
ML Pipeline for Multiclass MNIST Classification

Programmatic Optimization, Hyperparameter Tuning, and Advanced Ensemble Architecture

Raghav Iyengar

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Abstract

This report is about implementation of a machine learning pipeline using NumPy and SciPy, aiming to do Multiclass classification of MNIST dataset. XGBoostClassifier was tried as an option but it was observed that it did not give enough accuracy for the amount of time taken by the model. Simpler models like KNN and Softmax were giving good accuracy without less computation. The final solution is a stacked model `KFoldStackingEnsemble` that tries to do the classification faster by dimensionality reduction through PCA and LDA. This architecture, built using (`NumPyKNNClassifier` and `SoftmaxClassifier`), achieved a good validation accuracy showing that combining simple models and techniques can give good accuracy.

Contents

1	Summary of Models Used & System Architecture	2
1.1	Models Implemented	2
1.2	Pre-Processing Pipeline	2
1.3	Final System Architecture: K-Fold Stacking Ensemble	2
2	Summary of Hyper-parameter Tuning	3
2.1	NumPyKNNClassifier Tuning	3
2.2	SoftmaxClassifier Tuning	5
2.3	Analysis of Model Performance on LDA Features	6
3	Detailed Summary of Thoughts and Observations	6
3.1	Algorithm NumPyKNNClassifier:- Inherently low bias,high variance	6
3.2	Algorithm SoftmaxClassifier:- Inherently high bias,low variance	7
3.3	Algorithms for Dimensionality Reduction	7
3.4	Final Observation: Optimal Ensemble	7

1 Summary of Models Used & System Architecture

1.1 Models Implemented

Three models were built using Numpy and Scipy:

1. **SoftmaxClassifier (Linear Model):** Model made for multiclass logistic regression where the softmax function is used for probability. Optimized using stochastic mini-batch gradient descent and numerical stability accounted for. This model was used because of its good accuracy.
2. **NumPyKNNClassifier (Distance-Based Model):** A k-Nearest Neighbors classifier. Numpy vector operations were used to make the distance matrix computation faster.
3. **XGBoostClassifier (Tree Ensemble):** A gradient boosting implementation which was extended from the previous assignment. This did not give any sufficient accuracy increase for the amount of training time so it was excluded.

1.2 Pre-Processing Pipeline

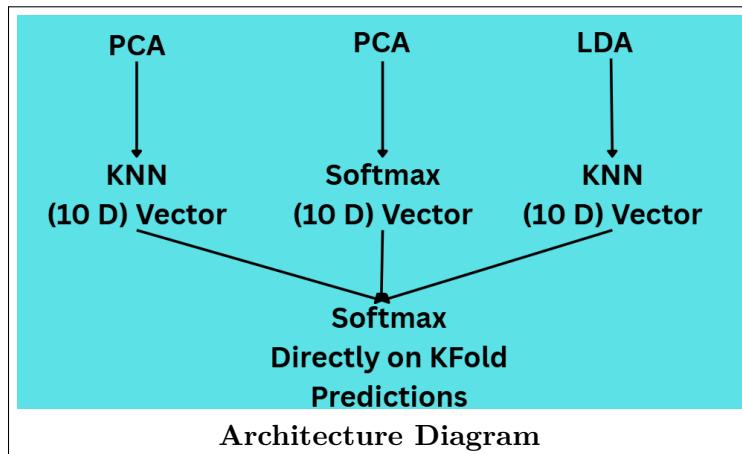
To make the models better and lesser time taking, the following pre processing techniques were used:

1. **NumPyScaler:** This works similar to the StandardScaler making each column standard variance 1 and mean centered.
2. **PCA:** To find the dimensions capturing most of the variance, PCA was performed on the Scaled dataset and different number of directions were used based on the required optimizations.
3. **LDA:** LDA was performed on the data to find the vector directions which maximize the class separability. This reduces the dimensions to 9.

This was done to prevent the complexity of high dimensions while using a model like KNN because the predict time will be very high.

1.3 Final System Architecture: K-Fold Stacking Ensemble

Instead of using parameterized weighted averages, cross validation predictions were used as robust predictions and then ensembled.



- **Level 0 (Base Learners):** Three models were used to generate cross-validated features for the final model.
 - `NumPyKNNClassifier` (on 50 PCA features)
 - `SoftmaxClassifier` (on 175 PCA features)
 - `NumPyKNNClassifier` (on 9 LDA features)
- **Meta-Feature Generation:** 3 Fold cross-validation was used to generate the features. 2/3rd of the dataset was used to makee predictions about one third of the dataset using all the three models. This was done to prevent overfitting on the training dataset.
- **Level 1 (Meta-Model):** A final `SoftmaxClassifier` was trained on the model to generate the predicted class

2 Summary of Hyper-parameter Tuning

2.1 NumPyKNNClassifier Tuning

Models with different number of components and K were run giving the results below

