

LendingClub End-to-End Project | Part 1 – Data Wrangling

```
1  /* Part 1: Data Wrangling for Lending Club Project
2    Data set: https://www.kaggle.com/datasets/ethon0426/lending-club-20072020q1
3    A raw subset for the 2018 year was created from the linked dataset and now I will clean it
4
5    loan_data_2018_RAW.csv is imported as a flat file with Microsoft SQL Server Management Studio
6    where no primary key is selected, and where NULL values are allowed in every column
7
8    Data cleaning steps:
9    1: Remove all duplicates
10   2: Standardize data
11   3: Manage NULL values and remove unnecessary rows/columns
12  */
13
14  USE loan_data_database
15
16  /*
17   1: Remove all duplicates
18  */
19
20  -----
21
22  -- NOTE: N is written before every string to declare
23  --       it as nvarchar as opposed to varchar for unicode
24  --       this is done here with the staging table and in every procedure created further below
25
26  -- create staging table by duplicating the raw table format and copying all raw data into it
27  -- this way if any mistakes are made, the untouched raw data table will still exist to reference
28  IF NOT EXISTS (SELECT * FROM sys.objects
29  WHERE [object_id] = OBJECT_ID(N'[dbo].[loan_data_2018_staging]') AND type in (N'U'))
30  SELECT * INTO loan_data_2018_staging FROM loan_data_2018_RAW
31
32  (57171 rows affected)
33
34  + dbo.loan_data_2018_RAW
35  + dbo.loan_data_2018_staging
```

```

38 -- create a CTE to isolate duplicate entries, where the data in all columns is exactly the same
39 -- 8 duplicate rows exist
40 WITH cte_duplicates AS
41 (
42     SELECT *,
43     ROW_NUMBER() OVER (
44         PARTITION BY id, loan_amount, term, interest_rate, installment, grade, sub_grade, employee_length, home_ownership,
45         annual_income, issue_date, loan_status, purpose, purpose2, address_state, debt_to_income, total_payment ORDER BY (SELECT NULL)) AS row_num
46     FROM loan_data_2018_staging
47 )
48 SELECT *
49 FROM cte_duplicates
50 WHERE row_num > 1

```

	id	loan_amount	term	interest_rate	installment	grade	sub_grade	employee_length	home_ownership	annual_income	issue_date	loan_status	purpose	purpose2	address_state	debt_to_income	total_payment	row_num
1	127935939	15750	36 months	0.200000002980232	585.330017089844	D	D4	5 years	Mortgage	83083.5234375	2018-02-18	Charged Off	Car	NULL	NC	0.200900003314018	9933.0400390625	2
2	129827583	24000	60 months	0.160099998116493	583.77001953125	C	C5	10+ years	Mortgage	100000	2018-03-18	Fully Paid	Debt	Consolidation	CA	0.320100009441376	28774.87890625	2
3	130464147	40000	36 months	0.119800001382828	1328.19995117188	B	B5	1 year	Mortgage	81000	2018-03-18	Fully Paid	Debt	Consolidation	AZ	0.318800002336502	42033.22265625	2
4	134042794	24800	60 months	0.0957999974489212	521.820007324219	B	B1	3 years	Mortgage	81323	2018-05-18	Fully Paid	Debt	Consolidation	MD	0.154100000858307	25036.634765625	2
5	139100306	15000	36 months	0.20890000462532	564.280029296875	D	D4	4 years	Rent	85000	2018-08-18	Charged Off	Other	NULL	TX	0.196099996566772	6727.83984375	2
6	140978011	25000	36 months	0.110600002110004	819.179992675781	B	B3	< 1 year	Rent	90000	2018-10-18	Current	Debt	Consolidation	WV	0.179900005459785	15610.509765625	2
7	144295663	14000	36 months	0.0755999982357025	435.880004882813	A	A3	3 years	Mortgage	85000	2018-11-18	Fully Paid	Debt	Consolidation	AK	0.121100001037121	15016.6494140625	2
8	144488901	17500	36 months	0.103299997746944	567.400024414063	B	B1	< 1 year	Rent	56786	2018-12-18	Current	Credit	Card	CA	0.172399997711182	9685.98046875	2

```

52 -- create a derived table with row_num column to count and delete duplicate rows that were isolated in the CTE above
53 -- 8 duplicate rows removed
54 DELETE dt_duplicates
55 FROM
56 (
57     SELECT *, row_num = ROW_NUMBER() OVER (PARTITION BY id, loan_amount, term, interest_rate, installment,
58     grade, sub_grade, employee_length, home_ownership, annual_income, issue_date, loan_status, purpose, purpose2,
59     address_state, debt_to_income, total_payment ORDER BY (SELECT NULL))
60     FROM loan_data_2018_staging
61 ) AS dt_duplicates
62 WHERE row_num > 1

```

(8 rows affected)

```

64 -- 57171 rows in raw table, 57163 rows in staging table, 8 duplicate rows removed
65 SELECT *
66 FROM loan_data_2018_staging

```

0.187800005078316	15819.2001953125	
(16.0 RTM)	-_ (60)	loan_data_database 00:00:01 57,163 rows

```
68 -----
69 /*
70 2: Standardize Data
71 */
72 -----
73
74 -- removing any potential trailing spaces from string fields
75 /*
76 UPDATE loan_data_2018_staging2
77 SET term = TRIM(term),
78     grade = TRIM(grade),
79     ... etc ... manually list out all string columns to trim
80 */
81 |
82 /*
83 NOTE:
84     the above approach for trimming string fields works, but it would
85     become impractical when dealing with a table with a large number
86     of string fields, therefore I will write a script to trim all
87     string fields to solve this efficiency issue
88 */
89
90 -----
```

```

90  -----
91  /*
92   stored procedure to trim all columns in a table
93  */
94  -----
95  IF EXISTS
96  (
97      SELECT type_desc, type
98      FROM sys.procedures WITH(NOLOCK)
99      WHERE NAME = 'trim_string_fields' AND type = 'P'
100  )
101  DROP PROCEDURE trim_string_fields
102  GO
103  CREATE PROCEDURE trim_string_fields @table nvarchar(max) AS DECLARE @query AS nvarchar(max)
104
105  -- trim all string values in every column and show which columns were trimmed in output
106  SET @query = STUFF((SELECT N', ' + QUOTENAME([name]) + N' = TRIM(' + QUOTENAME([name]) + N')'
107  FROM sys.columns
108  WHERE [object_id] = OBJECT_ID(@table) AND [system_type_id] IN(29,35,99,167,175,231)
109  FOR XML PATH(''),1,1,''))
110
111  SET @query = N'UPDATE ' + @table + N' SET' + @query
112  PRINT @query
113
114  EXEC(@query)
115  GO
116  -----
117  -- test the script by adding trailing/leading spaces to all 'PA' state values
118  -- (1880 rows affected) trailing space address_state
119  UPDATE loan_data_2018_staging
120  SET address_state = ' PA '
121  WHERE address_state = 'PA'
122
123  -- we can see that all 'PA' values now have the trailing/leading spaces added
124  SELECT * FROM loan_data_2018_staging

```

	id	loan_amount	term	interest_rate	installment	grade	sub_grade	employee_length	home_ownership	annual_income	issue_date	loan_status	purpose	purpose2	address_state	debt_to_income	total_payment
1	138408587	10000	36 months	0.0666999965906143	307.269989013672	A	A2	6 years	Rent	28787	2018-08-18	Current	Credit	Card	PA	0.158899992704391	6141.68994140625
2	138228354	5000	36 months	0.0846000015735626	157.75	A	A5	10+ years	Mortgage	35000	2018-08-18	Fully Paid	Debt	Consolidation	WV	0.269499987363815	5341.96337890625
3	138384636	40000	60 months	0.0784000009298325	808	A	A4	3 years	Mortgage	480000	2018-08-18	Current	Small	Business	FL	0.124600000679493	16950.580078125

```

126 -- run procedure to trim all columns including our test above
127 EXEC trim_string_fields loan_data_2018_staging
128

```

```

129 -- after executing the trim script above, we see that it works
130 SELECT * FROM loan_data_2018_staging
131

```

	id	loan_amount	term	interest_rate	installment	grade	sub_grade	employee_length	home_ownership	annual_income	issue_date	loan_status	purpose	purpose2	address_state	debt_to_income	total_payment
1	138408587	10000	36 months	0.0666999965906143	307.269989013672	A	A2	6 years	Rent	28787	2018-08-18	Current	Credit	Card	PA	0.158899992704391	6141.68994140625
2	138228354	5000	36 months	0.0846000015735626	157.75	A	A5	10+ years	Mortgage	35000	2018-08-18	Fully Paid	Debt	Consolidation	WV	0.269499987363815	5341.96337890625
3	138384636	40000	60 months	0.0784000009298325	808	A	A4	3 years	Mortgage	480000	2018-08-18	Current	Small	Business	FL	0.124600000679493	16950.580078125

```

131 -----
132
133 -- visually inspect all string field categories to check for errors/inconsistencies
134

```

```

135 -- it appears the purpose field was split into two columns

```

```

136 SELECT DISTINCT purpose, purpose2
137 FROM loan_data_2018_staging
138 ORDER BY purpose2

```

	purpose	purpose2
7	Wedding	NULL
8	Moving	NULL
9	Other.	NULL
10	Vacation	NULL
11	NULL	NULL
12	Small	Business
13	Credit	Card

```

140 -- by joining purpose1 with purpose2 on the id, we prove that 6 categories need to be merged, namely:
141 -- "Credit Card", "Debt Consolidation", "Home Improvement", "Major Purchase",
142 -- "Renewable Energy", and "Small Business"

```

```

143 SELECT DISTINCT t1.purpose, t2.purpose2
144 FROM loan_data_2018_staging t1
145 FULL OUTER JOIN loan_data_2018_staging t2
146 ON t1.id = t2.id
147

```

```

148 -- purpose_merged gives us the merged columns as they should be

```

```

149 SELECT DISTINCT purpose, purpose2, ISNULL(purpose, '') + ' ' + ISNULL(purpose2, '') AS purpose_merged
150 FROM loan_data_2018_staging

```




	purpose	purpose2	purpose_merged
1	Small	Business	Small Business
2	Vacation	NULL	Vacation
3	Major	Purchase	Major Purchase
4	Debt	Consolidation	Debt Consolidation
5	Renewable	Energy	Renewable Energy
6	NULL	NULL	
7	Medical	NULL	Medical
8	Wedding	NULL	Wedding
9	.Other.	NULL	.Other.
10	Credit	Card	Credit Card
11	Other	NULL	Other
12	Other.	NULL	Other.
13	Moving	NULL	Moving
14	Home	Improvement	Home Improvem...

```

152 | -- create a second staging table with a consolidated purpose column (in the same position) to store the merged values
153 | IF NOT EXISTS (SELECT * FROM sys.objects
154 | WHERE [object_id] = OBJECT_ID(N'[dbo].[loan_data_2018_staging2]') AND type in (N'U'))
155 | CREATE TABLE [dbo].[loan_data_2018_staging2](
156 |     [id] [int] NULL,
157 |     [loan_amount] [int] NULL,
158 |     [term] [nvarchar](50) NULL,
159 |     [interest_rate] [float] NULL,
160 |     [installment] [float] NULL,
161 |     [grade] [nvarchar](50) NULL,
162 |     [sub_grade] [nvarchar](50) NULL,
163 |     [employee_length] [nvarchar](50) NULL,
164 |     [home_ownership] [nvarchar](50) NULL,
165 |     [annual_income] [float] NULL,
166 |     [issue_date] [date] NULL,
167 |     [loan_status] [nvarchar](50) NULL,
168 |     [purpose] [nvarchar](50) NULL,
169 |     [address_state] [nvarchar](50) NULL,
170 |     [debt_to_income] [float] NULL,
171 |     [total_payment] [float] NULL,
172 | ) ON [PRIMARY]

```

Commands completed successfully.

- +  dbo.loan_data_2018_RAW
- +  dbo.loan_data_2018_staging
- +  **dbo.loan_data_2018_staging2**

```

174 | -- copy all needed fields from the original staging table into the second staging table
175 | INSERT INTO loan_data_2018_staging2
176 | SELECT id, loan_amount, term, interest_rate, installment, grade, sub_grade, employee_length, home_ownership,
177 |        annual_income, issue_date, loan_status, ISNULL(purpose, '') + ' ' + ISNULL(purpose2, '') AS purpose,
178 |        address_state, debt_to_income, total_payment
179 | FROM loan_data_2018_staging
180 |
181 | -- rerun procedure to trim all columns on the newly created loan_data_2018_staging2
182 | -- since values in the first purpose column that didn't have any corresponding value in the
183 | -- 2nd purpose column will be left with a trailing ' ' space
184 | EXEC trim_string_fields loan_data_2018_staging2
185 |
186 | -- all purpose categories have been merged into one and only 1 purpose column remains
187 | SELECT DISTINCT purpose
188 | FROM loan_data_2018_staging2
189 | ORDER BY purpose
190 |
191 | -- I will continue working with the 2nd staging table that has the most cleaned data

```


	purpose
1	
2	..Other.
3	.Other.
4	Car
5	Credit Card
6	Debt Consolidation
7	Home Improvement
8	House
9	Major Purchase
10	Medical
11	Moving
12	Other
13	Other.
14	Renewable Energy
15	Small Business
16	Vacation
17	Wedding

```

193 | -- purpose has some rows where the Other category has leading/trailing punctuation '.' marks
194 | SELECT DISTINCT purpose
195 | FROM loan_data_2018_staging2
196 | ORDER BY purpose
197 |
198 | -- remove all leading/trailing '.' dots from the purpose field and rerun query above to confirm
199 | UPDATE loan_data_2018_staging2
200 | SET purpose = TRIM('.', FROM purpose)

```

	purpose
1	
2	Car
3	Credit Card
4	Debt Consolidation
5	Home Improvement
6	House
7	Major Purchase
8	Medical
9	Moving
10	Other
11	Renewable Energy
12	Small Business
13	Vacation
14	Wedding

```
202 SELECT DISTINCT term
203 FROM loan_data_2018_staging2
204 ORDER BY term
```

```
206 SELECT DISTINCT grade
207 FROM loan_data_2018_staging2
208 ORDER BY grade
```

```
210 SELECT DISTINCT sub_grade
211 FROM loan_data_2018_staging2
212 ORDER BY sub_grade
```

```
214 SELECT DISTINCT employee_length
215 FROM loan_data_2018_staging2
216 ORDER BY employee_length
```

```
217
218 -- home_ownership has 2 redundant Rent labels as 'Renter/Renting' which need to be consolidated to Rent
```

```
219 SELECT DISTINCT home_ownership
220 FROM loan_data_2018_staging2
221 ORDER BY home_ownership
```

	home_ownership
1	NULL
2	Mortgage
3	Own
4	Rent
5	Renter
6	Renting

```

223 | -- fixed 14 rows showing as Renter or Renting and renamed to Rent, rerun query above to confirm only Rent remains
224 | UPDATE loan_data_2018_staging2
225 | SET home_ownership = 'Rent'
226 | WHERE home_ownership IN ('Renter', 'Renting')

```

(14 rows affected)

	home_ownership
1	NULL
2	Mortgage
3	Own
4	Rent

```
228 SELECT DISTINCT loan_status
229 FROM loan_data_2018_staging2
230 ORDER BY loan_status
231
232 SELECT DISTINCT address_state
233 FROM loan_data_2018_staging2
234 ORDER BY address_state
235
236 -- issue dates are already standardized so the part below will be commented out
237 SELECT DISTINCT issue_date
238 FROM loan_data_2018_staging2
239 ORDER BY issue_date
240
241 /*
242 -- format all date entries to MMMM-yyyy format
243 SELECT DISTINCT FORMAT(issue_date, 'MMMM-yyyy') AS formatted_issue_date
244 FROM loan_data_2018_staging2
245 ORDER BY formatted_issue_date
246
247 -- update issue_date field to standardize date format
248 UPDATE loan_data_2018_staging2
249 SET issue_date = FORMAT(issue_date, 'MMMM-yyyy')
250 WHERE issue_date IS NOT NULL -- to ensure we don't try to format NULL values
251 */
```

	issue_date
1	NULL
2	2018-01-18
3	2018-02-18
4	2018-03-18
5	2018-04-18
6	2018-05-18
7	2018-06-18
8	2018-07-18
9	2018-08-18
10	2018-09-18
11	2018-10-18
12	2018-11-18
13	2018-12-18

```

253 | -- ensure there aren't any negative numerical values present; 0 found
254 | SELECT loan_amount, interest_rate, installment, debt_to_income, total_payment
255 | FROM loan_data_2018_staging2
256 | WHERE loan_amount < 0 OR interest_rate < 0 OR installment < 0 OR debt_to_income < 0 OR total_payment < 0

```

loan_amount	interest_rate	installment	debt_to_income	total_payment
-------------	---------------	-------------	----------------	---------------

0 rows

```

258  */
259  3: Manage NULL values and remove unnecessary rows/columns
260
261  NOTE:
262      I will write a script to find any potential NULL and/or
263      empty values for the same efficiency reason that I wrote
264      one to trim all columns
265  */
266
267  -----
268  /*
269      stored procedure to display all null values in a table
270  */
271  -----
272  IF EXISTS
273  (
274      SELECT type_desc, type
275      FROM sys.procedures WITH(NOLOCK)
276      WHERE NAME = 'select_null_values' AND type = 'P'
277  )
278  DROP PROCEDURE select_null_values
279  GO
280  CREATE PROCEDURE select_null_values @table nvarchar(max) AS DECLARE @query AS nvarchar(max)
281
282  -- extract the table schema without extracting any data within
283  SET @query = N'SELECT * FROM ' + @table + N' WHERE 1 = 0'
284
285  -- check for NULL values in every column
286  SELECT @query += N' OR ' + QUOTENAME(name) + N' IS NULL'
287  FROM sys.columns
288  WHERE [object_id] = OBJECT_ID(@table)
289
290  EXEC(@query)
291  GO

```

```

292 |-----
293 | -- run procedure to select NULL values in all columns, 3 rows with NULL data exist
294 | EXEC select_null_values loan_data_2018_staging2

```

	id	loan_amount	term	interest_rate	installment	grade	sub_grade	employee_length	home_ownership	annual_income	issue_date	loan_status	purpose	address_state	debt_to_income	total_payment
1	137917888	NULL	NULL	NULL	NULL	NULL	NULL	NULL	NULL	NULL	NULL	NULL		NULL	NULL	NULL
2	139305488	NULL	NULL	NULL	NULL	NULL	NULL	NULL	NULL	NULL	NULL	NULL		NULL	NULL	NULL
3	131407188	NULL	NULL	NULL	NULL	NULL	NULL	NULL	NULL	NULL	NULL	NULL		NULL	NULL	NULL

```

296 | -- delete 3 rows with NULL data
297 | -- these same 3 rows were also causing the NULL value to appear in several columns
298 | DELETE
299 | FROM loan_data_2018_staging2
300 | WHERE id = 131407188 OR id = 137917888 OR id = 139305488
301 |
302 | -- running one more time shows that no NULL values remain in the table
303 | EXEC select_null_values loan_data_2018_staging2

```

(3 rows affected)

id	loan_amount	term	interest_rate	installment	grade	sub_grade	employee_length	home_ownership	annual_income	issue_date	loan_status	purpose	address_state	debt_to_income	total_payment

```

305 | -- the annual_income column is not needed for my analysis so it must be removed
306 | ALTER TABLE loan_data_2018_staging2
307 | DROP COLUMN IF EXISTS annual_income

```

Commands completed successfully.

```

326 -----
327 /*
328     stored procedure to display all empty '' values in a table
329 */
330 -----
331 IF EXISTS
332 (
333     SELECT type_desc, type
334     FROM sys.procedures WITH(NOLOCK)
335     WHERE NAME = 'select_empty_values' AND type = 'P'
336 )
337 DROP PROCEDURE select_empty_values
338 GO
339 CREATE PROCEDURE select_empty_values @table nvarchar(max) AS DECLARE @query AS nvarchar(max)
340
341 SET @query = N'SELECT * FROM ' + @table + N' WHERE 1 = 0'
342
343 -- check for empty '' values (written as '''' since escape characters are needed) in every column
344 SELECT @query += N' OR ' + QUOTENAME(name) + N' LIKE ''''''
345 FROM sys.columns
346 WHERE [object_id] = OBJECT_ID(@table)
347
348 EXEC(@query)
349 GO
350 -----
351
352 -- no empty values remain in the table
353 EXEC select_empty_values loan_data_2018_staging2




```

id	loan_amount	term	interest_rate	installment	grade	sub_grade	employee_length	home_ownership	issue_date	loan_status	purpose	address_state	debt_to_income	total_payment

```

376 -- rename final cleaned staging table
377 EXEC sp_rename 'loan_data_2018_staging2', 'loan_data_2018'

```

- ⊕  **dbo.loan_data_2018**
- ⊕  dbo.loan_data_2018_RAW
- ⊕  dbo.loan_data_2018_staging


```

380 -- 57160 rows with 15 columns in final cleaned table
381 SELECT * FROM loan_data_2018
382 ORDER BY id

```

	id	loan_amount	term	interest_rate	installment	grade	sub_grade	employee_length	home_ownership	issue_date	loan_status	purpose	address_state	debt_to_income	total_payment
1	118187703	40000	36 months	0.0671999976038933	1229.96997070313	A	A3	10+ years	Mortgage	2018-01-18	Current	Credit Card	MD	0.220400005578995	34424.23046875
2	119987870	11200	60 months	0.19030000269413	290.720001220703	D	D3	10+ years	Rent	2018-02-18	Fully Paid	Debt Consolidation	RI	0.174999997019768	12192.6875
3	120068032	20475	60 months	0.180600002408028	520.599975585938	D	D2	10+ years	Mortgage	2018-01-18	Current	Debt Consolidation	VA	0.18520000576973	14556.259765625
4	120640856	40000	36 months	0.0531999990344048	1204.59997558594	A	A1	1 year	Mortgage	2018-01-18	Fully Paid	Small Business	TX	0.112999998033047	41776.55859375
5	121263466	25000	36 months	0.0797000005841255	783.070007324219	A	A5	3 years	Mortgage	2018-01-18	Fully Paid	Debt Consolidation	MD	0.28099998831749	27557.921875
6	121703857	30000	60 months	0.150399997830391	714.330017089844	C	C4	10+ years	Mortgage	2018-04-18	Current	Debt Consolidation	MN	0.273299992084503	18887.169921875
7	121785745	6000	36 months	0.15049999523163	208.139999389648	C	C4	7 years	Rent	2018-01-18	Current	Other	NI	0.149000003933907	5819.81982421875

Query executed successfully. (16.0 RTM) (60) loan_data_database 00:00:01 57,160 rows

LendingClub End-to-End Project | Part 2 – Exploratory Data Analysis

```

389 /* Part 2: Exploratory Data Analysis for Lending Club Project
390 Data set: https://www.kaggle.com/datasets/ethon0426/lending-club-20072020q1
391 A raw subset for the 2018 year was cleaned in Part 1 of the Project
392
393 Now I will perform exploratory data analysis on the cleaned subset
394 */
395
396 USE loan_data_database
397
398 -- 57160 total loan applications issued, 3253 issued in December alone, and 3606 in November
399 SELECT COUNT(id) AS total_loan_applications FROM loan_data_2018
400 SELECT COUNT(id) AS MTD_total_loan_applications FROM loan_data_2018 WHERE MONTH(issue_date) = 12
401 SELECT COUNT(id) AS prior_MTD_total_loan_applications FROM loan_data_2018 WHERE MONTH(issue_date) = 11
402

```

	total_loan_applications
1	57160

	MTD_total_loan_applications
1	3253

	prior_MTD_total_loan_applications
1	3606

```

403 | -- $1,138.2M in total funded loans, $65M funded in December alone, and $74M in November
404 | SELECT FORMAT(SUM(loan_amount), 'C', 'en-US') AS total_funded_amount FROM loan_data_2018
405 | SELECT FORMAT(SUM(loan_amount), 'C', 'en-US') AS MTD_total_funded_amount FROM loan_data_2018 WHERE MONTH(issue_date) = 12
406 | SELECT FORMAT(SUM(loan_amount), 'C', 'en-US') AS prior_MTD_total_funded_amount FROM loan_data_2018 WHERE MONTH(issue_date) = 11

```

total_funded_amount	
1	\$1,138,202,000.00

MTD_total_funded_amount	
1	\$65,045,200.00

prior_MTD_total_funded_amount	
1	\$74,097,350.00

```

408 | -- $840M in total payments received, $38.5M received in December alone, and $45.5M in November
409 | -- 2018 has not seen overall profitability as the total payments received are 73.8% of the total funded loans
410 | SELECT FORMAT(SUM(total_payment), 'C', 'en-US') AS total_amount_received FROM loan_data_2018
411 | SELECT FORMAT(SUM(total_payment), 'C', 'en-US') AS MTD_total_amount_received FROM loan_data_2018 WHERE MONTH(issue_date) = 12
412 | SELECT FORMAT(SUM(total_payment), 'C', 'en-US') AS prior_MTD_total_amount_received FROM loan_data_2018 WHERE MONTH(issue_date) = 11

```

total_amount_received	
1	\$840,021,972.47

MTD_total_amount_received	
1	\$38,506,403.32

prior_MTD_total_amount_received	
1	\$45,535,225.22

```

414 | -- avg interest rate of 14.75% for 2018, avg 14.33% interest rate in December alone, and 14.29% in in November
415 | SELECT ROUND(AVG(interest_rate) * 100, 2) AS avg_interest_rate FROM loan_data_2018
416 | SELECT ROUND(AVG(interest_rate) * 100, 2) AS MTD_avg_interest_rate FROM loan_data_2018 WHERE MONTH(issue_date) = 12
417 | SELECT ROUND(AVG(interest_rate) * 100, 2) AS prior_MTD_avg_interest_rate FROM loan_data_2018 WHERE MONTH(issue_date) = 11

```

	avg_interest_rate
1	14.75

	MTD_avg_interest_rate
1	14.33

	prior_MTD_avg_interest_rate
1	14.29

```

419 | -- avg DTI of 20.97% for 2018, 20.84% avg DTI in December alone, and 20.89% in November
420 | SELECT ROUND(AVG(debt_to_income) * 100, 2) AS avg_debt_to_income FROM loan_data_2018
421 | SELECT ROUND(AVG(debt_to_income) * 100, 2) AS MTD_avg_debt_to_income FROM loan_data_2018 WHERE MONTH(issue_date) = 12
422 | SELECT ROUND(AVG(debt_to_income) * 100, 2) AS prior_MTD_avg_debt_to_income FROM loan_data_2018 WHERE MONTH(issue_date) = 11

