4 4 4 4 ...
                                          : Factor w/ 2 levels "False", "True":
## $ IncomeVerifiable
2 2 2 2 2 2 2 2 2 2 ...
## $ StatedMonthlyIncome : num 3083 6125 2083 2875 9583 ...
## $ LoanKey
                                           : Factor w/ 113066 levels
"00003683605746079487FF7",..: 100337 69837 46303 70776 71387 86505 91250 5425
908 908 ...
## $ TotalProsperLoans
                                          : int NA NA NA NA 1 NA NA NA NA
NA ...
## $ TotalProsperPaymentsBilled : int NA NA NA NA 11 NA NA NA NA
NA ...
```

```
## $ OnTimeProsperPayments : int NA NA NA NA 11 NA NA NA NA
NA ...
## $ ProsperPaymentsLessThanOneMonthLate: int NA NA NA NA O NA NA NA NA
## $ ProsperPaymentsOneMonthPlusLate : int NA NA NA NA O NA NA NA NA
## $ ProsperPrincipalBorrowed
                                   : num NA NA NA NA 11000 NA NA NA NA
NA ...
## $ ProsperPrincipalOutstanding
                                   : num NA NA NA NA 9948 ...
## $ ScorexChangeAtTimeOfListing
                                    : int NA NA NA NA NA NA NA NA
NA ...
: int NA NA NA NA NA NA NA NA
## $ LoanMonthsSinceOrigination : int 78 0 86 16 6 3 11 10 3 3 ...
                                    : int 19141 134815 6466 77296
## $ LoanNumber
102670 123257 88353 90051 121268 121268 ...
## $ LoanOriginalAmount
                                   : int 9425 10000 3001 10000 15000
15000 3000 10000 10000 10000 ...
## $ LoanOriginationDate
                                    : Factor w/ 1873 levels "2005-11-15
00:00:00",...: 426 1866 260 1535 1757 1821 1649 1666 1813 1813 ...
## $ LoanOriginationQuarter : Factor w/ 33 levels "Q1 2006","Q1
2007",..: 18 8 2 32 24 33 16 16 33 33 ...
                                    : Factor w/ 90831 levels
## $ MemberKev
"00003397697413387CAF966",..: 11071 10302 33781 54939 19465 48037 60448 40951
26129 26129 ...
## $ MonthlyLoanPayment
                                    : num 330 319 123 321 564 ...
## $ LP_CustomerPayments
                                   : num 11396 0 4187 5143 2820 ...
## $ LP CustomerPrincipalPayments : num 9425 0 3001 4091 1563 ...
## $ LP_InterestandFees
                                    : num 1971 0 1186 1052 1257 ...
## $ LP_ServiceFees
                                   : num -133.2 0 -24.2 -108 -60.3 ...
## $ LP_CollectionFees
                                   : num 0 0 0 0 0 0 0 0 0 ...
## $ LP GrossPrincipalLoss
                                   : num 0 0 0 0 0 0 0 0 0 ...
## $ LP NetPrincipalLoss
                                   : num 0 0 0 0 0 0 0 0 0 ...
## $ LP NonPrincipalRecoverypayments : num 0 0 0 0 0 0 0 0 0 ...
## $ PercentFunded
                                    : num 1 1 1 1 1 1 1 1 1 1 ...
## $ Recommendations
                                    : int 0000000000...
                                   : int 0000000000...
##
   $ InvestmentFromFriendsCount
## $ InvestmentFromFriendsAmount
                                   : num 0 0 0 0 0 0 0 0 0 ...
## $ Investors
                                    : int 258 1 41 158 20 1 1 1 1 1 ...
## [1] "ListingKev"
## [2] "ListingNumber"
## [3] "ListingCreationDate"
   [4] "CreditGrade"
##
##
   [5] "Term"
## [6] "LoanStatus"
## [7] "ClosedDate"
## [8] "BorrowerAPR"
## [9] "BorrowerRate"
## [10] "LenderYield"
## [11] "EstimatedEffectiveYield"
## [12] "EstimatedLoss"
## [13] "EstimatedReturn"
## [14] "ProsperRating..numeric."
## [15] "ProsperRating..Alpha."
## [16] "ProsperScore"
## [17] "ListingCategory..numeric."
```

```
## [18] "BorrowerState"
## [19] "Occupation"
## [20] "EmploymentStatus"
## [21] "EmploymentStatusDuration"
## [22] "IsBorrowerHomeowner"
## [23] "CurrentlyInGroup"
## [24] "GroupKey"
## [25] "DateCreditPulled"
## [26] "CreditScoreRangeLower"
## [27] "CreditScoreRangeUpper"
## [28] "FirstRecordedCreditLine"
## [29] "CurrentCreditLines"
## [30] "OpenCreditLines"
## [31] "TotalCreditLinespast7years"
## [32] "OpenRevolvingAccounts"
## [33] "OpenRevolvingMonthlyPayment"
## [34] "InquiriesLast6Months"
## [35] "TotalInquiries"
## [36] "CurrentDelinquencies"
## [37] "AmountDelinquent"
## [38] "DelinquenciesLast7Years"
## [39] "PublicRecordsLast10Years"
## [40] "PublicRecordsLast12Months"
## [41] "RevolvingCreditBalance"
## [42] "BankcardUtilization"
## [43] "AvailableBankcardCredit"
## [44] "TotalTrades"
## [45] "TradesNeverDelinquent..percentage."
## [46] "TradesOpenedLast6Months"
## [47] "DebtToIncomeRatio"
## [48] "IncomeRange"
## [49] "IncomeVerifiable"
## [50] "StatedMonthlyIncome"
## [51] "LoanKey"
## [52] "TotalProsperLoans"
## [53] "TotalProsperPaymentsBilled"
## [54] "OnTimeProsperPayments"
## [55] "ProsperPaymentsLessThanOneMonthLate"
## [56] "ProsperPaymentsOneMonthPlusLate"
## [57] "ProsperPrincipalBorrowed"
## [58] "ProsperPrincipalOutstanding"
## [59] "ScorexChangeAtTimeOfListing"
## [60] "LoanCurrentDaysDelinquent"
## [61] "LoanFirstDefaultedCycleNumber"
## [62] "LoanMonthsSinceOrigination"
## [63] "LoanNumber"
## [64] "LoanOriginalAmount"
## [65] "LoanOriginationDate"
## [66] "LoanOriginationQuarter"
## [67] "MemberKey"
## [68] "MonthlyLoanPayment"
## [69] "LP CustomerPayments"
## [70] "LP CustomerPrincipalPayments"
## [71] "LP InterestandFees"
## [72] "LP ServiceFees"
## [73] "LP CollectionFees"
```

[74] "LP GrossPrincipalLoss"

```
## [75] "LP NetPrincipalLoss"
## [76] "LP NonPrincipalRecoverypayments"
## [77] "PercentFunded"
## [78] "Recommendations"
## [79] "InvestmentFromFriendsCount"
## [80] "InvestmentFromFriendsAmount"
## [81] "Investors"
## [1] 113937
## [1] 81
##
                  ListingKey
                                ListingNumber
##
  17A93590655669644DB4C06: 6
                                Min. : 4
## 349D3587495831350F0F648:
                            4 1st Qu.: 400919
                           4 Median: 600554
## 47C1359638497431975670B:
## 8474358854651984137201C:
                            4 Mean : 627886
## DE8535960513435199406CE:
                            4 3rd Qu.: 892634
## 04C13599434217079754AEE: 3 Max. :1255725
##
   (Other)
                        :113912
##
                    ListingCreationDate CreditGrade
                                                         Term
## 2013-10-02 17:20:16.550000000: 6 :84984 Min. :12.00
## 2013-08-28 20:31:41.107000000:
                                 4 C
                                           : 5649 1st Qu.:36.00
## 2013-09-08 09:27:44.853000000:
                                 4 D
                                           : 5153
                                                    Median :36.00
## 2013-12-06 05:43:13.830000000:
                                 4 B
                                            : 4389
                                                    Mean :40.83
  __ 00 11:44:58.283000000:
2013-08-21 07:25:22.360000000:
(Other)
                                            : 3509
## 2013-12-06 11:44:58.283000000:
                                  4 AA
                                                     3rd Ou.:36.00
##
                                  3
                                     HR
                                            : 3508
                                                    Max. :60.00
##
                          :113912 (Other): 6745
  (Other)
##
                 LoanStatus
                                           ClosedDate
## Current
                    :56576
                                               :58848
                     :38074 2014-03-04 00:00:00: 105
## Completed
                     :11992 2014-02-19 00:00:00: 100
## Chargedoff
                              2014-02-11 00:00:00:
## Defaulted
                      : 5018
   Past Due (1-15 days) : 806
##
                              2012-10-30 00:00:00:
## Past Due (31-60 days): 363
                              2013-02-26 00:00:00: 78
                                               :54633
## (Other)
                     : 1108 (Other)
##
   BorrowerAPR
                   BorrowerRate
                                  LenderYield
## Min. :0.00653 Min. :0.0000 Min. :-0.0100
  1st Qu.:0.15629 1st Qu.:0.1340 1st Qu.: 0.1242
##
## Median: 0.20976 Median: 0.1840 Median: 0.1730
                  Mean :0.1928 Mean : 0.1827
## Mean :0.21883
## 3rd Qu.:0.28381 3rd Qu.:0.2500 3rd Qu.: 0.2400
## Max. :0.51229 Max. :0.4975 Max. :0.4925
## NA's :25
## EstimatedEffectiveYield EstimatedLoss EstimatedReturn
## Min. :-0.183 Min. :0.005 Min. :-0.183
                        1st Qu.:0.042 1st Qu.: 0.074
##
  1st Qu.: 0.116
## Median : 0.162
                        Median : 0.072 Median : 0.092
                       Mean :0.080 Mean : 0.096
## Mean : 0.169
## 3rd Qu.: 0.224
                        3rd Qu.:0.112 3rd Qu.: 0.117
## Max. : 0.320
                       Max. :0.366 Max. : 0.284
## NA's :29084
                       NA's :29084 NA's :29084
  ProsperRating..numeric. ProsperRating..Alpha. ProsperScore
##
##
   Min. :1.000
                            :29084
                                            Min. : 1.00
                              :18345
   1st Qu.:3.000
##
                        С
                                            1st Qu.: 4.00
                       В
## Median :4.000
                              :15581
                                           Median : 6.00
## Mean :4.072
                       А
                              :14551
                                           Mean : 5.95
## 3rd Ou.:5.000
                       D
                              :14274
                                            3rd Qu.: 8.00
## Max. :7.000 E : 9795 Max.
## NA's :29084 (Other):12307 NA's
                                           Max. :11.00
                                                 :29084
```

```
## ListingCategory..numeric. BorrowerState
                    CA :14717
TX : 6842
NY : 6729
## Min. : 0.000
## 1st Qu.: 1.000
                            NY
                                   : 6729
## Median : 1.000
                         FL
IL
## Mean : 2.774
                                   : 6720
## 3rd Qu.: 3.000
                                   : 5921
## Max. :20.000
                                    : 5515
                          (Other):67493
##
##
                      Occupation EmploymentStatus
                         :28617
## Other
                                   Employed :67322
                        :13628 Full-time :26355
: 4478 Self-employed: 6134
## Professional
## Computer Programmer
## Executive
## Teacher
                          : 4311 Not available: 5347
                           : 3759 Other
                                                 : 3806
                                                 : 2255
## Administrative Assistant: 3688
## (Other) :55456 (Other) : 2718
## EmploymentStatusDuration IsBorrowerHomeowner CurrentlyInGroup
## Min. : 0.00 False:56459 False:101218
## 1st Qu.: 26.00
                           True :57478
                                               True : 12719
## Median : 67.00
## Mean : 96.07
## 3rd Qu.:137.00
## Max. :755.00
## NA's :7625
                      GroupKey
##
                                               DateCreditPulled
##
                          :100596 2013-12-23 09:38:12: 6
## 783C3371218786870A73D20: 1140 2013-11-21 09:09:41:
## 3D4D3366260257624AB272D: 916 2013-12-06 05:43:16:
## 6A3B336601725506917317E: 698 2014-01-14 20:17:49:
## FEF83377364176536637E50: 611 2014-02-09 12:14:41:
## C9643379247860156A00EC0: 342 2013-09-27 22:04:54:
## (Other) : 9634 (Other) :1
                                               :113912
## CreditScoreRangeLower CreditScoreRangeUpper
## Min. : 0.0 Min. : 19.0
## 1st Qu.:660.0
                        1st Qu.:679.0
## Median :000.5
## Mean :685.6
                        Median :699.0
                       Mean :704.6
3rd Qu.:739.0
                        Max. :899.0
## Max. :880.0
## NA's :591 NA's :591
##
          FirstRecordedCreditLine CurrentCreditLines OpenCreditLines
##
                   : 697 Min. : 0.00 Min. : 0.00
## 1993-12-01 00:00:00: 185
                                 1st Qu.: 7.00
                                                    1st Qu.: 6.00
## 1994-11-01 00:00:00: 178
## 1995-11-01 00:00:00: 168
                               Median :10.00
                                                   Median: 9.00
## 1995-11-01 00:00:00: 168 Mean :10.32

## 1990-04-01 00:00:00: 161 3rd Qu.:13.00

## 1995-03-01 00:00:00: 159 Max. :59.00

## (Other) :112389 NA's :7604
                                                    Mean : 9.26
                                                    3rd Qu.:12.00
                                                   Max. :54.00
                                                    NA's :7604
## TotalCreditLinespast7years OpenRevolvingAccounts
## Min. : 2.00 Min. : 0.00
   1st Qu.: 17.00
##
                              1st Qu.: 4.00
                             Median : 6.00
## Median : 25.00
## Mean : 26.75
                             Mean : 6.97
## 3rd Qu.: 35.00
                             3rd Qu.: 9.00
## Max. :136.00
                             Max. :51.00
## NA's :697
## OpenRevolvingMonthlyPayment InquiriesLast6Months TotalInquiries
```

```
## Min. : 0.0 Min. : 0.000 Min. : 0.000
## 1st Qu.: 114.0 1st Qu.: 0.000 1st Qu.: 2.000
## Median : 271.0 Median : 1.000 Median : 4.000
## Mean : 398.3 Mean : 1.435 Mean : 5.584
## 3rd Qu.: 525.0 3rd Qu.: 2.000 3rd Qu.: 7.000
## Max. :14985.0 Max. :105.000 Max. :379.000
## " NA's :697 NA's :1159
## CurrentDelinquencies AmountDelinquent DelinquenciesLast7Years
## CurrentDelinquencies AmountDelinquent DelinquenciesLa
## Min. : 0.0000 Min. : 0.0 Min. : 0.000
## 1st Qu.: 0.0000 1st Qu.: 0.0 1st Qu.: 0.000
## Median : 0.0000 Median : 0.0 Median : 0.000
## Mean : 0.5921 Mean : 984.5 Mean : 4.155
## 3rd Qu.: 0.0000 3rd Qu.: 0.0 3rd Qu.: 3.000
## Max. :83.0000 Max. :463881.0 Max. :99.000
## NA's :697 NA's :7622 NA's :990
## PublicRecordsLast10Years PublicRecordsLast12Months RevolvingCreditBalance
## Min. : 0.0000 Min. : 0.000 Min. : 0
## 1st Qu.: 0.0000 1st Qu.: 3121
## Median : 0.0000 Median : 0.000 Median : 8549
## Mean : 0.3126 Mean : 0.015 Mean : 17599
## 3rd Qu.: 0.0000 3rd Qu.: 0.000 3rd Qu.: 19521
## Max. :38.0000 Max. :20.000 Max. :1435667
## NA's :697 NA's :7604 NA's :7604
## BankcardUtilization AvailableBankcardCredit TotalTrades
## Min. :0.000 Min. : 0 Min. : 0.00
## TradesNeverDelinquent..percentage. TradesOpenedLast6Months
## Min. :0.000
                                                                        Min. : 0.000
## 1st Qu.:0.820
                                                                         1st Qu.: 0.000
## Median: 0.940
## Mean: 0.886
## 3rd Qu: 1.000
## Max: 1.000
## Median :0.940
                                                                        Median : 0.000
                                                                        Mean : 0.802
                                                                          3rd Qu.: 1.000
                                                                         Max. :20.000
## NA's :7544
                                                                         NA's :7544
## DebtToIncomeRatio IncomeRange IncomeVerifiable
## Min. : 0.000 $25,000-49,999:32192 False: 8669
## 1st Qu.: 0.140 $50,000-74,999:31050 True :105268
## Median : 0.220 $100,000+ :17337
## Mean : 0.276 $75,000-99,999:16916
## 3rd Qu.: 0.320 Not displayed : 7741
## Max. :10.010 $1-24,999 : 7274
## NA's :8554 (Other) : 1427
                                                                                  LoanKey TotalProsperLoans
## StatedMonthlyIncome
## Min. : 0 CB1B37030986463208432A1: 6 Min. :0.00
## 1st Qu.: 3200 2DEE3698211017519D7333F: 4 1st Qu.:1.00
## Median : 4667 9F4B37043517554537C364C: 4 Median :1.00
## Mean : 5608 D895370150591392337ED6D: 4 Mean :1.42
## 3rd Qu.: 6825 E6FB37073953690388BC56D: 4 3rd Qu.:2.00
## Max. :1750003 OD8F37036734373301ED419: 3 Max. :8.00
## (Other) :113912 NA's :91852
                                                                  :113912 NA's :91852
## TotalProsperPaymentsBilled OnTimeProsperPayments
## Min. : 0.00 Min. : 0.00
```