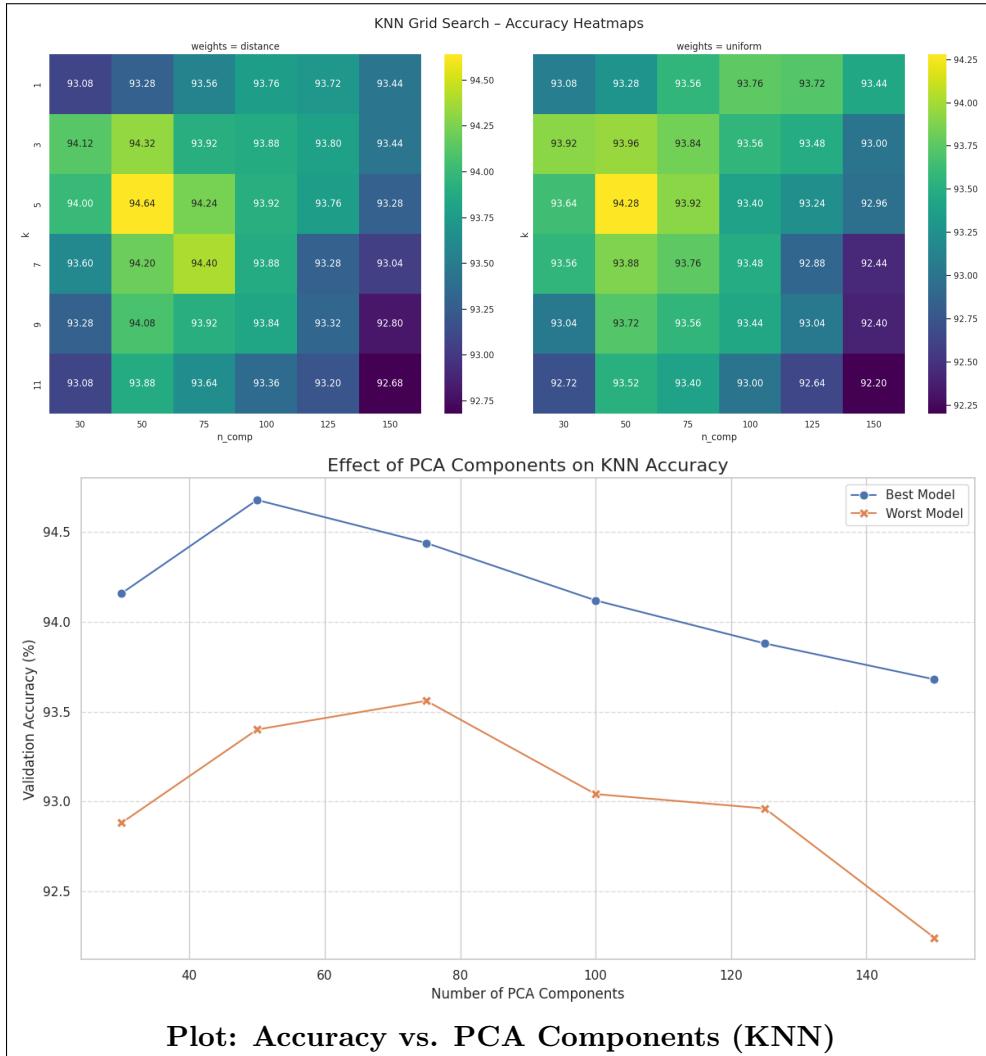


Figure 1: Performance peaked at 50 PCA components and dropped on either side.

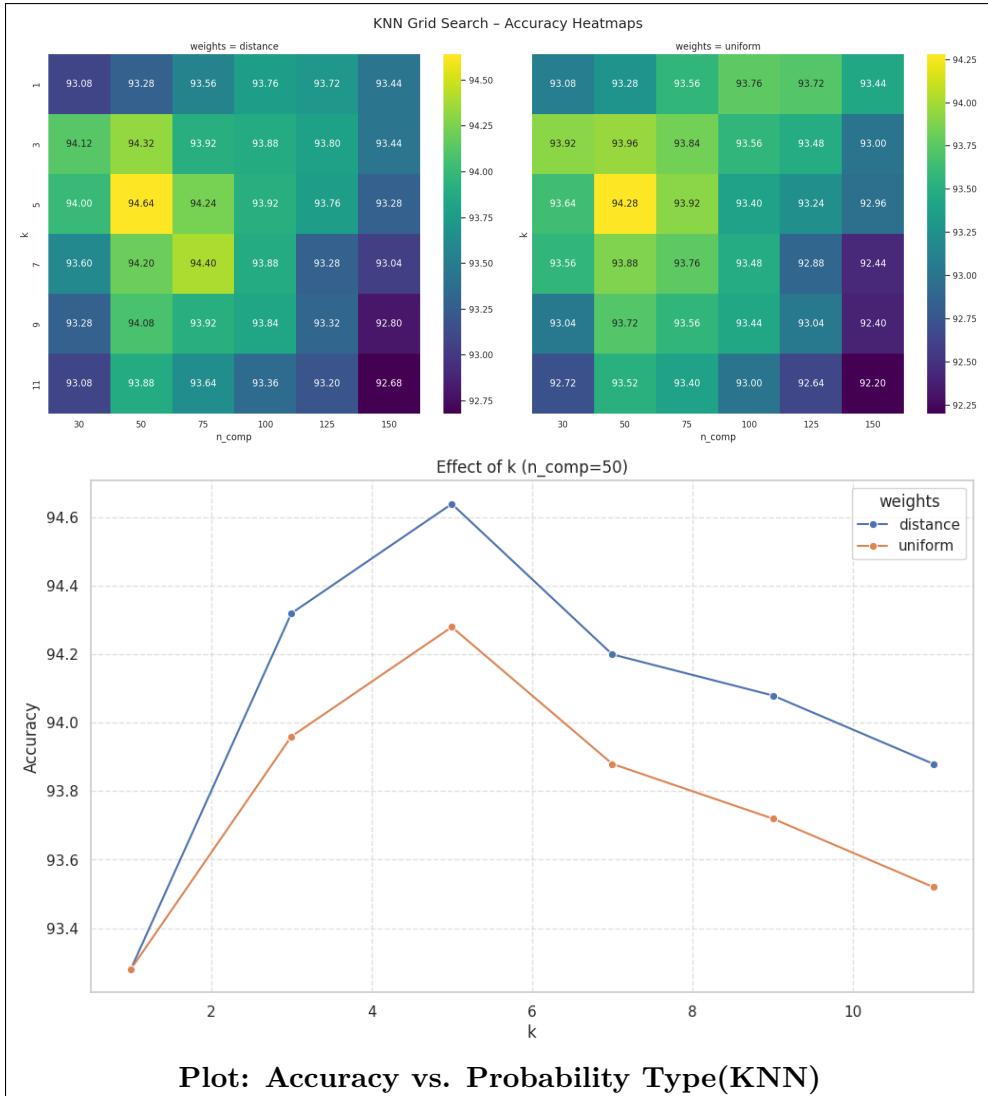


Figure 2: Distance proportional probabilistic choices work better.

- **Key Finding 1:** Taking around 50 components allowed the models to balance both the requirements of taking the variance into account as well as keeping noise in check.
- **Key Finding 2:** Giving probability proportional to distance choices works better than the equal probability.
- **Key Finding 3:** Low k values like 3-5 were performing well.
- **Best Model:** `PCA(n=50) + KNN(k=5, weights='distance')`.

2.2 SoftmaxClassifier Tuning

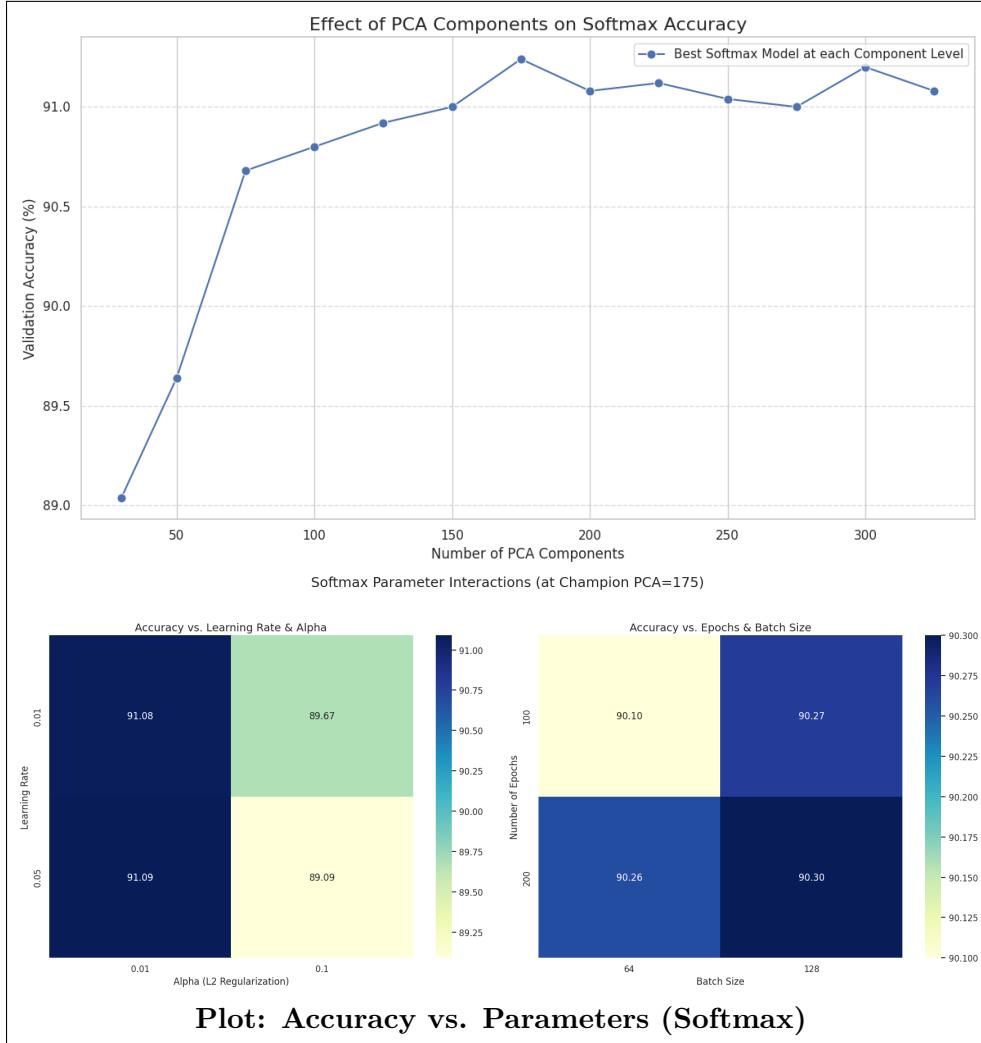


Figure 3: The linear Softmax model's accuracy improved as more PCA components were added.

- **Key Finding 1:** The linear Softmax model was more robust to noise as its accuracy consistently improved as more PCA components were added from below 89 percent with 30 components to just less than 91.25 percent with 175 components. After 175, there isn't much gain so we can stop there.
- **Key Finding 2:** The learning rate and regularization optimal quantities are inversely proportional.
- **Best Model:** `PCA(n=175) + Softmax(epochs=100, lr=0.05, alpha=0.01, batch=128)`.

2.3 Analysis of Model Performance on LDA Features

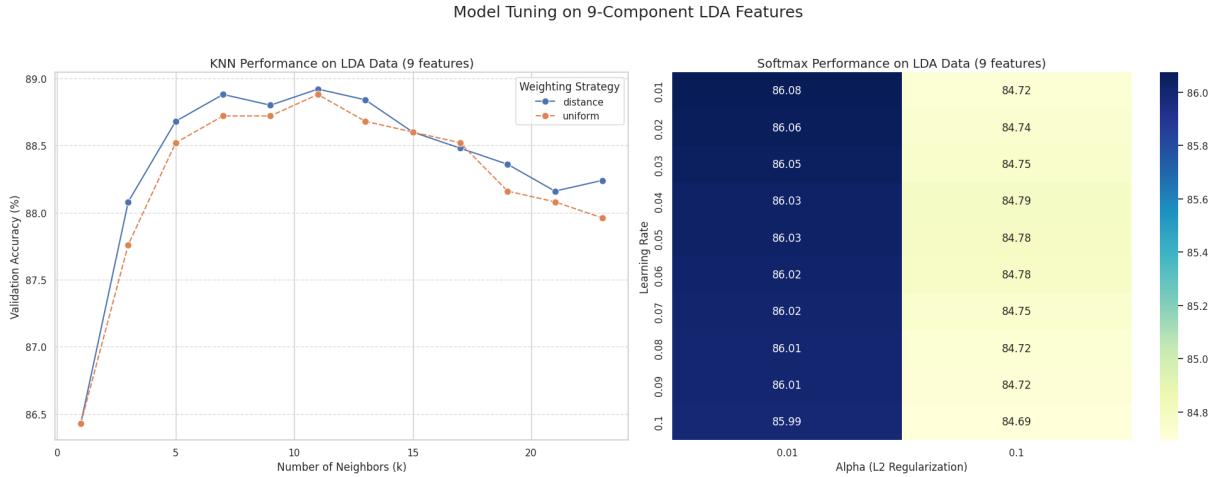


Figure 4: KNN Performance on Left/Softmax Performance on Right

- **KNN on LDA (Left Plot):** The 9 dimensional LDA is sparser, that is why we need 11-13 neighbors to give the best results. Again, distance proportionality makes gives higher accuracy.
- **Softmax on LDA (Right Plot):** The softmax classifier performed decently well on this dataset, but gave slightly lower accuracy. Here low regularization and learning rate gave best accuracy and the vice versa gave the worst.

3 Detailed Summary of Thoughts and Observations

The 5-minute (300s) runtime constraint invalidated the `XGBoostClassifier` because it achieved low accuracy when this training time was taken into consideration inspite of being a complex model.

Hence, for faster training, we have to make an ensemble of simpler models to get a higher accuracy.

We used KNN Classifier and Softmax Classifier for this. The main tuning was to find the optimal parameters for this accounting for accuracy and overfitting.