```

	avg_debt_to_income
1	20.97

	MTD_avg_debt_to_income
1	20.84

	prior_MTD_avg_debt_to_income
1	20.89

```

425 | -- 12.6% of all loans issued have been charged off, where this accounts for 7200 loans and $145.3M in funds where $54.5M was paid
426 | -- hence $90.8M total was charged off in 2018
427 | SELECT
428 |     (SELECT (COUNT(CASE WHEN loan_status = 'Charged Off' THEN id END) * 100.0) / COUNT(id) FROM loan_data_2018) AS bad_loan_percentage,
429 |     COUNT(id) AS bad_loan_applications,
430 |     FORMAT(SUM(loan_amount), 'C', 'en-US') AS bad_loan_funded_amount,
431 |     FORMAT(SUM(total_payment), 'C', 'en-US') AS bad_loan_amount_received,
432 |     FORMAT(SUM(loan_amount) - SUM(total_payment), 'C', 'en-US') AS total_charged_off
433 | FROM loan_data_2018
434 | WHERE loan_status = 'Charged Off'

```

	bad_loan_percentage	bad_loan_applications	bad_loan_funded_amount	bad_loan_amount_received	total_charged_off
1	12.555983205038	7177	\$145,335,150.00	\$54,515,127.14	\$90,820,022.86

```

436 -- 87.4% of all loans issued have been fully paid or are current, where this accounts for 50000 loans and $992.9M in funds where $785.5M was paid
437 -- hence $237.1M is the approximate total outstanding loan balance for 2018 ($207.4M remaining * avg annual interest of 14.7%)
438 -- outstanding loan balance does not account for fees, penalties, compounding, and other hidden charges as we don't have that data in the subset
439 SELECT
440     (SELECT (COUNT(CASE WHEN loan_status = 'Current' OR loan_status = 'Fully Paid' THEN id END) * 100.0) / COUNT(id) FROM loan_data_2018) AS good_loan_percentage,
441     COUNT(id) AS good_loan_applications,
442     FORMAT(SUM(loan_amount), 'C', 'en-US') AS good_loan_funded_amount,
443     FORMAT(SUM(total_payment), 'C', 'en-US') AS good_loan_amount_received,
444     FORMAT((SUM(loan_amount) - SUM(total_payment)) * (1 + AVG(interest_rate)), 'C', 'en-US') AS total_outstanding_balance
445 FROM loan_data_2018
446 WHERE loan_status = 'Current' OR loan_status = 'Fully Paid'

```

	good_loan_percentage	good_loan_applications	good_loan_funded_amount	good_loan_amount_received	total_outstanding_balance
1	87.444016794961	49983	\$992,866,850.00	\$785,506,845.33	\$237,135,503.36

```

449 -- the average interest rate for bad loans at 17.45% is only slightly higher than the rates for good loans averaging 14.43%
450 -- of the 50000 loans in good standing: 17525 are fully paid and 32458 are current
451 -- of the loans that are current about 63.2% of the total funded amount (not including interest) has been repaid
452 -- all loan applicants had similar average DTI levels at around 21%
453 SELECT
454     loan_status AS Loan_Status,
455     COUNT(id) AS Total_Loan_Applications,
456     FORMAT(SUM(loan_amount), 'C', 'en-US') AS Total_Funded_Amount,
457     FORMAT(SUM(total_payment), 'C', 'en-US') AS Total_Amount_Received,
458     AVG(interest_rate * 100) AS Avg_Interest_Rate,
459     AVG(debt_to_income * 100) AS Avg_Debt_to_Income
460 FROM loan_data_2018
461 GROUP BY loan_status
462 ORDER BY loan_status ASC
463
464 SELECT
465     loan_status AS Loan_Status,
466     FORMAT(SUM(total_payment), 'C', 'en-US') AS MTD_Total_Amount_Received,
467     FORMAT(SUM(loan_amount), 'C', 'en-US') AS MTD_Total_Funded_Amount
468 FROM loan_data_2018
469 WHERE MONTH(issue_date) = 12 AND YEAR(issue_date) = 2018
470 GROUP BY loan_status
471 ORDER BY loan_status ASC

```

	Loan_Status	Total_Loan_Applications	Total_Funded_Amount	Total_Amount_Received	Avg_Interest_Rate	Avg_Debt_to_Income
1	Charged Off	7177	\$145,335,150.00	\$54,515,127.14	17.4487070358383	21.2798634478354
2	Current	32458	\$668,935,375.00	\$422,602,080.49	14.2068948060269	21.0670001237622
3	Fully Paid	17525	\$323,931,475.00	\$362,904,764.84	14.6416428452637	20.6537980119699

	Loan_Status	MTD_Total_Amount_Received	MTD_Total_Funded_Amount
1	Charged Off	\$1,614,785.17	\$6,184,775.00
2	Current	\$21,944,025.52	\$45,214,425.00
3	Fully Paid	\$14,947,592.63	\$13,646,000.00

473 | -- Loans which were fully paid resulted in a profit of 12% (which is less than avg interest of 14.6%) or \$39M, whereas chargeoffs were a loss of 62.5% or \$90.8M
474 | -- hence net loss was \$51.8M among Fully Paid and Charged Off loans

```
475 | SELECT
476 |     loan_status AS Loan_Status,
477 |     (SUM(total_payment) / SUM(loan_amount)-1) AS Profit_Multiple,
478 |     FORMAT((SUM(loan_amount) * (SUM(total_payment) / SUM(loan_amount)-1)), 'C', 'en-US') AS Profit
479 | FROM loan_data_2018 WHERE loan_status = 'Fully Paid' OR loan_status = 'Charged Off'
480 | GROUP BY loan_status
```

	Loan_Status	Profit_Multiple	Profit
1	Fully Paid	0.120313377516796	\$38,973,289.84
2	Charged Off	-0.624900602927515	(\$90,820,022.86)

```
484 | -- the purpose of most loans was for debt consolidation (58.5% of all loans) and credit cards (21.5% of all loans)
485 | SELECT
486 |     purpose AS Loan_Purpose,
487 |     COUNT(id) AS Total_Loan_Applications,
488 |     COUNT(id) * 100.0 / (SELECT COUNT(id) FROM loan_data_2018) AS Total_Loan_Applications_Percentage,
489 |     FORMAT(SUM(loan_amount), 'C', 'en-US') AS Total_Funded_Amount,
490 |     FORMAT(SUM(total_payment), 'C', 'en-US') AS Total_Received_Amount
491 | FROM loan_data_2018
492 | -- WHERE loan_status = 'Charged Off' -- used to compare between charged off loans only and all loans to see if the distribution of loans in each category varies
493 | GROUP BY purpose
494 | ORDER BY Total_Loan_Applications DESC
```