```
## Max. :141.00 Max. :141.00
## NA's :91852 NA's :91852
## ProsperPaymentsLessThanOneMonthLate ProsperPaymentsOneMonthPlusLate
## Min. : 0.00 Min. : 0.00
## 1st Qu.: 0.00
## Median: 0.00
## Mean: 0.61
## 3rd Qu.: 0.00
                                            1st Qu.: 0.00
                                            Median: 0.00
                                            Mean : 0.05
                                            3rd Qu.: 0.00
## Max. :42.00
                                           Max. :21.00
## NA's :91852
                                           NA's :91852
## ProsperPrincipalBorrowed ProsperPrincipalOutstanding
## Min. : 0 Min. : 0
## 1st Qu.: 3500 1st Qu.: 0
## Median : 6000 Median : 1627
## Mean : 8472 Mean : 2930
                              Mean : 2930
3rd Qu.: 4127
## Mean : 8472
## 3rd Qu.:11000
## Max. :72499 Max. :23451
## NA's :91852 NA's :91852
## ScorexChangeAtTimeOfListing LoanCurrentDaysDelinquent
## Min. :-209.00 Min. : 0.0
                            1st Qu.: 0.0

Median: 0.0

Mean: 152.8

3rd Qu.: 0.0

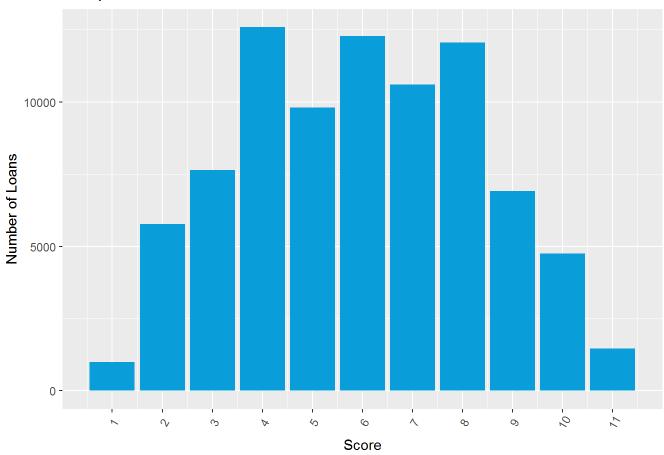
Max. :2704.0
## 1st Qu.: -35.00
## Median : -3.00
## Mean : -3.22
## 3rd Qu.: 25.00
## Max. : 286.00
## NA's :95009
## LoanFirstDefaultedCycleNumber LoanMonthsSinceOrigination LoanNumber
## Min. : 0.00 Min. : 1
## 1st Qu.: 9.00
                                     1st Qu.: 6.0
                                                                  1st Qu.: 37332
## Median :14.00
                                    Median : 21.0
                                                                  Median : 68599
                                  Mean : 31.9
3rd Qu.: 65.0
Max. :100.0
                                                             Mean : 69444
3rd Qu.:101901
Max. :136486
## Mean :16.27
                                     3rd Qu.: 65.0
## 3rd Qu.:22.00
## Max. :44.00
## NA's
           :96985
## LoanOriginalAmount LoanOriginationDate LoanOriginationQuarter
## (Other) :111428 (Other):60343
## MemberKey MonthlyLoanPayment LP_CustomerPayment
## 63CA34120866140639431C9: 9 Min.: 0.0 Min.: -2.35
                                        MonthlyLoanPayment LP CustomerPayments
## 16083364744933457E57FB9: 8 1st Qu.: 131.6 1st Qu.: 1005.76
## 3A2F3380477699707C81385: 8 Median : 217.7 Median : 2583.83
## 4D9C3403302047712AD0CDD: 8 Mean : 272.5 Mean : 4183.08
## 739C338135235294782AE75: 8 3rd Qu.: 371.6 3rd Qu.: 5548.40
## 7E1733653050264822FAA3D: 8 Max. :2251.5 Max. :40702.39
## (Other) :113888
## LP CustomerPrincipalPayments LP InterestandFees LP ServiceFees
## Min. : 0.0 Min. : -2.35 Min. :-664.87
## 1st Qu.: 500.9 1st Qu.: 274.87 1st Qu.: -73.18
```

```
Median : 700.84 Median : -34.44
## Median : 1587.5
## Mean : 3105.5
                             Mean : 1077.54 Mean : -54.73
                             3rd Qu.: 1458.54 3rd Qu.: -13.92
##
   3rd Qu.: 4000.0
                             Max. :15617.03 Max. : 32.06
## Max. :35000.0
##
## LP CollectionFees LP GrossPrincipalLoss LP NetPrincipalLoss
## Min. :-9274.75 Min. : -94.2 Min. : -954.5
## 1st Qu.: 0.00 1st Qu.: 0.0
## Median: 0.00 Median: 0.0
                                        1st Qu.:
                                        Median :
                                                   0.0
## Mean : -14.24 Mean : 700.4 Mean : 681.4
## 3rd Qu.: 0.00 3rd Qu.: 0.0 3rd Qu.: 0.0
## Max. : 0.00 Max. :25000.0 Max. :25000.0
##
## LP NonPrincipalRecoverypayments PercentFunded Recommendations
                                Min. :0.7000 Min. : 0.00000
## Min. : 0.00
                                ## 1st Qu.:
             0.00
## Median :
             0.00
## Mean : 25.14
                               Mean :0.9986 Mean : 0.04803
## 3rd Qu.: 0.00
                                3rd Qu.:1.0000 3rd Qu.: 0.00000
## Max. :21117.90
                               Max. :1.0125 Max. :39.00000
##
## InvestmentFromFriendsCount InvestmentFromFriendsAmount Investors
## Min. : 0.00000 Min. : 0.00 Min. : 1.00
                          1st Qu.: 0.00
Median: 0.00
##
  1st Qu.: 0.00000
                                                      1st Qu.:
                                                     Median : 44.00
## Median : 0.00000
                          Mean : 16.55
3rd Qu.: 0.00
Max. :25000.00
## Mean : 0.02346
                                                     Mean : 80.48
                                                 3rd Qu.: 115.00
Max. :1189.00
## 3rd Qu.: 0.00000
## Max. :33.00000
##
```

Univariate Plots Section

```
## Min. 1st Qu. Median Mean 3rd Qu. Max. NA's
## 1.00 4.00 6.00 5.95 8.00 11.00 29084
## Warning: Removed 29084 rows containing non-finite values (stat count).
```

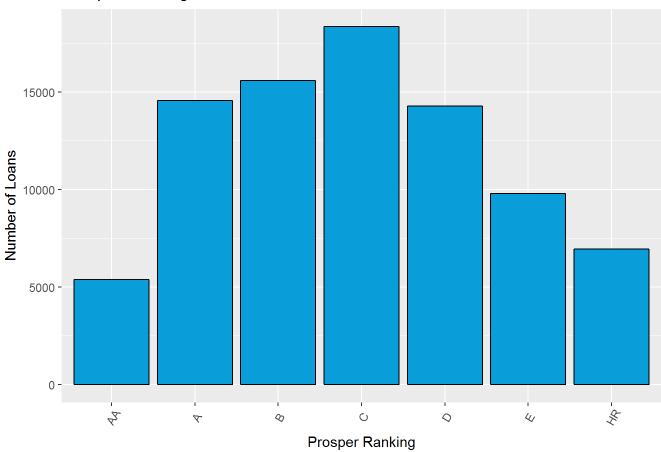
Prosper Score Distribution



Here we can see a normal distribution chart if we did not count NA. It looks like most loans are under 4-8 score ranges.

A AA B C D E HR ## 29084 14551 5372 15581 18345 14274 9795 6935

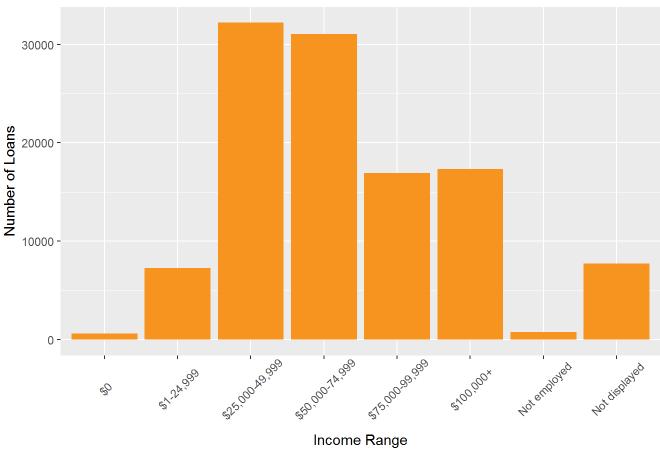
Prosper Ranking Distribution



From the chart above, the majority of loan are being rated as "A", "B", "c" and "D".

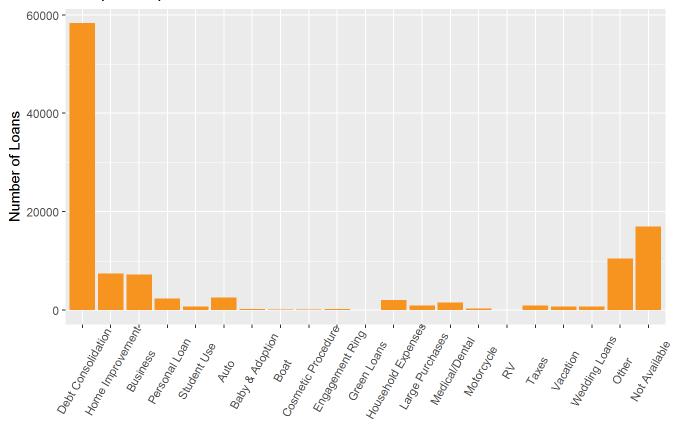
##	\$0	\$1-24 , 999	\$25,000-49,999	\$50,000-74,999	\$75 , 000-99 , 999
##	621	7274	32192	31050	16916
##	\$100,000+	Not employed	Not displayed		
##	17337	806	7741		





People from income range 25,000-49,999 and 50,000-74,999 are the main borrower. However, people with income under 25000 are not major borrower. So, probably Prosper.com has some income limits on borrowers or Prosper.com main target are future home-owners or car-owners, but not for student loan.

Prosper Purpose of loan Distribution



Listing Catagory

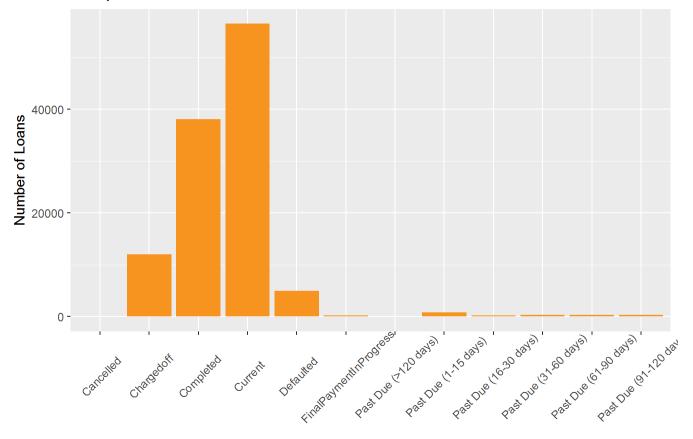
Business	Home Improvement	Debt Consolidation	##
7189	7433	58308	##
Auto	Student Use	Personal Loan	##
2572	756	2395	##
Cosmetic Procedure	Boat	Baby & Adoption	##
91	85	199	##
Household Expenses	Green Loans	Engagement Ring	##
1996	59	ŧ 217	##
Motorcycle	Medical/Dental	Large Purchases	##
304	1522	‡ 876	##
Vacation	Taxes	ŧ RV	##
768	885	‡ 52	##
Not Available	Other	Wedding Loans	##
16965	10494	† 771	##

From this chart, we could see most people want money for debit consolidation, home improvement and business. Debit consolidation has a surprising amount which almost 50% of total number of loans. These data shows that most borrower already have their first job and is looking to pay their debt.

##	Cancelled	Chargedoff	Completed
##	5	11992	38074

```
##
                                         Defaulted FinalPaymentInProgress
                   Current
##
                     56576
                                               5018
##
     Past Due (>120 days)
                              Past Due (1-15 days)
                                                     Past Due (16-30 days)
##
                        16
                                                806
                                                                         265
    Past Due (31-60 days)
##
                             Past Due (61-90 days) Past Due (91-120 days)
##
                       363
                                                313
                                                                         304
```

Prosper Loan Status Distribution

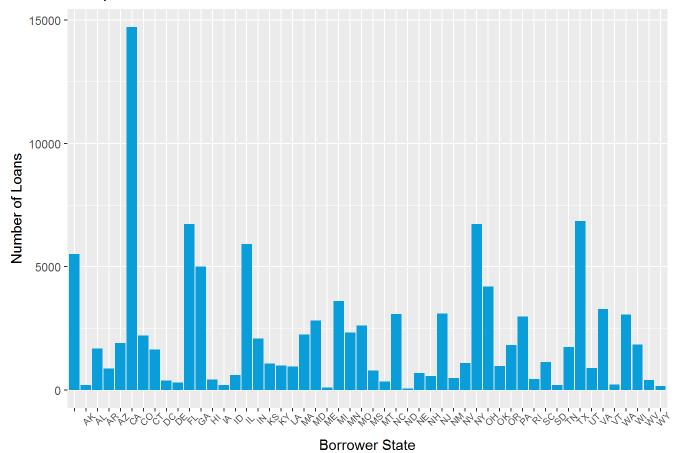


Loan Status

Here we can see that more than 50% of loans are under active status and almost 38% are complete. However, more than 10% loans are defaulted or past due, which probably will reduce the company revenue.

##		AK	AL	AR	ΑZ	CA	CO	CT	DC	DE	FL	GA
##	5515	200	1679	855	1901	14717	2210	1627	382	300	6720	5008
##	ΗI	IA	ID	IL	IN	KS	KY	LA	MA	MD	ME	MI
##	409	186	599	5921	2078	1062	983	954	2242	2821	101	3593
##	MN	MO	MS	MT	NC	ND	NE	NH	NJ	NM	NV	NY
##	2318	2615	787	330	3084	52	674	551	3097	472	1090	6729
##	ОН	OK	OR	PA	RI	SC	SD	TN	TX	UT	VA	VT
##	4197	971	1817	2972	435	1122	189	1737	6842	877	3278	207
##	WA	WI	WV	WY								
##	3048	1842	391	150								

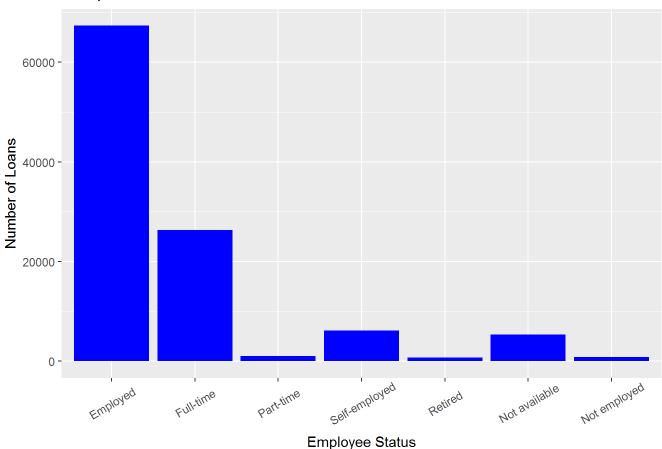
Prosper Borrower State Distribution



From this chart, we could see the top borrower is from "CA", which is almost 15,000 borrowers. Other popular states are "FL", "IL", "NY" and "TX".

##	Employed	Full-time	Part-time	Self-employed	Retired
##	67322	26355	1088	6134	795
##	Not available	Not employed	others		NA's
##	5347	835	0	0	6061

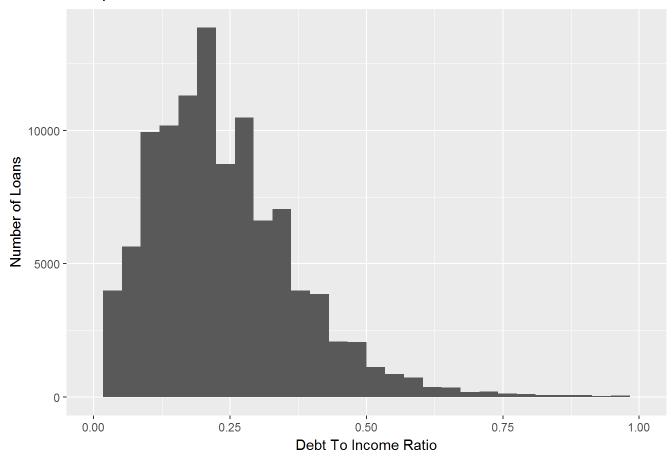
Prosper Borrower State Distribution



Most borrowers are employed or had a full-time job. Surprisingly, there're few part-time job borrowers. Probabaly because part-time job is not stable or do not have enough income to pay back the loan.

```
## Min. 1st Qu. Median Mean 3rd Qu. Max. NA's
## 0.000 0.140 0.220 0.276 0.320 10.010 8554
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
## Warning: Removed 9353 rows containing non-finite values (stat bin).
```

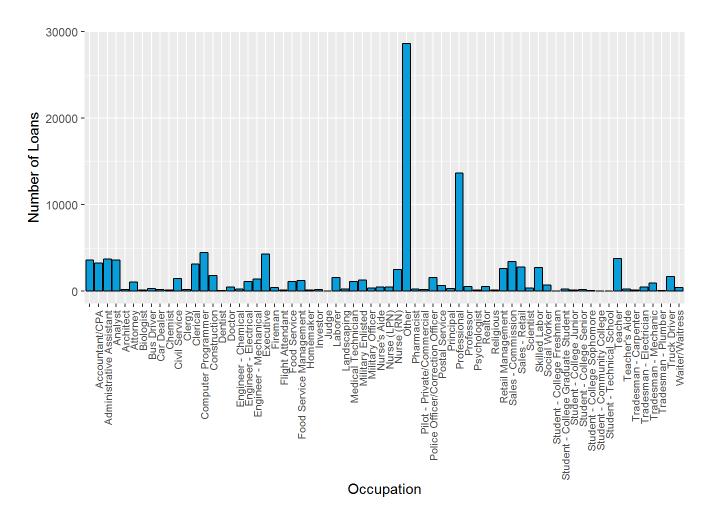
Prosper Debt to Income ratio Distribution



Most borrower's debt to income ratio is between 0-0.6. The peak is around 0.23. However, there're still some borrower's ratio is above 0.75.

##		Accountant/CPA
##	3588	3233
##	Administrative Assistant	Analyst
##	3688	3602
##	Architect	Attorney
##	213	1046
##	Biologist	Bus Driver
##	125	316
##	Car Dealer	Chemist
##	180	145
##	Civil Service	Clergy
##	1457	196
##	Clerical	Computer Programmer
##	3164	4478
##	Construction	Dentist
##	1790	68
##	Doctor	Engineer - Chemical
##	494	225
##	Engineer - Electrical	Engineer - Mechanical
##	1125	1406

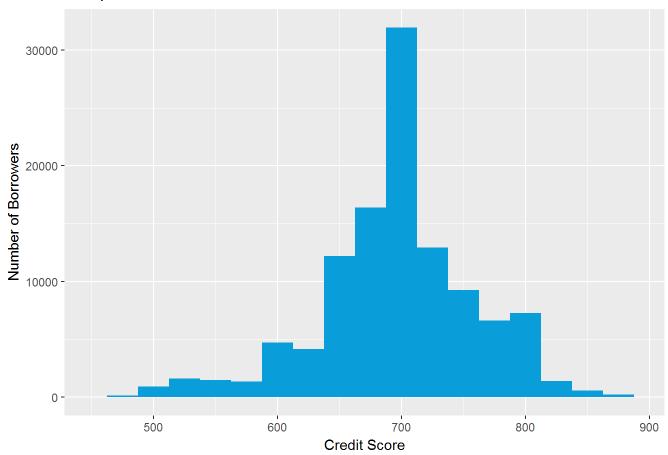
Fireman	Executive	##
422	4311	#
Food Service	Flight Attendant	##
1123	123	##
Homemaker	Food Service Management	##
120	1239	##
Judge	Investor	##
22	214	##
Landscaping	Laborer	##
236	1595	##
Military Enlisted	Medical Technician	##
1272	1117	##
Nurse's Aide	Military Officer	##
491	346	##
Nurse (RN)	Nurse (LPN)	##
2489	492	##
Pharmacist	Other	##
257	28617	##
Police Officer/Correction Officer	Pilot - Private/Commercial	##
1578	199	##
Principal	Postal Service	##
312	627	##
Professor	Professional	##
557	13628	##
Realtor	Psychologist	##
543	145	##
Retail Management	Religious	##
2602	124	##
Sales - Retail	Sales - Commission	##
2797	3446	##
Skilled Labor	Scientist	##
2746	372	##
Student - College Freshman	Social Worker	##
41	741	##
Student - College Junior	Student - College Graduate Student	
112	245	##
Student - College Sophomore	Student - College Senior	##
69	188	##
Student - Technical School	Student - Community College	##
16	28	
Teacher's Aide	zo Teacher	## ##
reacher's Alde	3759	## ##
Tradesman - Electrician	Tradesman - Carpenter	## ##
477	120	##
Tradesman - Plumber	Tradesman - Mechanic	##
102	951	##
Waiter/Waitress	Truck Driver	##
436	1675	##



Here we can see borrowers are from all over of different industry. There are also many people choose "Other" or "professional" not sepcifying their occupation.

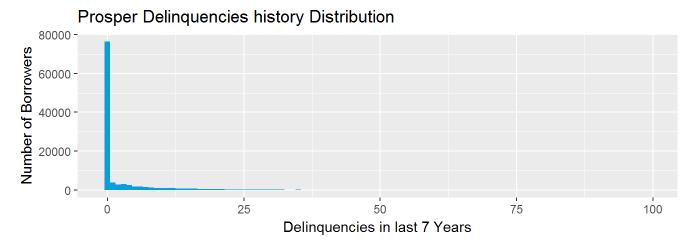
```
## Min. 1st Qu. Median Mean 3rd Qu. Max. NA's
## 9.5 669.5 689.5 695.1 729.5 889.5 591
## Warning: Removed 766 rows containing non-finite values (stat_bin).
## Warning: Removed 1 rows containing missing values (geom_bar).
```

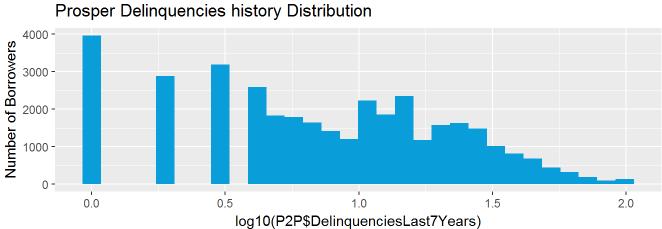
Prosper credit score of borrower Distribution



Here we could see a normal distribution chart. The main average of credit score from borrowers are around 700. Most borrowers are from 600-800.

```
## Min. 1st Qu. Median Mean 3rd Qu. Max. NA's
## 0.000 0.000 0.000 4.155 3.000 99.000 990
## Warning: Removed 990 rows containing non-finite values (stat_bin).
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
## Warning: Removed 77429 rows containing non-finite values (stat_bin).
```

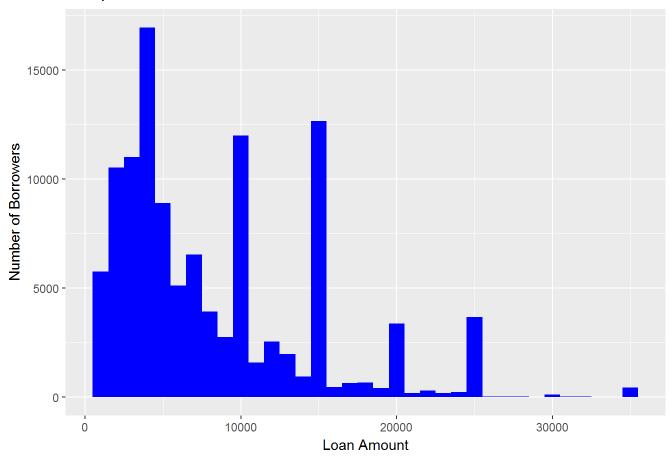




Here we could see a positive skewed chart and a log transformation chart. Most people do not have delinquency record. And people with few delinquency record can still get the loan.

##	Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
##	1000	4000	6500	8337	12000	35000

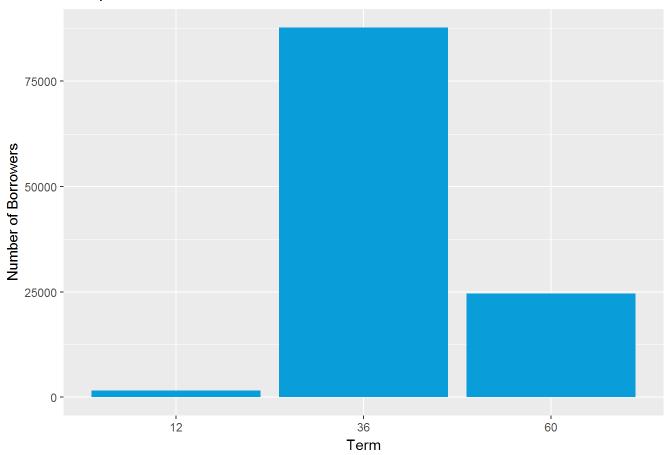
Prosper Loan Amount Distribution



Here we can see main loan amount are 4000, 10000 and 15000. And minimum loan amount is 1000.