3.1 Algorithm NumPyKNNClassifier:- Inherently low bias,high variance

- **Things to keep in mind:** Non linear model. When we use high k, KNN becomes a low variance, high bias model but when we use low k, the KNN becomes a low bias, high variance model.
- **Hyperparameter Tuning & Bias-Variance Evaluation:**
 - **k (n_neighbors):** The primary knob for controlling variance.
 - * Low k ($k = 1$) → high variance, low bias. High accuracy.
 - * High k (e.g., $k = 23$) → high bias, low variance. Accuracy dropped.
 - * **Optimal range:** $k \in [3, 7]$ which lies between these two.
 - **weights:**
 - * **uniform:** higher bias, lower variance.
 - * **distance:** lower bias, higher variance.

* **Result:** `distance > uniform`.

Optimal Parameters - PCA($n = 50$) + KNN($k = 5$, weights = <code>distance</code>)
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3.2 Algorithm SoftmaxClassifier:- Inherently high bias, low variance

- **Things to keep in mind:** Linear model
- **Hyperparameter Tuning & Bias-Variance Evaluation:**
 - **alpha (L2 regularization):**
 - * High α (0.1) → increased bias → accuracy drops
 - * Low α (0.01) → decreased bias → best performance
 - **n_epochs, lr, batch_size:** These are for the optimization of the model. Could be used in the best possible state because the running doesn't take very long.

Optimal Parameters - PCA($n = 175$) + Softmax(100, 0.05, 0.01, 128)
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3.3 Algorithms for Dimensionality Reduction

- **PCA (Variance-Driven):**
 - Low components (30) → high bias, low variance. Good for KNN, bad for Softmax.
 - High components (150) → low bias, high variance. Good for Softmax, bad for KNN.
- **LDA (Class-Separation-Driven):**
 - High-bias: compresses 784 → 9 features.
 - Best results: KNN > Softmax.

3.4 Final Observation: Optimal Ensemble

The final ensemble combines complementary models:

1. High-variance model: PCA(50) + KNN
2. High-bias model: PCA(275) + Softmax
3. Specialists: LDA(9) + KNN
4. Stacking with KFold to reduce overfitting

This ensemble achieved **93.64%** accuracy and **93.55%** macro f1-score.

Appendix: Full Hyperparameter Tunings Results

Table 1: PCA+KNN Hyperparameter Results

n_comp	k	weights	Accuracy	F1-Score	Fit (s)
50.000000	5.000000	distance	94.640000	94.590000	0.000000
75.000000	7.000000	distance	94.400000	94.350000	0.000000
50.000000	3.000000	distance	94.320000	94.250000	0.000000
50.000000	5.000000	uniform	94.280000	94.220000	0.000000
75.000000	5.000000	distance	94.240000	94.200000	0.000000
50.000000	7.000000	distance	94.200000	94.150000	0.000000
30.000000	3.000000	distance	94.120000	94.040000	0.000000
50.000000	9.000000	distance	94.080000	94.040000	0.000000
30.000000	5.000000	distance	94.000000	93.940000	0.000000
50.000000	3.000000	uniform	93.960000	93.880000	0.000000
30.000000	3.000000	uniform	93.920000	93.840000	0.000000
75.000000	3.000000	distance	93.920000	93.860000	0.000000
75.000000	5.000000	uniform	93.920000	93.880000	0.000000
75.000000	9.000000	distance	93.920000	93.870000	0.000000
100.000000	5.000000	distance	93.920000	93.870000	0.000000
50.000000	7.000000	uniform	93.880000	93.830000	0.000000
50.000000	11.000000	distance	93.880000	93.830000	0.000000
100.000000	3.000000	distance	93.880000	93.820000	0.000000
100.000000	7.000000	distance	93.880000	93.830000	0.000000
75.000000	3.000000	uniform	93.840000	93.780000	0.000000
100.000000	9.000000	distance	93.840000	93.780000	0.000000
125.000000	3.000000	distance	93.800000	93.740000	0.000000
75.000000	7.000000	uniform	93.760000	93.690000	0.000000
100.000000	1.000000	uniform	93.760000	93.700000	0.000000
100.000000	1.000000	distance	93.760000	93.700000	0.000000
125.000000	5.000000	distance	93.760000	93.710000	0.000000
50.000000	9.000000	uniform	93.720000	93.670000	0.000000
125.000000	1.000000	uniform	93.720000	93.660000	0.000000
125.000000	1.000000	distance	93.720000	93.660000	0.000000
30.000000	5.000000	uniform	93.640000	93.570000	0.000000
75.000000	11.000000	distance	93.640000	93.580000	0.000000
30.000000	7.000000	distance	93.600000	93.550000	0.000000
30.000000	7.000000	uniform	93.560000	93.500000	0.000000
75.000000	1.000000	uniform	93.560000	93.500000	0.000000
75.000000	1.000000	distance	93.560000	93.500000	0.000000
75.000000	9.000000	uniform	93.560000	93.500000	0.000000
100.000000	3.000000	uniform	93.560000	93.510000	0.000000
50.000000	11.000000	uniform	93.520000	93.470000	0.000000
100.000000	7.000000	uniform	93.480000	93.420000	0.000000
125.000000	3.000000	uniform	93.480000	93.420000	0.000000
100.000000	9.000000	uniform	93.440000	93.380000	0.000000
150.000000	1.000000	uniform	93.440000	93.370000	0.000000
150.000000	1.000000	distance	93.440000	93.370000	0.000000

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n_comp	k	weights	Accuracy	F1-Score	Fit (s)
150.000000	3.000000	distance	93.440000	93.370000	0.000000
75.000000	11.000000	uniform	93.400000	93.340000	0.000000
100.000000	5.000000	uniform	93.400000	93.330000	0.000000
100.000000	11.000000	distance	93.360000	93.310000	0.000000
125.000000	9.000000	distance	93.320000	93.260000	0.000000
30.000000	9.000000	distance	93.280000	93.220000	0.000000
50.000000	1.000000	uniform	93.280000	93.210000	0.000000
50.000000	1.000000	distance	93.280000	93.210000	0.000000
125.000000	7.000000	distance	93.280000	93.220000	0.000000
150.000000	5.000000	distance	93.280000	93.230000	0.000000
125.000000	5.000000	uniform	93.240000	93.180000	0.000000
125.000000	11.000000	distance	93.200000	93.150000	0.000000
30.000000	1.000000	uniform	93.080000	93.010000	0.000000
30.000000	1.000000	distance	93.080000	93.010000	0.000000
30.000000	11.000000	distance	93.080000	93.020000	0.000000
30.000000	9.000000	uniform	93.040000	92.970000	0.000000
125.000000	9.000000	uniform	93.040000	92.980000	0.000000
150.000000	7.000000	distance	93.040000	92.970000	0.000000
100.000000	11.000000	uniform	93.000000	92.920000	0.000000
150.000000	3.000000	uniform	93.000000	92.930000	0.000000
150.000000	5.000000	uniform	92.960000	92.900000	0.000000
125.000000	7.000000	uniform	92.880000	92.810000	0.000000
150.000000	9.000000	distance	92.800000	92.730000	0.000000
30.000000	11.000000	uniform	92.720000	92.650000	0.000000
150.000000	11.000000	distance	92.680000	92.620000	0.000000
125.000000	11.000000	uniform	92.640000	92.560000	0.000000
150.000000	7.000000	uniform	92.440000	92.350000	0.000000
150.000000	9.000000	uniform	92.400000	92.310000	0.000000
150.000000	11.000000	uniform	92.200000	92.120000	0.000000