	Loan_Purpose	Total_Loan_Applications	Total_Loan_Applications_Percentage	Total_Funded_Amount	Total_Received_Amount
1	Debt Consolidation	33435	58.493701889433	\$705,984,675.00	\$520,128,743.04
2	Credit Card	12309	21.534289713086	\$243,634,925.00	\$179,159,762.15
3	Other	3887	6.800209937018	\$57,036,725.00	\$43,271,613.84
4	Home Improvement	3292	5.759272218334	\$64,872,775.00	\$48,595,017.44
5	Major Purchase	1062	1.857942617214	\$20,045,100.00	\$14,592,753.09
6	Medical	686	1.200139958012	\$8,738,600.00	\$6,503,100.16
7	House	606	1.060181945416	\$11,288,125.00	\$8,363,314.25
8	Moving	576	1.007697690692	\$5,744,950.00	\$4,215,035.76
9	Small Business	550	0.962211336599	\$12,800,325.00	\$9,179,222.55
10	Car	365	0.638558432470	\$4,554,350.00	\$3,395,482.21
11	Vacation	328	0.573827851644	\$2,703,775.00	\$1,973,778.91
12	Renewable Energy	63	0.110216934919	\$782,775.00	\$636,401.81
13	Wedding	1	0.001749475157	\$14,900.00	\$7,747.28

```

496 | -- 40% of all loans issued were to those who had 10+ years employment history
497 | SELECT
498 |     employee_length AS Employee_Length,
499 |     COUNT(id) AS Total_Loan_Applications,
500 |     COUNT(id) * 100.0 / (SELECT COUNT(id) FROM loan_data_2018) AS Total_Loan_Applications_Percentage,
501 |     FORMAT(SUM(loan_amount), 'C', 'en-US') AS Total_Funded_Amount,
502 |     FORMAT(SUM(total_payment), 'C', 'en-US') AS Total_Received_Amount
503 | FROM loan_data_2018
504 | -- WHERE loan_status = 'Charged Off'
505 | GROUP BY employee_length
506 | ORDER BY Total_Loan_Applications DESC

```

	Employee_Length	Total_Loan_Applications	Total_Loan_Applications_Percentage	Total_Funded_Amount	Total_Received_Amount
1	10+ years	23090	40.395381385584	\$480,092,350.00	\$354,889,446.00
2	2 years	5136	8.985304408677	\$96,791,650.00	\$72,241,496.57
3	3 years	5025	8.791112666200	\$97,214,050.00	\$71,642,651.47
4	4 years	3885	6.796710986703	\$73,316,900.00	\$54,383,099.65
5	< 1 year	3830	6.700489853044	\$77,442,200.00	\$55,684,570.52
6	5 years	3793	6.635759272218	\$73,374,325.00	\$54,298,067.56
7	1 year	3402	5.951714485654	\$63,480,500.00	\$46,044,174.18
8	6 years	2869	5.019244226731	\$56,186,300.00	\$41,518,071.22
9	7 years	2303	4.029041287613	\$44,750,625.00	\$33,425,779.28
10	8 years	2143	3.749125262421	\$42,138,975.00	\$31,304,296.69
11	9 years	1684	2.946116165150	\$33,414,125.00	\$24,590,319.34

```

508 | -- 58% of loans issued to those who have a mortgage on their house, 32% to renters, and 10% to those who own their house outright
509 | SELECT
510 |     home_ownership AS Home_Ownership,
511 |     COUNT(id) AS Total_Loan_Applications,
512 |     COUNT(id) * 100.0 / (SELECT COUNT(id) FROM loan_data_2018) AS Total_Loan_Applications_Percentage,
513 |     FORMAT(SUM(loan_amount), 'C', 'en-US') AS Total_Funded_Amount,
514 |     FORMAT(SUM(total_payment), 'C', 'en-US') AS Total_Received_Amount
515 | FROM loan_data_2018
516 | -- WHERE loan_status = 'Charged Off'
517 | GROUP BY home_ownership
518 | ORDER BY Total_Loan_Applications DESC

```

	Home_Ownership	Total_Loan_Applications	Total_Loan_Applications_Percentage	Total_Funded_Amount	Total_Received_Amount
1	Mortgage	33169	58.028341497550	\$701,010,800.00	\$519,683,997.03
2	Rent	18274	31.969909027291	\$327,711,775.00	\$237,404,542.66
3	Own	5717	10.001749475157	\$109,479,425.00	\$82,933,432.78

```

520 -- CA is the top state with issued loans at 12.3% of all loans, followed by TX at 8.7%, and FL and NY tied for third at 6.8%
521 -- IA is the only state with no loans issued
522 SELECT
523     address_state AS State,
524     COUNT(id) AS Total_Loan_Applications,
525     COUNT(id) * 100.0 / (SELECT COUNT(id) FROM loan_data_2018) AS Total_Loan_Applications_Percentage,
526     FORMAT(SUM(loan_amount), 'C', 'en-US') AS Total_Funded_Amount,
527     FORMAT(SUM(total_payment), 'C', 'en-US') AS Total_Received_Amount
528 -- (SUM(loan_amount)-SUM(total_payment))/SUM(loan_amount) AS funding_gap_percent
529 FROM loan_data_2018
530 -- WHERE loan_status = 'Charged Off'
531 GROUP BY address_state
532 ORDER BY Total_Loan_Applications DESC

```

	State	Total_Loan_Applications	Total_Loan_Applications_Percentage	Total_Funded_Amount	Total_Received_Amount
1	CA	7019	12.279566130160	\$144,476,925.00	\$108,563,639.50
2	TX	4952	8.663400979706	\$101,243,725.00	\$74,891,556.69
3	FL	3888	6.801959412176	\$74,955,000.00	\$54,571,346.45
4	NY	3855	6.744226731980	\$78,296,475.00	\$56,146,690.01
5	IL	2340	4.093771868439	\$47,692,500.00	\$35,019,564.38
6	GA	1971	3.448215535339	\$40,388,550.00	\$29,330,374.74
7	NJ	1914	3.348495451364	\$40,430,125.00	\$29,565,126.81
8	PA	1880	3.289013296011	\$37,095,650.00	\$27,351,635.75
9	OH	1867	3.266270118964	\$34,986,050.00	\$25,638,098.68
10	VA	1643	2.874387683694	\$34,449,250.00	\$25,448,940.21
11	MD	1596	2.792162351294	\$33,735,000.00	\$24,555,303.83
12	NC	1565	2.737928621413	\$30,631,725.00	\$22,310,330.67
13	MI	1513	2.646955913226	\$28,954,675.00	\$21,700,257.83
14	CO	1325	2.318054583624	\$26,519,700.00	\$20,679,953.29
15	AZ	1307	2.286564030790	\$24,787,075.00	\$18,711,776.14
16	WA	1302	2.277816655003	\$25,682,000.00	\$20,660,972.06
17	MA	1250	2.186843946815	\$25,649,025.00	\$19,020,403.61
18	MN	1067	1.866689993002	\$20,988,375.00	\$15,058,203.79
19	IN	1061	1.856193142057	\$20,235,750.00	\$14,975,923.76
20	TN	1039	1.817704688593	\$19,823,200.00	\$14,535,993.42
21	CT	1002	1.752974107767	\$20,218,175.00	\$14,474,533.16
22	MO	994	1.738978306508	\$18,902,050.00	\$13,687,993.75
23	WI	897	1.569279216235	\$16,740,875.00	\$12,487,145.20
24	NV	843	1.474807557732	\$16,392,300.00	\$12,707,247.53
25	SC	772	1.350594821553	\$15,247,475.00	\$11,409,347.64

26	AL	685	1.198390482855	\$13,081,950.00	\$9,229,371.83
27	LA	683	1.194891532540	\$13,388,925.00	\$9,291,666.26
28	OR	677	1.184394681595	\$12,877,600.00	\$9,979,152.08
29	KY	572	1.000699790062	\$10,124,050.00	\$7,249,637.86
30	OK	558	0.976207137858	\$11,302,425.00	\$7,939,040.52
31	KS	491	0.858992302309	\$9,582,250.00	\$6,769,880.18
32	AR	436	0.762771168649	\$8,294,200.00	\$5,805,124.18
33	UT	421	0.736529041287	\$8,169,750.00	\$6,567,368.40
34	MS	383	0.670048985304	\$7,525,050.00	\$5,280,051.31
35	NE	364	0.636808957312	\$6,379,200.00	\$4,731,400.58
36	WV	354	0.619314205738	\$6,966,250.00	\$5,214,387.47
37	NH	314	0.549335199440	\$6,543,350.00	\$4,949,197.73
38	NM	284	0.496850944716	\$5,402,100.00	\$3,879,498.56
39	HI	268	0.468859342197	\$5,563,800.00	\$3,996,057.31
40	RI	250	0.437368789363	\$4,773,250.00	\$3,406,322.21
41	ID	224	0.391882435269	\$4,073,150.00	\$3,229,931.03
42	ME	220	0.384884534639	\$4,063,450.00	\$3,111,372.98
43	DE	182	0.318404478656	\$3,583,525.00	\$2,517,650.85
44	WY	156	0.272918124562	\$2,936,600.00	\$2,272,245.45
45	AK	154	0.269419174247	\$3,361,600.00	\$2,485,352.51
46	MT	147	0.257172848145	\$2,649,375.00	\$1,969,317.32
47	VT	146	0.255423372988	\$2,458,775.00	\$1,788,511.05
48	ND	134	0.234429671098	\$2,752,450.00	\$2,083,537.06
49	SD	110	0.192442267319	\$1,918,900.00	\$1,384,207.37
50	DC	85	0.148705388383	\$1,908,375.00	\$1,389,331.45

```

534 -- May is the month with most loans issued at 10% of all loans, while December is the month with the least at 5.7%
535 -- Loan applications trended up from Jan. through May, then trended down from May through the rest of the year
536 SELECT
537     MONTH(issue_date) AS Mnth,
538     DATENAME(MONTH, issue_date) AS 'Month',
539     COUNT(id) AS Total_Loan_Applications,
540     COUNT(id) * 100.0 / (SELECT COUNT(id) FROM loan_data_2018) AS Total_Loan_Applications_Percentage,
541     FORMAT(SUM(loan_amount), 'C', 'en-US') AS Total_Funded_Amount,
542     FORMAT(SUM(total_payment), 'C', 'en-US') AS Total_Received_Amount
543 FROM loan_data_2018
544 -- WHERE loan_status = 'Charged Off'
545 GROUP BY MONTH(issue_date), DATENAME(MONTH, issue_date)
546 ORDER BY Mnth

```

	Mnth	Month	Total_Loan_Applications	Total_Loan_Applications_Percentage	Total_Funded_Amount	Total_Received_Amount
1	1	January	4565	7.986354093771	\$92,398,725.00	\$79,264,216.92
2	2	February	4326	7.568229531140	\$86,815,275.00	\$72,694,970.79
3	3	March	5333	9.329951014695	\$106,329,750.00	\$86,589,651.71
4	4	April	5492	9.608117564730	\$108,591,650.00	\$85,519,103.56
5	5	May	5699	9.970258922323	\$111,167,375.00	\$84,516,780.22
6	6	June	4865	8.511196641007	\$94,388,200.00	\$71,063,507.64
7	7	July	5307	9.284464660601	\$104,000,650.00	\$76,874,141.60
8	8	August	5470	9.569629111266	\$107,553,425.00	\$76,170,421.09
9	9	September	4614	8.072078376487	\$92,962,300.00	\$62,523,015.26
10	10	October	4630	8.100069979006	\$94,852,100.00	\$60,764,535.13
11	11	November	3606	6.308607417774	\$74,097,350.00	\$45,535,225.22
12	12	December	3253	5.691042687193	\$65,045,200.00	\$38,506,403.32

```

548 -- 58% of loans issued for a 36 month term while 42% issued for a 60 month term
549 SELECT
550     term AS Term,
551     COUNT(id) AS Total_Loan_Applications,
552     COUNT(id) * 100.0 / (SELECT COUNT(id) FROM loan_data_2018) AS Total_Loan_Applications_Percentage,
553     FORMAT(SUM(loan_amount), 'C', 'en-US') AS Total_Funded_Amount,
554     FORMAT(SUM(total_payment), 'C', 'en-US') AS Total_Received_Amount
555 FROM loan_data_2018
556 -- WHERE loan_status = 'Charged Off'
557 GROUP BY Term
558 ORDER BY Term

```

	Term	Total_Loan_Applications	Total_Loan_Applications_Percentage	Total_Funded_Amount	Total_Received_Amount
1	36 months	33139	57.975857242827	\$564,465,475.00	\$466,477,998.39
2	60 months	24021	42.024142757172	\$573,736,525.00	\$373,543,974.08

```

560  /*
561  -- 62% of all loans issued were to C - G graded applicants which represent moderate through very high risk loans (and therefore carry higher average interest rates)
562     since all applicants in the dataset have manageable DTI levels, these C-G grades have average through very low credit scores,
563     may have a history of missed payments, and potentially carry significant risk factors
564  -- 38% of all loans issued were to A and B graded applicants having high to good credit scores with overall positive credit history representing low risk loans
565  -- While the average interest rate only jumps by about 4% from A to B grades, the interest rate more than doubles from A to C grades, and more than doubles from C to G grades
566  -- Only 2.3% of all A and B graded loans issued were charged off (about 18% of all bad loans),
567     meaning 82% of all charge offs came from C - G graded loans, hence there is a strong negative correlation between grade and bad loans issued
568  */
569  SELECT
570      grade AS Grade,
571      COUNT(id) AS Total_Loan_Applications,
572      COUNT(id) * 100.0 / (SELECT COUNT(id) FROM loan_data_2018) AS Total_Loan_Applications_Percentage,
573      AVG(interest_rate) AS Average_Interest_Rate,
574      AVG(debt_to_income) AS Average_DTI
575  FROM loan_data_2018
576  -- WHERE loan_status = 'Charged Off'
577  GROUP BY grade
578  ORDER BY grade ASC

```

	Grade	Total_Loan_Applications	Total_Loan_Applications_Percentage	Average_Interest_Rate	Average_DTI
1	A	9015	15.771518544436	0.0714684080747816	0.19682044370665
2	B	12636	22.106368089573	0.109755286315103	0.205070995602753
3	C	17611	30.810006997900	0.14739977402695	0.211282925462931
4	D	13245	23.171798460461	0.195359275817714	0.216341978155721
5	E	3792	6.634009797060	0.253063528786732	0.220699235227132
6	F	717	1.254373687893	0.294253139607245	0.225109763090165
7	G	144	0.251924422673	0.308143063965771	0.239577082296213

```

569  SELECT
570      grade AS Grade,
571      COUNT(id) AS Total_Loan_Applications,
572      COUNT(id) * 100.0 / (SELECT COUNT(id) FROM loan_data_2018) AS Total_Loan_Applications_Percentage,
573      AVG(interest_rate) AS Average_Interest_Rate,
574      AVG(debt_to_income) AS Average_DTI
575  FROM loan_data_2018
576  WHERE loan_status = 'Charged Off'
577  GROUP BY grade
578  ORDER BY grade ASC

```

	Grade	Total_Loan_Applications	Total_Loan_Applications_Percentage	Average_Interest_Rate	Average_DTI
1	A	332	0.580825752274	0.0734542168744178	0.194529819470572
2	B	1007	1.761721483554	0.110860774472963	0.204482919380232
3	C	2274	3.978306508047	0.148142261301203	0.211964819567211
4	D	2412	4.219734079776	0.197066459675037	0.215559079665474
5	E	900	1.574527641707	0.254597000430028	0.220034888941381
6	F	204	0.356892932120	0.294812746784266	0.222345588144426
7	G	48	0.083974807557	0.308150008320808	0.238149997778237

```

581  /*
582  Overall Findings:
583  - Even in the best case scenario where the roughly 237.1M outstanding loan balance had been paid off for current loans by the end of 2018,
584    2018 would've still seen a loss of  $(1 - ((\text{total\_amount\_received} + \text{total\_outstanding\_balance}) / \text{total\_funded\_amount})) * 100 = 3.86\%$ 
585    indicating that far too many loans were charged off at 12.6% of all loans issued
586  - The distribution of all loans issued by fields like purpose, employee length, home ownership, apps by state, apps by month, and apps by term
587    remain consistent regardless of the loan status, hence there is no significant correlation between loan status and these other fields
588  - As the loan grade increases (A being the highest) the likelihood of a charge off significantly decreases (especially after increasing from C to B),
589    hence there is a strong negative correlation between grade and bad loans issued
590  - The first half of the year is the most opportune for capturing as much business as possible
591  */
592
593  SELECT
594    (1 - ((SUM(total_payment) + ((SUM(loan_amount) - SUM(total_payment)) * (1 + AVG(interest_rate))))
595    / SUM(loan_amount))) * 100 AS best_case_scenario_net_profit_percent
596  FROM loan_data_2018

```

	best_case_scenario_net_profit_percent
1	-3.86339933778701

```

598 -- contingency table (cross-tabulation) of the categorical loan_status and grade variables to analyze correlation
599 -- 4th column added to view Bad_Good Ratio of loans per grade for easy interpretation
600 -- as the Bad_Good_Ratio increases the loan grade decreases (G being the lowest)
601 -- overall risk profile of the corresponding loans increases
602 SELECT
603     COALESCE(bad_loans.grade, good_loans.grade) AS Grade,
604     COALESCE(bad_loans.Total_Bad_Loans, 0) AS Total_Bad_Loans,
605     COALESCE(good_loans.Total_Good_Loans, 0) AS Total_Good_Loans,
606     CASE
607         WHEN COALESCE(good_loans.Total_Good_Loans, 0) = 0 THEN NULL
608         ELSE CAST(COALESCE(bad_loans.Total_Bad_Loans, 0) AS FLOAT) /
609             CAST(COALESCE(good_loans.Total_Good_Loans, 1) AS FLOAT)
610     END AS Bad_Good_Ratio
611 FROM
612     (SELECT
613         grade AS grade,
614         COUNT(id) AS Total_Bad_Loans
615     FROM loan_data_2018
616     WHERE loan_status = 'Charged Off'
617     GROUP BY grade) AS bad_loans
618 FULL OUTER JOIN
619     (SELECT
620         grade AS grade,
621         COUNT(id) AS Total_Good_Loans
622     FROM loan_data_2018
623     WHERE loan_status = 'Current' OR loan_status = 'Fully Paid'
624     GROUP BY grade) AS good_loans
625 ON
626     bad_loans.grade = good_loans.grade
627 ORDER BY
628     Grade ASC

```

	Grade	Total_Bad_Loans	Total_Good_Loans	Bad_Good_Ratio
1	A	332	8683	0.0382356328457906
2	B	1007	11629	0.0865938601771434
3	C	2274	15337	0.148268892221425
4	D	2412	10833	0.222653004707837
5	E	900	2892	0.311203319502075
6	F	204	513	0.39766081871345
7	G	48	96	0.5

LendingClub End-to-End Project | Part 3 - Tableau Dashboard

<https://public.tableau.com/app/profile/alexander.j.porter/viz/2018LendingClubDashboard/Summary>

LendingClub End-to-End Project | Part 4 - Presentation

[https://github.com/Alexander-J-Porter/LendingClub-Loan-End-to-End-Project-using-Excel-SQL-Tableau/blob/main/LendingClub Loan Presentation.pdf](https://github.com/Alexander-J-Porter/LendingClub-Loan-End-to-End-Project-using-Excel-SQL-Tableau/blob/main/LendingClub%20Loan%20Presentation.pdf)