

Machine Learning 441 Assignment 1

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Task 1: Analytics Base Table

1. 581012
2. 62
3. Continuous features

Table 1: Data Quality Report for Continuous Features

Feature	Count	% Miss.	Card.	Min.	1 st Qrt.	Mean	Median	3 rd Qrt.	Max.	Std. Dev.
1	581012	0.00	1978	2054845.65	3104928.15	3271134.43	3311628.60	3496222.05	4264440.30	309481.13
2	581012	0.51	361	0.00	58.00	155.66	127.00	260.00	360.00	111.91
3	581012	0.00	576099	0.00	145.49	389.92	318.12	652.53	903.48	280.34
4	581012	0.05	67	0.00	9.00	14.10	13.00	18.00	66.00	7.49
5	581012	0.51	569	-691.00	108.00	269.42	218.00	384.00	1397.00	212.56
6	581012	0.00	581012	-173.07	6.99	46.42	29.91	68.97	600.95	58.30
7	581012	0.51	577988	-1.00	-0.50	-0.00	-0.00	0.50	1.00	0.58
8	581012	0.00	5811	0.00	1106.00	8158.11	1997.00	3328.00	510165098.00	1185156.02
9	581012	0.51	207	0.00	198.00	212.14	218.00	231.00	254.00	26.77
11	581012	0.00	255	0.00	119.00	142.53	143.00	168.00	254.00	38.27
12	581012	0.51	5826	0.00	1024.00	1980.43	1710.00	2550.00	7173.00	1324.25
61	581012	0.00	581012	1.00	145253.75	290506.50	290506.50	435759.25	581012.00	167723.86

4. Target feature data quality report:

Table 3: Data Quality Report for target feature

Feature	Count	%miss	Card.	Mode	Mode Freq.	Mode %	2 nd Mode	2 nd Mode Freq	2 nd Mode %
T	581012	0.00	7	2.00	283269	48.75	1.00	211815	36.46

Task 2: Data Quality Issues

Feature	Data Quality Issue	Justification
A2	Missing Values	Many records have missing values for A2 as observed in Table 1. There are 0.51% of the observations with A2 missing.
A2, A3	Perfect Correlation	The features A2 and A3 are perfectly correlated as observed in Figure 5. This means that they are an <i>exact</i> linear transformation of each other.
A4, A5, A7, A9, A10, A12, A13, A16, A18, A20, A21	Missing Values	Testing revealed that there is a certain 2947 rows with all these features missing simultaneously.
A8	Extreme outliers / skew	The maximum value of this feature is much higher than the mean and the 3rd quartile, indicating that there may be an error. The extreme nature of these outliers is shown in Figure 1. Through testing it was found that there are 18 observations greater than 50,000,000 with the 99.9th percentile sitting at 6699.0.
A9, A11	Strongly Correlated Features	Figure 5 indicates that A9 and A11 are strongly correlated with a coefficient of -0.78. After further investigation, we observed that there is likely a mathematical or physical constraint between the two features due to the relationship exhibited in Figure 6.
A10	Abnormal Data Type	There are 0.99% of the observations in feature A10 that have the value 'a' whilst all the other observations have numerical values.
A16	Cardinality of 1	This categorical variable is redundant since all the observations have the same value.
A17	Cardinality of 1	This categorical variable is redundant since all the observations have the same value.
A21	Missing Values	Many records, specifically 70.16% have missing values for A21 as observed in Table 2.
A35	Class imbalance	The 2nd class for this feature only has 3 observations. This rounds down to 0.00% shown in Table 2.
A61	ID Feature	This feature has a perfectly uniform distribution observed in Figure 2. Testing revealed that the feature value is exactly the same row number for all observations.
A61, A1	Strongly Correlated Features	The features A1 and A61 are strongly correlated, as observed in Figure 5 with a correlation coefficient of -0.95.
A62	Cardinality of 1	This categorical variable is redundant since all the observations have the same value.

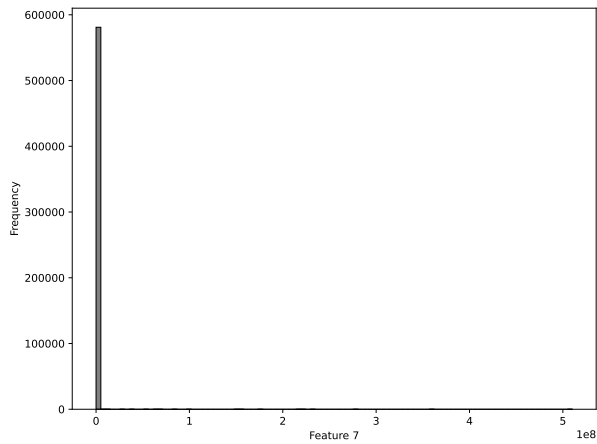


Figure 1: Feature 8 distribution with 100 bins

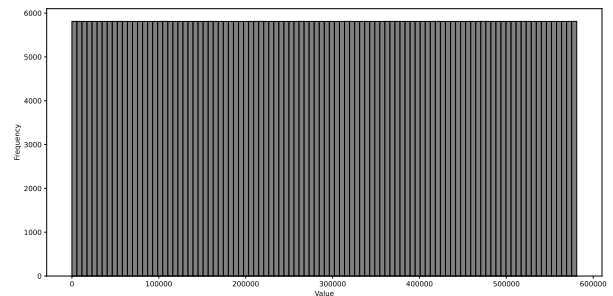


Figure 2: Feature 61 distribution with 100 bins

Task 3: Addressing The Data Quality Issues

Table 4: Data Quality Fixes

Feature	Data Quality Issue	Handling Strategy	Justification
A2	Missing Values	Remove Feature	See below.
A2, A3	Perfect Correlation	Remove Feature A2	Since the features are perfectly correlated, we are essentially storing the same information twice which provides no value to a machine learning model. Since A2 has missing values we might as well remove it to deal with two quality issues at the same time.
A4, A5, A7, A9, A10, A12, A13, A16, A18, A20, A21	Missing Values	Remove Observations	Since testing revealed that there is a certain 2947 rows with all these features missing simultaneously, it makes sense to simply remove all these observations since they have a large amount of missing information. It is useful to note that the distribution of the target variable in these observations matches the overall distribution of the target variable for the whole dataset observed in Figure 7. This means that we are likely not losing useful information about a certain feature when removing these observations.

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Table 4 – continued from previous page

Feature	Data Quality Issue	Handling Strategy	Justification
A8	Outlier observations	Observation removal	There must be an error with the observations in question because the outliers are so incredibly skewed. Through testing it was found that there are 18 observations greater than 50,000,000 with the 99.9th percentile sitting at 6699.0.
A9, A11	Strongly Correlated Features	Nothing	Despite these features being strongly correlated, there is still a chance that the features' relationship represents makes sense in the domain context. Therefore we elect to not remove either feature.
A10	Abnormal Data Type	Imputation	Testing revealed that the levels of this suspected categorical variable range from 100–254 predominantly. Since many observations have the value 'a', we can guess this error is not at random and the character 'a' should be replaced by 255. This fits the distribution seen in Figure 4.
A16	Cardinality of 1	Drop Feature	This categorical variable can be removed since it will have no effect on a model's performance.
A17	Cardinality of 1	Drop Feature	This categorical variable can be removed since it will have no effect on a model's performance.
A21	Missing Values	Drop Feature	Testing revealed that the correlation between missingness of A21 and the target is -0.0004. Due to the high number of missing values we should remove the feature.
A35	Class imbalance	Drop Feature	Since there are only three observations in the second class for this feature, it should be removed without affecting model performance.
A61	ID Feature	Drop Feature	It was confirmed this feature perfectly fits the description of an ID column, meaning it has no predictive power and can be removed.
A61, A1	Strongly Correlated Features	Drop Feature	Since we have already established that A61 is a redundant ID feature and have determined to remove it, we have solved the issue.
A62	Cardinality of 1	Drop Feature	This categorical variable can be removed since it will have no effect on a model's performance.