```
## Min. 1st Qu. Median Mean 3rd Qu. Max.
## 12.00 36.00 36.00 40.83 36.00 60.00
```

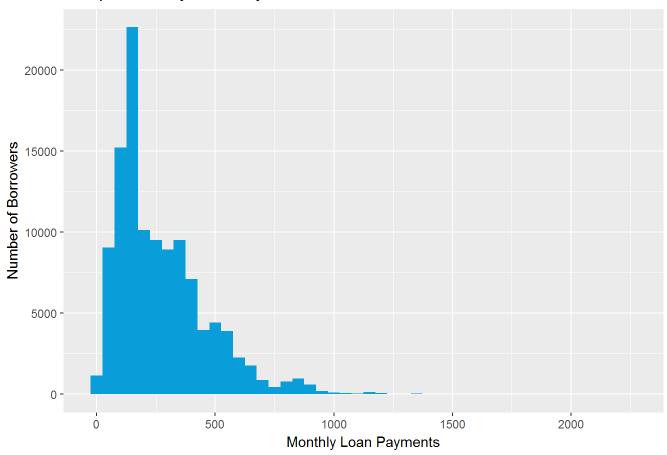
Prosper Term Distribution



Based on chart and summary, we could see most people choose 36 months, then 60 months and last 12 months. 36 months loan is more than tirple of 60 months loan.

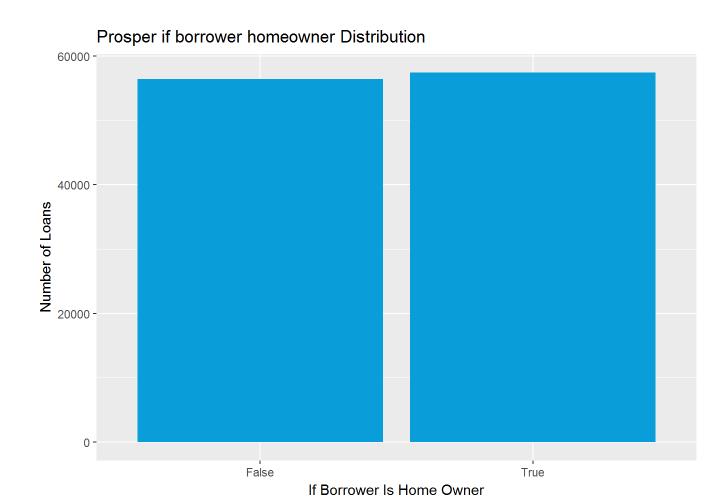
```
## Min. 1st Qu. Median Mean 3rd Qu. Max.
## 0.0 131.6 217.7 272.5 371.6 2252.0
```

Prosper Monthly Loan Payments Distribution



From this chart, we could see most borrower's monthly loan payments are between \$0-\$500. Mean and median monthly loan payment is \$272.50 and \$217.70.

False True ## 56459 57478



Here we could see almost half of borrower are homeowners and other half are not.

Univariate Analysis

What is the structure of your dataset?

The dataset has 87 variables and 113937 oberservations. And the loans cover the period from 2005 to 2014. The dataset contains classes int, numeric, data and factor.

What is/are the main feature(s) of interest in your dataset?

My main features of interest are: 1.Borrower rate. I am interested to know the relationship between borrower rate and other variables relationship. What factors will mostly affect borrower rate? Like prosper score,income range, loan amount etc. 2.Number of investors. I am interested to know which loans are more attractive for investing. What are the main characteristics of those loans?

What other features in the dataset do you think will help support your investigation into your feature(s) of interest?

Other features could be creidt score, debt to income ratio, prosper rating, occupation and employment status.

Did you create any new variables from existing variables in the dataset?

No, I did not

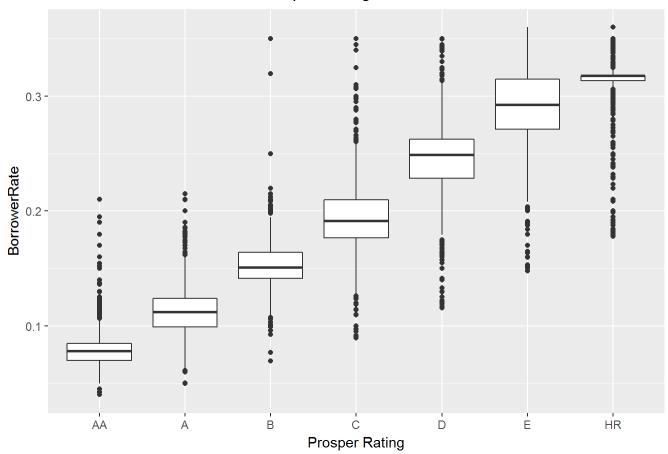
Of the features you investigated, were there any unusual distributions? Did you perform any operations on the data to tidy, adjust, or change the form of the data? If so, why did you do this?

DUring my investigation, I find some unusual distributions. When I do income range and job research, I find income lower than \$25000 and part-time job borrowers are not main borrowers. I'm wodering if prosper set up any income limits for borrowers. I did not perform any operations on the data.

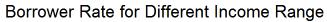
Bivariate Plots Section

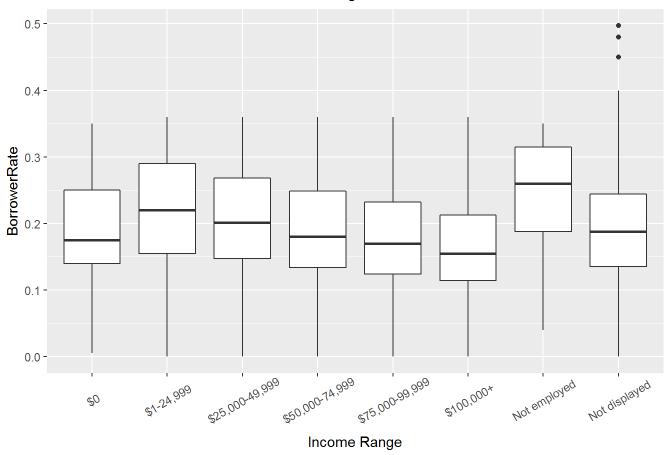
```
## P2P$ProsperRating..Alpha.: AA
   Min. 1st Qu. Median Mean 3rd Qu.
## 0.04000 0.06990 0.07790 0.07912 0.08450 0.21000
## P2P$ProsperRating..Alpha.: A
## Min. 1st Qu. Median Mean 3rd Qu. Max.
## 0.0498 0.0990 0.1119 0.1129 0.1239 0.2150
## ------
## P2P$ProsperRating..Alpha.: B
## Min. 1st Qu. Median Mean 3rd Qu.
## 0.0693 0.1414 0.1509 0.1545 0.1639 0.3500
## P2P$ProsperRating..Alpha.: C
## Min. 1st Qu. Median Mean 3rd Qu. Max.
## 0.0895 0.1765 0.1914 0.1944 0.2099 0.3500
## P2P$ProsperRating..Alpha.: D
## Min. 1st Qu. Median Mean 3rd Qu. Max.
## 0.1157 0.2287 0.2492 0.2464 0.2625 0.3500
## ------
## P2P$ProsperRating..Alpha.: E
## Min. 1st Qu. Median Mean 3rd Qu. Max.
```

Borrower Rate for Different Prosper Rating



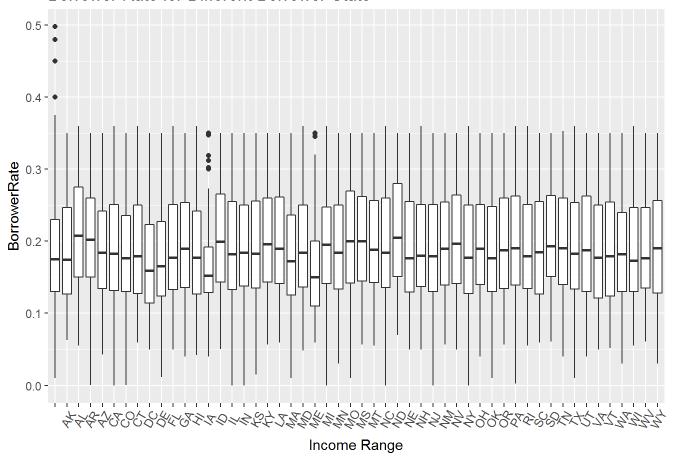
From this chart, we could see the reversed relationship between borrower rate and prosper rating. The higher prosper rating, the lower borrower rate.





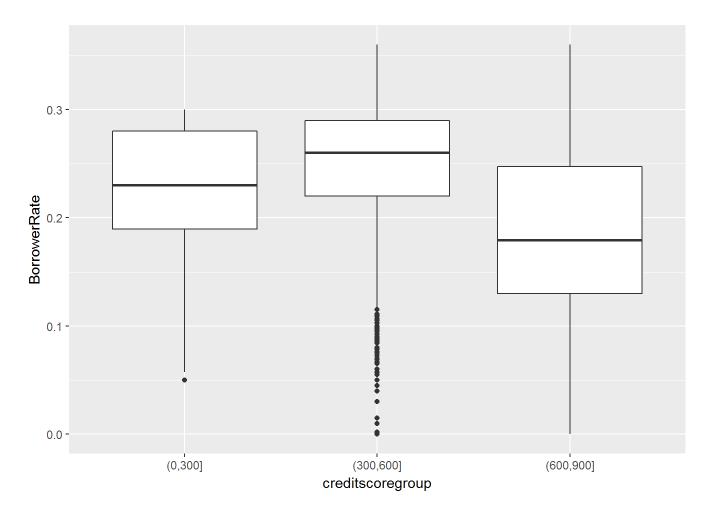
It looks like borrower with higher income will get lower borrower rate.

Borrower Rate for Different Borrower State



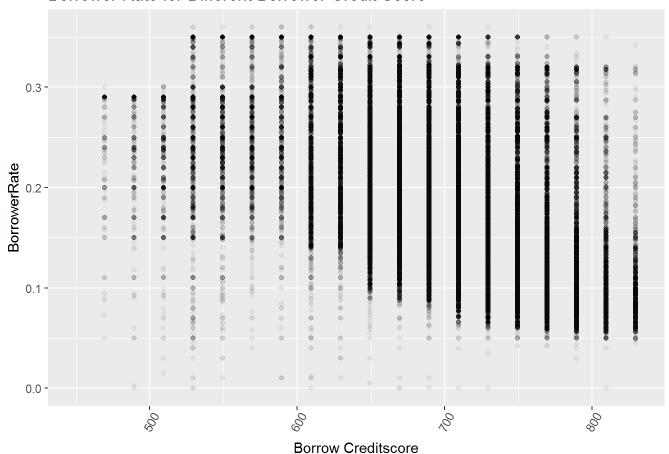
From this chart, we could see borrower rate has not a strong relationship with borrower state.

