CSCI4150U: Data Mining Lab 04

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Part I:

1. Preprocessing (German Credit Data)

This dataset contains a mixture of ordinal, nominal and numeric features. The ordinal and nominal features must be processed such that Minkowski distance provides a meaningful metric during k nearest neighbors classifications while retaining as much information as possible. We begin with the following features which have either a completely arbitrary ordering or contain only 2 unique values:

- attribute 4: Purpose
 - a list of purchases on credit
- attribute 9: Personal status and sex
 - an arbitrarily ordered list of marital status and sex
- attribute 19: Telephone
 - either have a Telephone (yes) or do not (no)
- attribute 20: Foreign worker
 - either is a foreign worker (yes) or is not (no)

We can use **one-hot encoding** for Attributes 4 and 9 as these features have multiple unique values and **label encoding** for attributes 19 and 20 as these features have only two unique values which makes ordering of label encoding irrelevant.

Figure 1: [Attributes 4 and 9 one-hot encoding]

Figure 2: [Attributes 19 and 20 label encoding]

The next set of features appear to have a clear ordinal ranking with least credit-worthy on the left to most credit-worthy on the right:

- Attribute 1: Status of existing checking account
 - A11 (< 0 DM) \to A12 (0 <= & < 200 DM) \to A13 (>= 200 DM / salary assignments for at least 1 year) \to A14 (no checking account)
- Attribute 3: Credit history
 - A34 (critical account / other credits existing) \rightarrow A33 (delay in paying off in the past) \rightarrow A32 (existing credits paid back duly till now) \rightarrow A31 (all credits at this bank paid back duly) \rightarrow A30 (no credits taken / all credits paid back duly)
- Attribute 6: Savings account / bonds
 - A61 (< 100 DM) → A62 (100 <= & < 500 DM) → A63 (500 <= & < 1000 DM) → A64 (>= 1000 DM) → A65 (unknown / no savings account)
- $\bullet\,$ Attribute 7: Present employment since
 - A71 (unemployed) \rightarrow A72 (< 1 year) \rightarrow A73 (1 <= & < 4 years) \rightarrow A74 (4 <= & < 7 years) \rightarrow A75 (>= 7 years)
- Attribute 12: Property
 - A124 (unknown / no property) \rightarrow A123 (car or other, not in attribute 6) \rightarrow A122 (building society savings agreement / life insurance) \rightarrow A121 (real estate)
- Attribute 17: Job
 - A171 (unemployed / unskilled non-resident) \rightarrow A172 (unskilled resident) \rightarrow A173 (skilled employee / official) \rightarrow A174 (management / self-employed / highly qualified employee / officer)

We can visualize the value distributions of each feature for each class: