

# genre\_model\_upper\_bound\_to\_accuracy

June 1, 2020

## 1 Calculate the upper bound for accuracy of any model trained on our training data.

The data is of the form  $(X, y)$  with  $X_i \in \{0, 1\}^{\times p}$  ( $p = 1494$ ), and  $y \in \{0, 1\}$ . There are 12376 training samples. Let  $\{\bar{X}_a\}_{1 \leq a \leq 6230}$  be unique representatives of the inputs in the training set; That is, for all  $i$  there exists  $a$  such that  $X_i = \bar{X}_a$ . For each  $\bar{X}_a$  the number of female artists ( $\text{fem}(\bar{X}_a)$ ) and male artists ( $\text{mal}(\bar{X}_a)$ ) with  $X_i = \bar{X}_a$  are calculated. Define a classifier on the set of training data  $f_0 : \{X_i\}_{i=1}^{12376} \rightarrow \{0, 1\}$  as

$$f(X_i) = \operatorname{argmax}_{\{\text{male}, \text{female}\}} \{\text{mal}(\bar{X}_a), \text{fem}(\bar{X}_a)\} \text{ if } X_i = \bar{X}_a$$

Then extend  $f_0$  to  $f : \{0, 1\}^{\times p} \rightarrow \{0, 1\}$ . When  $f$  is only used on the training data, the extension from  $f_0$  to  $f$  is irrelevant, and  $f_0$  gives rise to an optimal classifier. However, to generalize to data which includes points in  $\{0, 1\}^{\times p}$  that were not in the training set, a rule is needed to make the extension.

This notebook shows that even on the training data  $f_0$  has an expected error of 26.8%, or an accuracy of 73.2%.

Questions: - for the DNN classifier the 1-fold CV accuracy has a mean of 76% with std 1%. How?  
- for the DNN classifier the training accuracy can be close to 80%. How? Is it memorizing the order and particular

```
[4]: import numpy as np
import pandas as pd

import matplotlib.pyplot as plt
import seaborn as sns; sns.set()

import re
```

Import the cleaned data:

```
[5]: %ls -lt ../../data/genre_lists/data_ready_for_model/
```

```
[6]: %store -r now
now
#now = '2020-05-11-14-35'
```