```
## (0,300] (300,600] (600,900] 
## 133 6632 106581
```



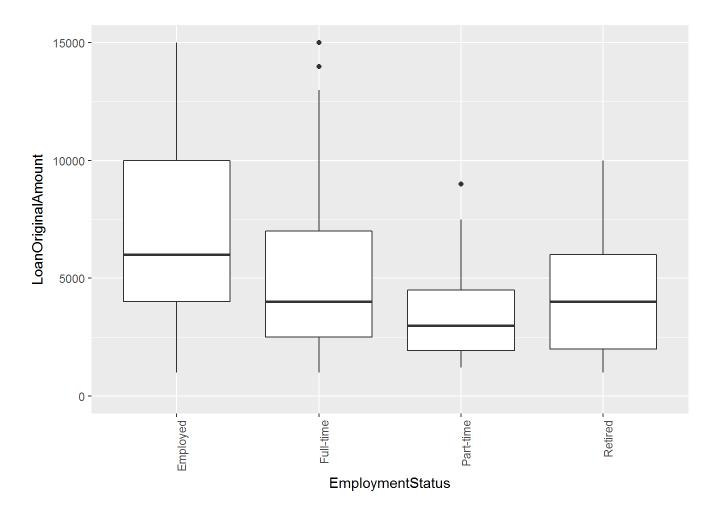
Warning: Removed 981 rows containing missing values (geom_point).

Borrower Rate for Different Borrower Credit Score



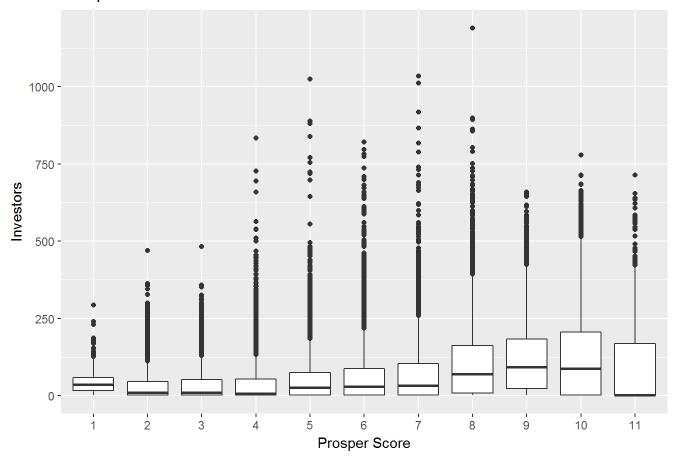
Here we use two charts to explore the realtionship between borrower credit score and borrower rate. From first chart, we could see the relationship between different credit score group and borrower rate. Borrower with 600-900 credit score has lowest borrower rate, which average is arount 0.18. From second chart, we could see a scotter plot chart with borrower's credit score 500 or above. We could clearly see that the higher credit score, the lower borrower rate.

Warning: Removed 22 rows containing non-finite values (stat boxplot).



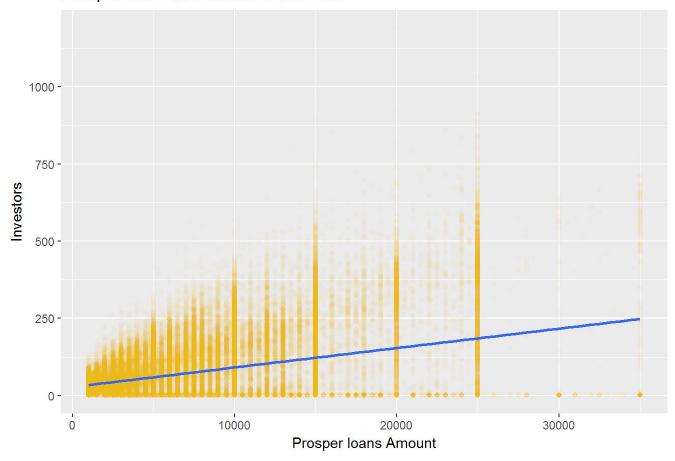
Employed and retired borrower seems to have higher loan amount than part-time.

Prosper Score vs Number of Investors



From the chart, we could see the higher the Prosper Score, the more investors. The peak is prosper socre equeal to 10.

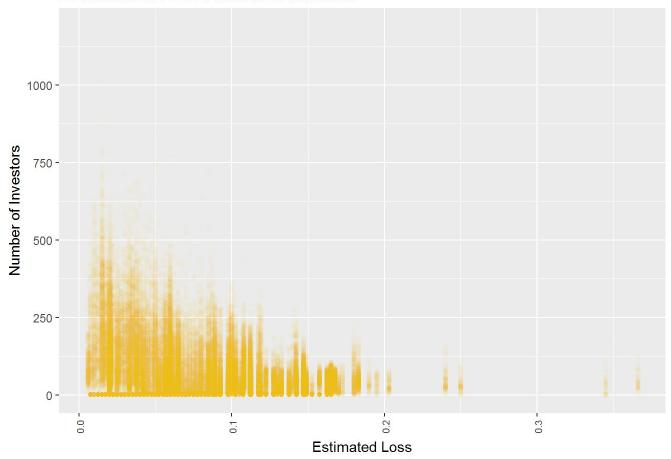
Prosper loans vs Number of Investors



Here we could see that the higher loans amount, the more investors.

Warning: Removed 29084 rows containing missing values (geom point).





We can found the lower the estmated loss, the more investors.

Bivariate Analysis

Talk about some of the relationships you observed in this part of the investigation. How did the feature(s) of interest vary with other features in the dataset?

1.Borrow Rate: There's reversed relationship between borrower rate and prosper rating. The higher prosper rating, the lower borrower rate. And borrower with higher income will get lower borrower rate. The higher credit score, the lower borrower rate. Employed and retired borrower seems to have higher loan amount than part-time. 2.Investors: We can find the higher the Prosper Score, the more investors. The peak is prosper socre equeal to 10. We also could see that the higher loans amount, the more investors. We can found the lower the estmated loss, the more investors.

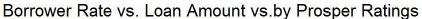
Did you observe any interesting relationships between the other features (not the main feature(s) of interest)?

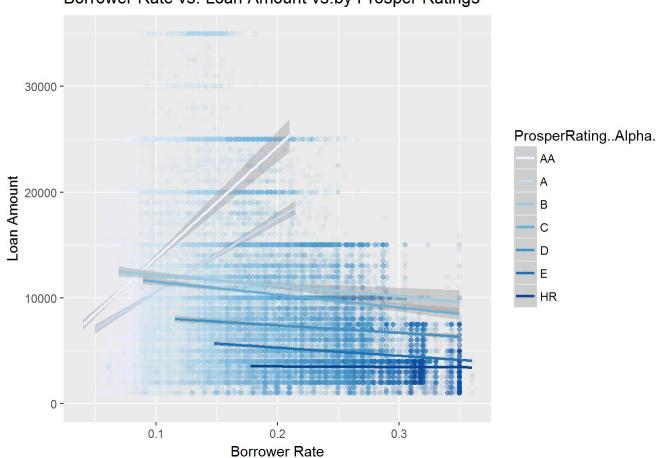
No

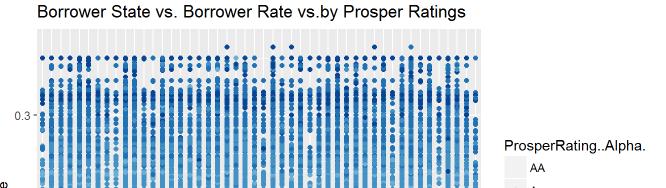
What was the strongest relationship you found?

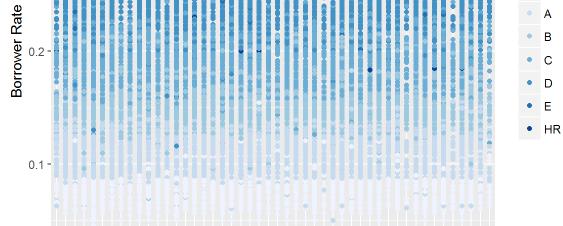
The reverse relationship between borrower rate and prosper rating.

Multivariate Plots Section





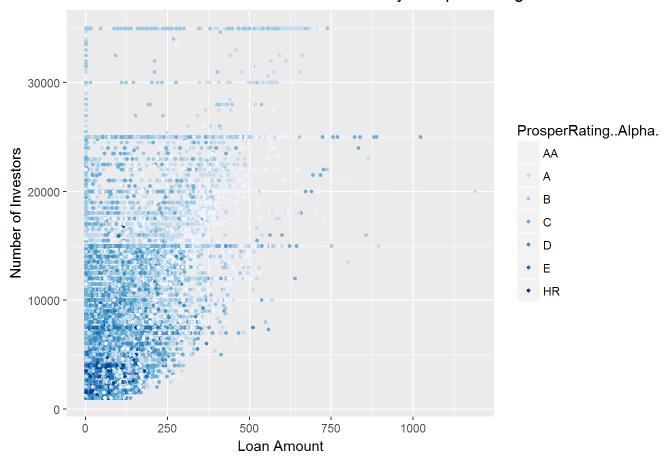




Each state is almost same, but still the lower the prosper rating, the higher borrower rate.

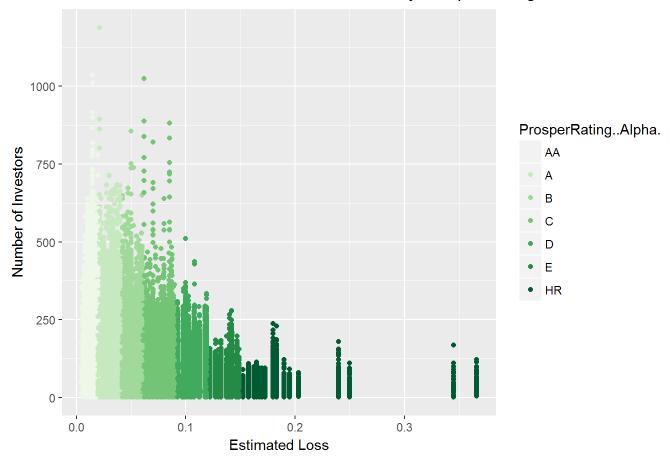
Borrower State

Number of Investors vs. Loan Amount vs. by Prosper Ratings



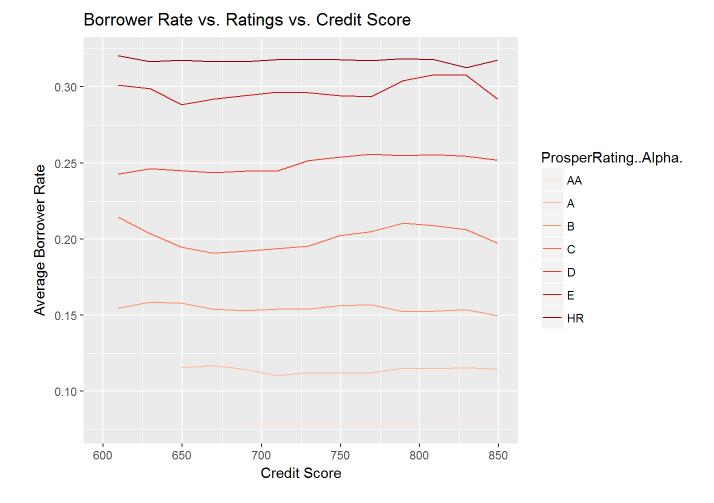
From this chart, we could see the relationship between Investors, loan amount and prosper ratings. The higher the prosper ratings, there will be more investors. However, most investors are interested in loan amount less than \$30,000. \$10,000 to \$20,000 are most popular investor's choice no matter what rating.

Number of Investors vs. Estimated Loss vs. by Prosper Ratings



The higher the prosper rating, the lower the estimated loss, the more the number of investors.

Warning: Removed 139 rows containing non-finite values (stat summary).



This chart summrize the relationship between borrower rate, rating and borrower credit score. We can see loans with high rating usually require higher credit score.

Multivariate Analysis

Talk about some of the relationships you observed in this part of the investigation. Were there features that strengthened each other in terms of looking at your feature(s) of interest?

Prosper rating and investors are strong correlated. Here we can see prosper rating, investors, loan amount and estimated loss are affecting each other.

Were there any interesting or surprising interactions between features?

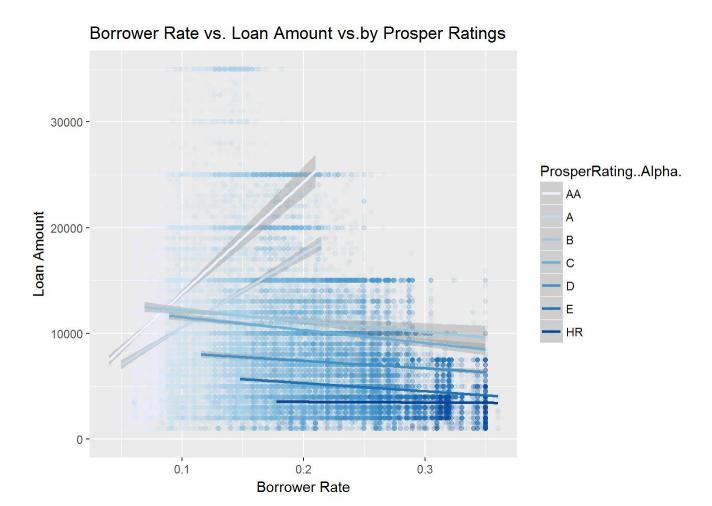
There were interesting interactions between prosper rating, loan amount and number of investors. It looks like up to \$25,000, higher loan amount means more investors, but when it comes above \$25,000, there were only few investors.

OPTIONAL: Did you create any models with your dataset? Discuss the strengths and limitations of your model.

No

Final Plots and Summary

Plot One

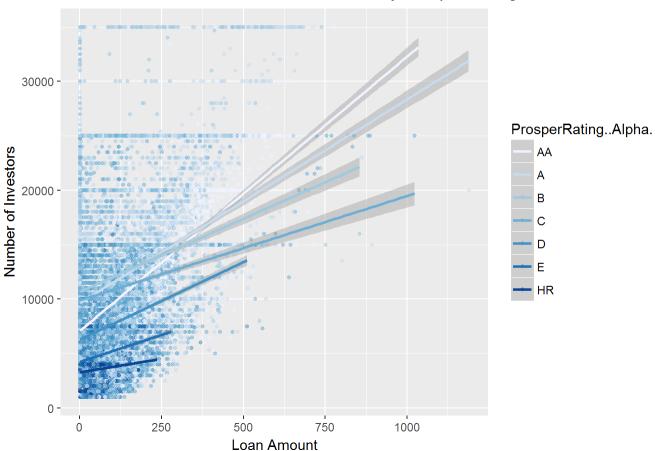


Description One

This chart basically summarize the realtionship betwwen borrower rate, loan amount and prosper ratings. We can clearly see that high rating loans usually got low borrower rate. And some large loan amount (more than \$25,000) are only available for high rating loans. And we also can see loans with higher borrower rate and lower rating usually has small loan amount.

Plot Two





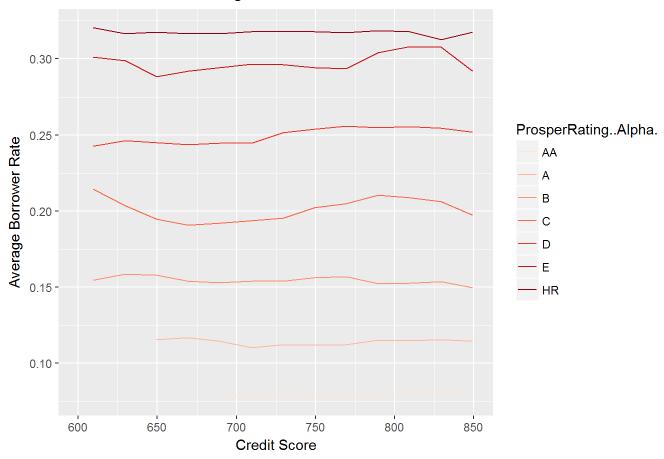
Description Two

This chart summrize the relationship between number of investors, loan amount and prosper ratings. We can see most investors are interested in investing loan amount less than \$25,000. For loans above \$25,000, investors are Only interested in high rating loans, like "A" or "AA". When investors was interested on low rating loans, they only invest loans with smaller loan amount (less than \$15,000). Probaly they are worried about risk with large loan amount.

Plot Three

Warning: Removed 139 rows containing non-finite values (stat summary).





Description Three

This chart summrize the relationship between borrower rate, rating and borrower credit score. We can see loans with high rating usually require higher credit score. "AA" rating loans require borrower's to have almost 700 credit score in order to qualify to apply. And other rating loan's minimum credit score start from 600. We also can see higher rating loans are always with lower borrower rate.

Reflection

I used to be an Investment Consultant for more than 6 years, so I had some general ideas about loans and investments. So I choose this database as my research topic. After I reviewed the dataset, I was interested to find the relationship between loan rating, borrower rate and number of investors. I wanted to learn more about P2P company and see what factors will affect investors to make their decisions.

I was new to computer system. So at the beginning, I was struggling with ggplot2 syntax. After lots of practise, reserching, reading and learning on Google, I gradually become comfortable with ggplot2 sytax. Before I did this project, I did not learn deeply about P2P company like Prosper.com. But, this topic was quite interesting to me and related to my previous job. I also checked Prosper.com to see the meaning of each column and how they operate their business. I also took time to read database and wanted to find most related factors.

Then I imported the databased and used the ggplot2 to see the correlation with each factors and variables. After I did two variable comprison and three variable comprison, I finlly find the relationship between borrower rate, number of investors, ratings, creidt socre and loan amount. I think these are important factors which will most affect the profit of the loan.

After I finished my project, I got some future ideas about this database. When I do the quick reserch, I realized that there was a global financial crisis in 2008, but it was not obvious on these database. So if I will continue to do future work on this dataset, I will split the timeframe between 2004-2009 and 2009-2014 to see if there was any financial pressure or effects on number of borrowers, borrower rate and number of investors.