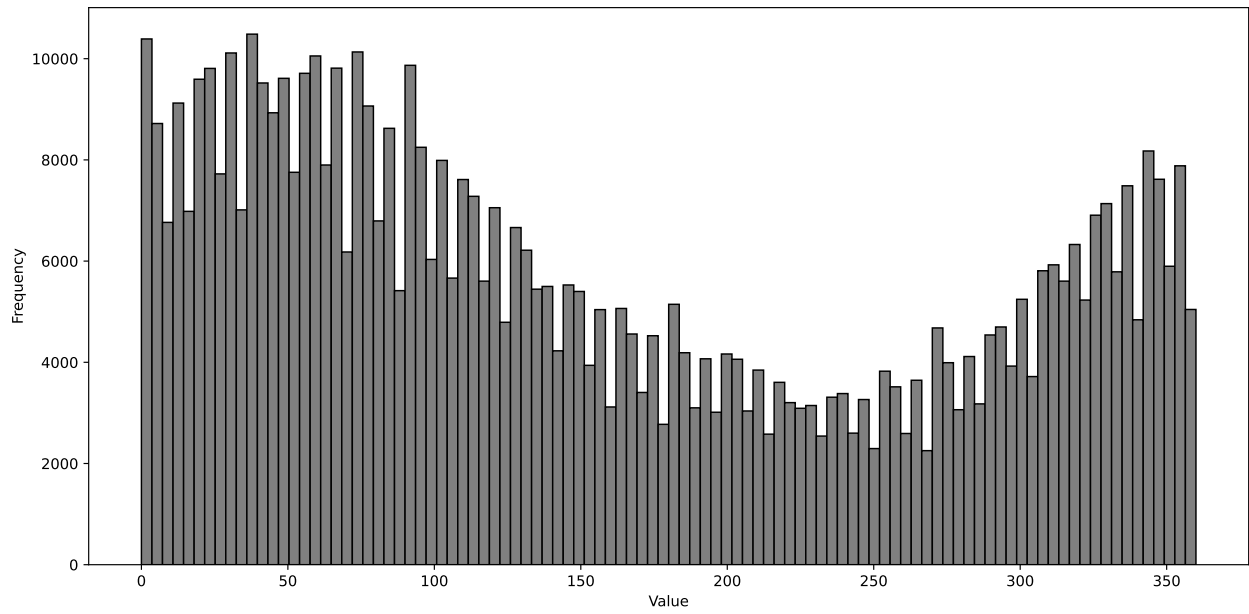


Figure 3: Feature 2 distribution with 100 bins

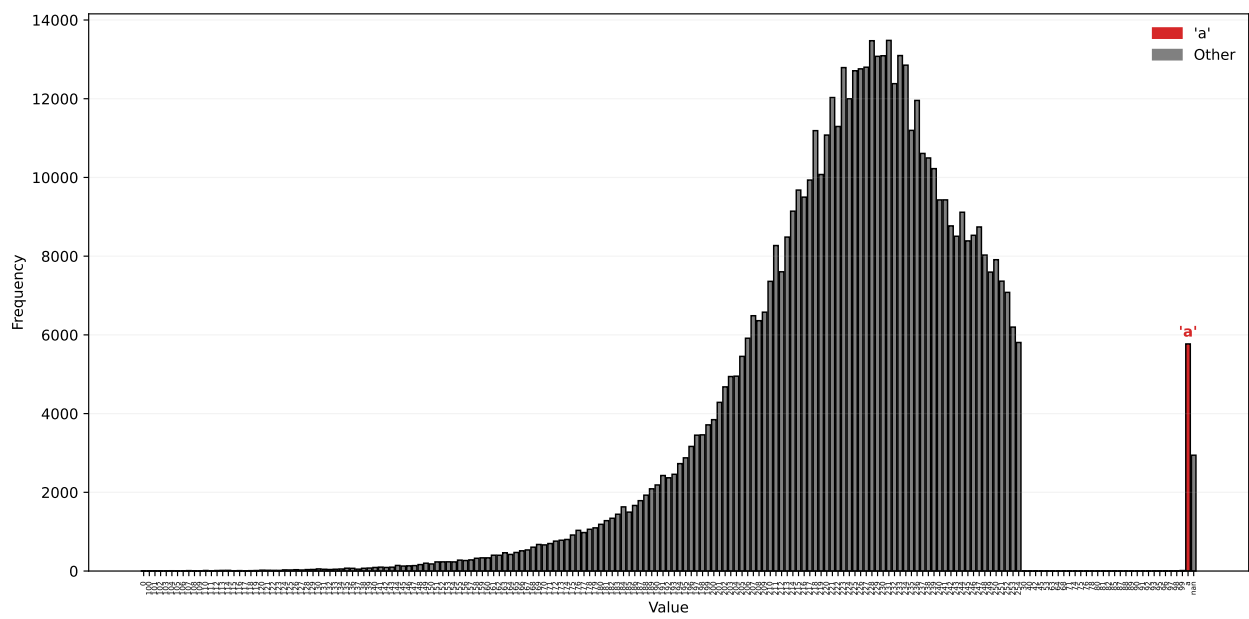


Figure 4: Feature 10 distribution ordered by value

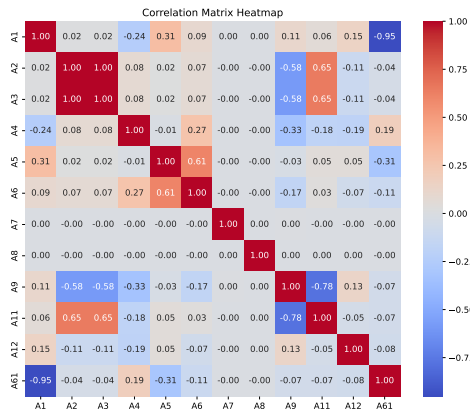


Figure 5: Correlation heatmap

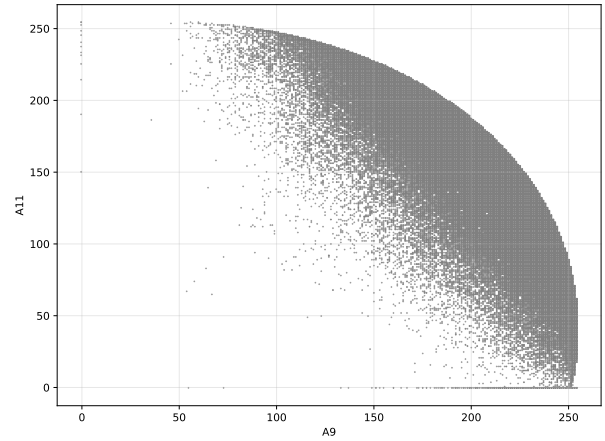


Figure 6: A9 vs A11 scatter plot

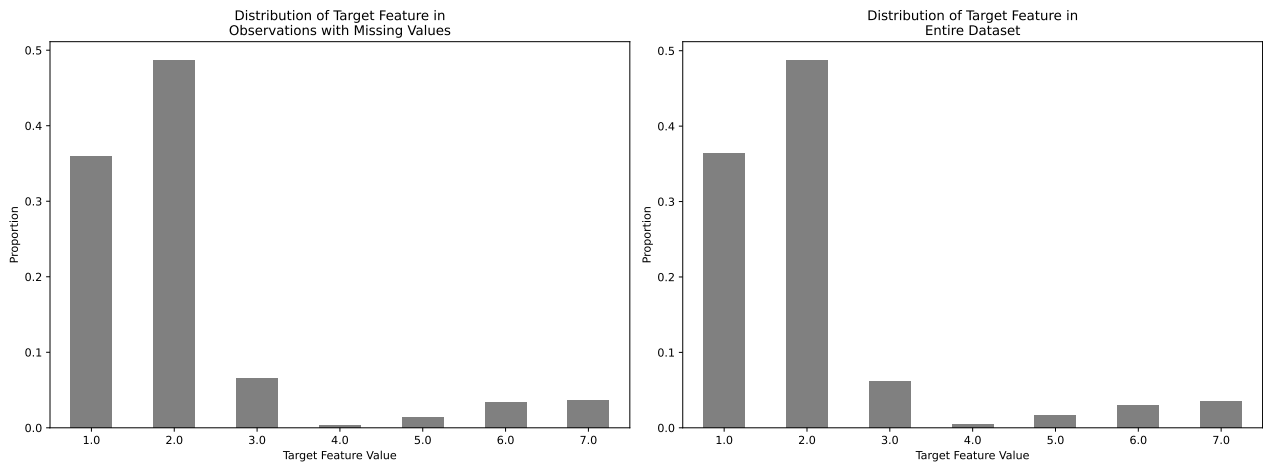


Figure 7: Target value distribution comparison

Table 2: Categorical Data Quality Report

Feature	Count	%miss	Card.	Mode	Mode Freq.	Mode %	2 nd Mode	2 nd Mode Freq	2 nd Mode %
10	578069	0.51	186	231	13482	2.33	228	13474	2.33
13	578069	0.51	2	0.00	318579	55.11	1.00	259490	44.89
14	581012	0.00	2	0	551128	94.86	1	29884	5.14
15	581012	0.00	2	0	327648	56.39	1	253364	43.61
16	578069	0.51	1	0.00	578069	100.00		0	0.00
17	581012	0.00	1	0	581012	100.00		0	0.00
18	578069	0.51	3	0.00	538608	93.17	1.00	36589	6.33
19	581012	0.00	2	0	291278	50.13	1	289734	49.87
20	577566	0.59	2	0.00	574553	99.48	1.00	3013	0.52
21	173355	70.16	2	0.00	172455	99.48	1.00	900	0.52
22	581012	0.00	2	0	573487	98.70	1	7525	1.30
23	581012	0.00	2	0	576189	99.17	1	4823	0.83
24	581012	0.00	2	0	568616	97.87	1	12396	2.13
25	581012	0.00	2	0	579415	99.73	1	1597	0.27
26	581012	0.00	2	0	574437	98.87	1	6575	1.13
27	581012	0.00	2	0	580907	99.98	1	105	0.02
28	581012	0.00	2	0	580833	99.97	1	179	0.03
29	581012	0.00	2	0	579865	99.80	1	1147	0.20
30	581012	0.00	2	0	548378	94.38	1	32634	5.62
31	581012	0.00	2	0	568602	97.86	1	12410	2.14
32	581012	0.00	2	0	551041	94.84	1	29971	5.16
33	581012	0.00	2	0	563581	97.00	1	17431	3.00
34	581012	0.00	2	0	580413	99.90	1	599	0.10
35	581012	0.00	2	0	581009	100.00	1	3	0.00
36	581012	0.00	2	0	578167	99.51	1	2845	0.49
37	581012	0.00	2	0	577590	99.41	1	3422	0.59
38	581012	0.00	2	0	579113	99.67	1	1899	0.33
39	581012	0.00	2	0	576991	99.31	1	4021	0.69
40	581012	0.00	2	0	571753	98.41	1	9259	1.59
41	581012	0.00	2	0	580174	99.86	1	838	0.14
42	581012	0.00	2	0	547639	94.26	1	33373	5.74
43	581012	0.00	2	0	523260	90.06	1	57752	9.94
44	581012	0.00	2	0	559734	96.34	1	21278	3.66
45	581012	0.00	2	0	580538	99.92	1	474	0.08
46	581012	0.00	2	0	578423	99.55	1	2589	0.45
47	581012	0.00	2	0	579926	99.81	1	1086	0.19
48	581012	0.00	2	0	580066	99.84	1	946	0.16
49	581012	0.00	2	0	465765	80.16	1	115247	19.84
50	581012	0.00	2	0	550842	94.81	1	30170	5.19
51	581012	0.00	2	0	555346	95.58	1	25666	4.42
52	581012	0.00	2	0	528493	90.96	1	52519	9.04
53	581012	0.00	2	0	535858	92.23	1	45154	7.77
54	581012	0.00	2	0	579401	99.72	1	1611	0.28
55	581012	0.00	2	0	579121	99.67	1	1891	0.33
56	581012	0.00	2	0	580893	99.98	1	119	0.02
57	581012	0.00	2	0	580714	99.95	1	298	0.05
58	581012	0.00	2	0	565439	97.32	1	15573	2.68
59	581012	0.00	2	0	567206	97.62	1	13806	2.38
60	581012	0.00	2	0	572262	98.49	1	8750	1.51
62	581012	0.00	1	1	581012	100.00		0	0.00