Table 2: PCA+Softmax Hyperparameter Results

Model Configuration (PCA, ep, lr, a, b)	Accuracy	F1-Score	Fit (s)
PCA=175, Softmax (ep=100, lr=0.05, a=0.01, b=128)	91.240000	91.130000	3.510000
PCA=300, Softmax (ep=100, lr=0.05, a=0.01, b=64)	91.200000	91.090000	6.440000
PCA=175, Softmax (ep=100, lr=0.01, a=0.01, b=64)	91.160000	91.040000	3.850000
PCA=300, Softmax (ep=100, lr=0.05, a=0.01, b=128)	91.160000	91.050000	5.700000
PCA=175, Softmax (ep=200, lr=0.01, a=0.01, b=128)	91.120000	91.000000	7.470000
PCA=225, Softmax (ep=100, lr=0.05, a=0.01, b=64)	91.120000	90.990000	5.510000
PCA=175, Softmax (ep=200, lr=0.01, a=0.01, b=64)	91.080000	90.960000	8.330000
PCA=175, Softmax (ep=200, lr=0.05, a=0.01, b=64)	91.080000	90.970000	8.640000
PCA=200, Softmax (ep=200, lr=0.05, a=0.01, b=128)	91.080000	90.970000	8.730000
PCA=225, Softmax (ep=100, lr=0.05, a=0.01, b=128)	91.080000	90.960000	4.310000
PCA=325, Softmax (ep=200, lr=0.05, a=0.01, b=128)	91.080000	90.970000	10.730000
PCA=325, Softmax (ep=200, lr=0.05, a=0.01, b=128)	91.080000	90.970000	4.280000
PCA=175, Softmax (ep=100, lr=0.05, a=0.01, b=64)	91.040000	90.910000	4.180000
PCA=225, Softmax (ep=200, lr=0.05, a=0.01, b=128)	91.040000	90.930000	9.220000
PCA=250, Softmax (ep=100, lr=0.05, a=0.01, b=128)	91.040000	90.930000	4.750000
PCA=250, Softmax (ep=200, lr=0.01, a=0.01, b=64)	91.040000	90.930000	11.120000
PCA=300, Softmax (ep=200, lr=0.01, a=0.01, b=64)	91.040000	90.930000	13.820000
PCA=150, Softmax (ep=100, lr=0.05, a=0.01, b=128)	91.000000	90.880000	2.850000
PCA=150, Softmax (ep=200, lr=0.05, a=0.01, b=128)	91.000000	90.880000	6.610000
PCA=175, Softmax (ep=200, lr=0.05, a=0.01, b=128)	91.000000	90.880000	6.950000
PCA=200, Softmax (ep=100, lr=0.05, a=0.01, b=64)	91.000000	90.860000	4.560000
PCA=200, Softmax (ep=100, lr=0.05, a=0.01, b=128)	91.000000	90.880000	4.320000
PCA=225, Softmax (ep=100, lr=0.01, a=0.01, b=128)	91.000000	90.880000	4.400000
PCA=250, Softmax (ep=200, lr=0.05, a=0.01, b=64)	91.000000	90.900000	11.430000
PCA=275, Softmax (ep=100, lr=0.05, a=0.01, b=64)	91.000000	90.860000	6.120000
PCA=275, Softmax (ep=200, lr=0.05, a=0.01, b=64)	91.000000	90.880000	11.420000
PCA=175, Softmax (ep=100, lr=0.01, a=0.01, b=128)	90.960000	90.840000	3.150000
PCA=325, Softmax (ep=200, lr=0.05, a=0.01, b=64)	90.960000	90.840000	14.520000
PCA=325, Softmax (ep=200, lr=0.05, a=0.01, b=64)	90.960000	90.840000	5.050000
PCA=125, Softmax (ep=100, lr=0.05, a=0.01, b=64)	90.920000	90.810000	3.350000
PCA=150, Softmax (ep=100, lr=0.01, a=0.01, b=64)	90.920000	90.800000	3.880000
PCA=275, Softmax (ep=200, lr=0.05, a=0.01, b=128)	90.920000	90.800000	7.970000
PCA=300, Softmax (ep=200, lr=0.05, a=0.01, b=128)	90.920000	90.800000	9.260000
PCA=325, Softmax (ep=100, lr=0.05, a=0.01, b=64)	90.920000	90.800000	6.710000
PCA=325, Softmax (ep=200, lr=0.01, a=0.01, b=64)	90.920000	90.810000	13.880000
PCA=325, Softmax (ep=100, lr=0.05, a=0.01, b=64)	90.920000	90.800000	5.460000
PCA=325, Softmax (ep=200, lr=0.01, a=0.01, b=64)	90.920000	90.810000	12.980000
PCA=125, Softmax (ep=100, lr=0.05, a=0.01, b=128)	90.880000	90.750000	2.560000
PCA=150, Softmax (ep=200, lr=0.01, a=0.01, b=64)	90.880000	90.760000	7.920000
PCA=150, Softmax (ep=200, lr=0.01, a=0.01, b=128)	90.880000	90.760000	6.420000
PCA=200, Softmax (ep=100, lr=0.01, a=0.01, b=64)	90.880000	90.770000	5.090000
PCA=200, Softmax (ep=200, lr=0.01, a=0.01, b=64)	90.880000	90.770000	9.150000
PCA=225, Softmax (ep=200, lr=0.05, a=0.01, b=64)	90.880000	90.770000	10.450000
PCA=250, Softmax (ep=100, lr=0.01, a=0.01, b=64)	90.880000	90.770000	5.660000
PCA=250, Softmax (ep=100, lr=0.05, a=0.01, b=64)	90.880000	90.760000	5.620000
PCA=250, Softmax (ep=200, lr=0.01, a=0.01, b=128)	90.880000	90.770000	10.280000
PCA=325, Softmax (ep=100, lr=0.05, a=0.01, b=128)	90.880000	90.760000	5.290000
PCA=325, Softmax (ep=100, lr=0.05, a=0.01, b=128)	90.880000	90.760000	4.480000

Table 2 – continued from previous page

Model Configuration (PCA, ep, lr, a, b)	Accuracy	F1-Score	Fit (s)
PCA=125, Softmax (ep=200, lr=0.05, a=0.01, b=128)	90.840000	90.720000	6.070000
PCA=150, Softmax (ep=100, lr=0.05, a=0.01, b=64)	90.840000	90.720000	3.540000
PCA=150, Softmax (ep=200, lr=0.05, a=0.01, b=64)	90.840000	90.710000	7.810000
PCA=200, Softmax (ep=200, lr=0.01, a=0.01, b=128)	90.840000	90.730000	8.140000
PCA=225, Softmax (ep=200, lr=0.01, a=0.01, b=64)	90.840000	90.730000	10.500000
PCA=250, Softmax (ep=200, lr=0.05, a=0.01, b=128)	90.840000	90.730000	9.740000
PCA=275, Softmax (ep=100, lr=0.05, a=0.01, b=128)	90.840000	90.710000	5.520000
PCA=325, Softmax (ep=100, lr=0.01, a=0.01, b=64)	90.840000	90.740000	7.330000
PCA=325, Softmax (ep=100, lr=0.01, a=0.01, b=64)	90.840000	90.740000	3.010000
PCA=100, Softmax (ep=100, lr=0.05, a=0.01, b=128)	90.800000	90.680000	2.170000
PCA=275, Softmax (ep=100, lr=0.01, a=0.01, b=64)	90.800000	90.690000	6.340000
PCA=275, Softmax (ep=200, lr=0.01, a=0.01, b=128)	90.800000	90.700000	11.020000
PCA=300, Softmax (ep=100, lr=0.01, a=0.01, b=64)	90.800000	90.700000	7.050000
PCA=300, Softmax (ep=200, lr=0.05, a=0.01, b=64)	90.800000	90.670000	13.110000
PCA=100, Softmax (ep=200, lr=0.05, a=0.01, b=64)	90.760000	90.650000	5.010000
PCA=125, Softmax (ep=200, lr=0.01, a=0.01, b=64)	90.760000	90.640000	6.610000
PCA=200, Softmax (ep=200, lr=0.05, a=0.01, b=64)	90.760000	90.650000	9.740000
PCA=225, Softmax (ep=200, lr=0.01, a=0.01, b=128)	90.760000	90.650000	9.080000
PCA=275, Softmax (ep=200, lr=0.01, a=0.01, b=64)	90.760000	90.660000	11.870000
PCA=325, Softmax (ep=100, lr=0.01, a=0.01, b=128)	90.760000	90.640000	6.380000
PCA=325, Softmax (ep=100, lr=0.01, a=0.01, b=128)	90.760000	90.640000	2.460000
PCA=125, Softmax (ep=100, lr=0.01, a=0.01, b=64)	90.720000	90.590000	3.530000
PCA=125, Softmax (ep=200, lr=0.01, a=0.01, b=128)	90.720000	90.590000	5.210000
PCA=150, Softmax (ep=100, lr=0.01, a=0.01, b=128)	90.720000	90.590000	3.310000
PCA=200, Softmax (ep=100, lr=0.01, a=0.01, b=128)	90.720000	90.590000	3.680000
PCA=250, Softmax (ep=100, lr=0.01, a=0.01, b=128)	90.720000	90.610000	4.750000
PCA=325, Softmax (ep=200, lr=0.01, a=0.01, b=128)	90.720000	90.610000	10.110000
PCA=325, Softmax (ep=200, lr=0.01, a=0.01, b=128)	90.720000	90.610000	8.740000
PCA=75, Softmax (ep=100, lr=0.05, a=0.01, b=128)	90.680000	90.550000	1.850000
PCA=125, Softmax (ep=200, lr=0.05, a=0.01, b=64)	90.680000	90.550000	7.570000
PCA=225, Softmax (ep=100, lr=0.01, a=0.01, b=64)	90.680000	90.560000	5.310000
PCA=275, Softmax (ep=100, lr=0.01, a=0.01, b=128)	90.680000	90.570000	5.330000
PCA=300, Softmax (ep=200, lr=0.01, a=0.01, b=128)	90.680000	90.570000	10.490000
PCA=300, Softmax (ep=100, lr=0.01, a=0.01, b=128)	90.640000	90.540000	5.850000
PCA=75, Softmax (ep=200, lr=0.05, a=0.01, b=128)	90.600000	90.470000	4.280000
PCA=100, Softmax (ep=100, lr=0.05, a=0.01, b=64)	90.600000	90.480000	2.700000
PCA=100, Softmax (ep=200, lr=0.05, a=0.01, b=128)	90.520000	90.400000	4.140000
PCA=125, Softmax (ep=100, lr=0.01, a=0.01, b=128)	90.480000	90.340000	2.910000
PCA=75, Softmax (ep=100, lr=0.05, a=0.01, b=64)	90.400000	90.280000	2.040000
PCA=100, Softmax (ep=200, lr=0.01, a=0.01, b=64)	90.400000	90.270000	5.250000
PCA=100, Softmax (ep=100, lr=0.01, a=0.01, b=64)	90.320000	90.190000	2.430000
PCA=75, Softmax (ep=200, lr=0.05, a=0.01, b=64)	90.240000	90.090000	4.560000
PCA=75, Softmax (ep=200, lr=0.01, a=0.01, b=64)	90.160000	90.010000	4.900000
PCA=100, Softmax (ep=100, lr=0.01, a=0.01, b=128)	90.080000	89.940000	2.060000
PCA=75, Softmax (ep=100, lr=0.01, a=0.01, b=64)	90.040000	89.890000	1.930000
PCA=75, Softmax (ep=200, lr=0.01, a=0.01, b=128)	90.000000	89.840000	4.140000
PCA=100, Softmax (ep=200, lr=0.01, a=0.01, b=128)	90.000000	89.870000	4.410000
PCA=250, Softmax (ep=200, lr=0.01, a=0.1, b=64)	89.800000	89.680000	12.810000
PCA=175, Softmax (ep=100, lr=0.01, a=0.1, b=128)	89.760000	89.630000	3.500000

Table 2 – continued from previous page

Model Configuration (PCA, ep, lr, a, b)	Accuracy	F1-Score	Fit (s)
PCA=225, Softmax (ep=200, lr=0.01, a=0.1, b=128)	89.760000	89.640000	9.110000
PCA=225, Softmax (ep=200, lr=0.05, a=0.1, b=128)	89.760000	89.650000	9.950000
PCA=250, Softmax (ep=100, lr=0.01, a=0.1, b=128)	89.760000	89.630000	4.780000
PCA=150, Softmax (ep=100, lr=0.01, a=0.1, b=64)	89.720000	89.610000	3.750000
PCA=175, Softmax (ep=100, lr=0.01, a=0.1, b=64)	89.720000	89.590000	4.260000
PCA=225, Softmax (ep=200, lr=0.01, a=0.1, b=64)	89.720000	89.600000	10.640000
PCA=250, Softmax (ep=100, lr=0.01, a=0.1, b=64)	89.720000	89.590000	5.880000
PCA=300, Softmax (ep=100, lr=0.01, a=0.1, b=64)	89.720000	89.590000	6.480000
PCA=300, Softmax (ep=200, lr=0.01, a=0.1, b=64)	89.720000	89.600000	13.760000
PCA=150, Softmax (ep=200, lr=0.05, a=0.1, b=128)	89.680000	89.570000	6.960000
PCA=200, Softmax (ep=200, lr=0.01, a=0.1, b=64)	89.680000	89.560000	9.740000
PCA=250, Softmax (ep=200, lr=0.01, a=0.1, b=128)	89.680000	89.560000	9.890000
PCA=275, Softmax (ep=200, lr=0.01, a=0.1, b=64)	89.680000	89.550000	13.270000
PCA=50, Softmax (ep=100, lr=0.05, a=0.01, b=64)	89.640000	89.510000	1.840000
PCA=150, Softmax (ep=200, lr=0.01, a=0.1, b=64)	89.640000	89.530000	7.000000
PCA=200, Softmax (ep=100, lr=0.01, a=0.1, b=64)	89.640000	89.520000	4.610000
PCA=250, Softmax (ep=200, lr=0.05, a=0.1, b=128)	89.640000	89.530000	10.780000
PCA=300, Softmax (ep=100, lr=0.01, a=0.1, b=128)	89.640000	89.510000	5.520000
PCA=325, Softmax (ep=100, lr=0.01, a=0.1, b=64)	89.640000	89.510000	7.300000
PCA=325, Softmax (ep=100, lr=0.01, a=0.1, b=64)	89.640000	89.510000	7.000000
PCA=50, Softmax (ep=200, lr=0.01, a=0.01, b=64)	89.600000	89.480000	3.160000
PCA=175, Softmax (ep=200, lr=0.01, a=0.1, b=64)	89.600000	89.480000	8.920000
PCA=175, Softmax (ep=200, lr=0.01, a=0.1, b=128)	89.600000	89.480000	7.800000
PCA=200, Softmax (ep=200, lr=0.01, a=0.1, b=128)	89.600000	89.480000	8.390000
PCA=300, Softmax (ep=200, lr=0.05, a=0.1, b=128)	89.600000	89.480000	5.770000
PCA=75, Softmax (ep=100, lr=0.01, a=0.01, b=128)	89.560000	89.400000	1.630000
PCA=200, Softmax (ep=100, lr=0.01, a=0.1, b=128)	89.560000	89.430000	3.680000
PCA=300, Softmax (ep=200, lr=0.01, a=0.1, b=128)	89.560000	89.430000	11.960000
PCA=325, Softmax (ep=100, lr=0.01, a=0.1, b=128)	89.560000	89.430000	6.340000
PCA=325, Softmax (ep=100, lr=0.01, a=0.1, b=128)	89.560000	89.430000	5.840000
PCA=125, Softmax (ep=100, lr=0.01, a=0.1, b=64)	89.520000	89.410000	2.860000
PCA=275, Softmax (ep=100, lr=0.01, a=0.1, b=64)	89.520000	89.390000	5.980000
PCA=275, Softmax (ep=200, lr=0.05, a=0.1, b=128)	89.520000	89.400000	11.130000
PCA=300, Softmax (ep=100, lr=0.05, a=0.1, b=128)	89.520000	89.370000	6.060000
PCA=325, Softmax (ep=200, lr=0.01, a=0.1, b=64)	89.520000	89.390000	9.490000
PCA=325, Softmax (ep=200, lr=0.05, a=0.1, b=128)	89.520000	89.390000	6.300000
PCA=325, Softmax (ep=200, lr=0.01, a=0.1, b=64)	89.520000	89.390000	7.660000
PCA=325, Softmax (ep=200, lr=0.05, a=0.1, b=128)	89.520000	89.390000	8.230000
PCA=50, Softmax (ep=100, lr=0.05, a=0.01, b=128)	89.480000	89.350000	1.430000
PCA=150, Softmax (ep=200, lr=0.01, a=0.1, b=128)	89.480000	89.360000	5.960000
PCA=150, Softmax (ep=200, lr=0.05, a=0.1, b=64)	89.480000	89.360000	8.280000
PCA=175, Softmax (ep=200, lr=0.05, a=0.1, b=128)	89.480000	89.380000	7.500000
PCA=225, Softmax (ep=100, lr=0.01, a=0.1, b=64)	89.480000	89.350000	5.360000
PCA=225, Softmax (ep=100, lr=0.01, a=0.1, b=128)	89.480000	89.350000	4.320000
PCA=275, Softmax (ep=200, lr=0.01, a=0.1, b=128)	89.480000	89.350000	10.750000
PCA=325, Softmax (ep=200, lr=0.01, a=0.1, b=128)	89.480000	89.350000	6.180000
PCA=325, Softmax (ep=200, lr=0.01, a=0.1, b=128)	89.480000	89.350000	5.170000
PCA=50, Softmax (ep=200, lr=0.01, a=0.01, b=128)	89.440000	89.320000	2.880000
PCA=125, Softmax (ep=100, lr=0.05, a=0.1, b=128)	89.440000	89.320000	2.580000

Table 2 – continued from previous page

Model Configuration (PCA, ep, lr, a, b)	Accuracy	F1-Score	Fit (s)
PCA=125, Softmax (ep=200, lr=0.05, a=0.1, b=64)	89.440000	89.310000	6.460000
PCA=150, Softmax (ep=100, lr=0.01, a=0.1, b=128)	89.440000	89.320000	3.090000
PCA=200, Softmax (ep=200, lr=0.05, a=0.1, b=128)	89.440000	89.330000	8.250000
PCA=275, Softmax (ep=100, lr=0.05, a=0.1, b=128)	89.440000	89.310000	5.680000
PCA=300, Softmax (ep=200, lr=0.05, a=0.1, b=64)	89.440000	89.320000	8.910000
PCA=50, Softmax (ep=200, lr=0.05, a=0.01, b=64)	89.400000	89.260000	4.580000
PCA=50, Softmax (ep=200, lr=0.05, a=0.01, b=128)	89.400000	89.280000	3.510000
PCA=150, Softmax (ep=100, lr=0.05, a=0.1, b=128)	89.400000	89.290000	3.050000
PCA=200, Softmax (ep=200, lr=0.05, a=0.1, b=64)	89.400000	89.300000	9.040000
PCA=275, Softmax (ep=100, lr=0.01, a=0.1, b=128)	89.400000	89.270000	5.160000
PCA=275, Softmax (ep=200, lr=0.05, a=0.1, b=64)	89.400000	89.280000	12.380000
PCA=325, Softmax (ep=100, lr=0.05, a=0.1, b=128)	89.400000	89.260000	6.700000
PCA=325, Softmax (ep=100, lr=0.05, a=0.1, b=128)	89.400000	89.260000	3.380000
PCA=125, Softmax (ep=200, lr=0.01, a=0.1, b=128)	89.360000	89.240000	5.760000
PCA=225, Softmax (ep=100, lr=0.05, a=0.1, b=128)	89.360000	89.230000	4.700000
PCA=50, Softmax (ep=100, lr=0.01, a=0.01, b=64)	89.320000	89.190000	1.940000
PCA=125, Softmax (ep=100, lr=0.01, a=0.1, b=128)	89.320000	89.200000	2.430000
PCA=150, Softmax (ep=100, lr=0.05, a=0.1, b=64)	89.320000	89.190000	3.570000
PCA=225, Softmax (ep=200, lr=0.05, a=0.1, b=64)	89.320000	89.210000	11.720000
PCA=325, Softmax (ep=200, lr=0.05, a=0.1, b=64)	89.320000	89.180000	9.910000
PCA=325, Softmax (ep=200, lr=0.05, a=0.1, b=64)	89.320000	89.180000	11.820000
PCA=100, Softmax (ep=100, lr=0.01, a=0.1, b=64)	89.280000	89.150000	2.310000
PCA=125, Softmax (ep=100, lr=0.05, a=0.1, b=64)	89.280000	89.140000	3.380000
PCA=125, Softmax (ep=200, lr=0.01, a=0.1, b=64)	89.280000	89.170000	7.080000
PCA=175, Softmax (ep=200, lr=0.05, a=0.1, b=64)	89.280000	89.180000	9.000000
PCA=200, Softmax (ep=100, lr=0.05, a=0.1, b=128)	89.280000	89.150000	4.100000
PCA=250, Softmax (ep=100, lr=0.05, a=0.1, b=128)	89.280000	89.150000	5.130000
PCA=250, Softmax (ep=200, lr=0.05, a=0.1, b=64)	89.280000	89.170000	12.340000
PCA=325, Softmax (ep=100, lr=0.05, a=0.1, b=64)	89.280000	89.130000	7.480000
PCA=325, Softmax (ep=100, lr=0.05, a=0.1, b=64)	89.280000	89.130000	4.730000
PCA=300, Softmax (ep=100, lr=0.05, a=0.1, b=64)	89.240000	89.080000	6.810000
PCA=100, Softmax (ep=100, lr=0.05, a=0.1, b=128)	89.200000	89.060000	1.910000
PCA=100, Softmax (ep=200, lr=0.01, a=0.1, b=128)	89.200000	89.080000	4.910000
PCA=125, Softmax (ep=200, lr=0.05, a=0.1, b=128)	89.200000	89.090000	5.360000
PCA=275, Softmax (ep=100, lr=0.05, a=0.1, b=64)	89.200000	89.060000	6.670000
PCA=100, Softmax (ep=200, lr=0.05, a=0.1, b=64)	89.120000	88.980000	5.880000
PCA=175, Softmax (ep=100, lr=0.05, a=0.1, b=128)	89.120000	88.980000	3.420000
PCA=50, Softmax (ep=100, lr=0.01, a=0.01, b=128)	89.080000	88.950000	1.720000
PCA=100, Softmax (ep=200, lr=0.01, a=0.1, b=64)	89.080000	88.960000	5.940000
PCA=30, Softmax (ep=200, lr=0.01, a=0.01, b=64)	89.040000	88.890000	3.170000
PCA=250, Softmax (ep=100, lr=0.05, a=0.1, b=64)	89.000000	88.860000	5.700000
PCA=30, Softmax (ep=200, lr=0.01, a=0.01, b=128)	88.920000	88.780000	2.840000
PCA=100, Softmax (ep=100, lr=0.01, a=0.1, b=128)	88.920000	88.780000	1.920000
PCA=100, Softmax (ep=100, lr=0.05, a=0.1, b=64)	88.920000	88.770000	2.630000
PCA=200, Softmax (ep=100, lr=0.05, a=0.1, b=64)	88.920000	88.760000	4.730000
PCA=100, Softmax (ep=200, lr=0.05, a=0.1, b=128)	88.880000	88.760000	4.860000
PCA=30, Softmax (ep=100, lr=0.05, a=0.01, b=64)	88.840000	88.700000	1.910000
PCA=30, Softmax (ep=100, lr=0.05, a=0.01, b=128)	88.840000	88.700000	1.290000
PCA=75, Softmax (ep=100, lr=0.05, a=0.1, b=64)	88.840000	88.690000	2.400000

Table 2 – continued from previous page

Model Configuration (PCA, ep, lr, a, b)	Accuracy	F1-Score	Fit (s)
PCA=30, Softmax (ep=100, lr=0.01, a=0.01, b=64)	88.800000	88.650000	1.020000
PCA=75, Softmax (ep=100, lr=0.01, a=0.1, b=128)	88.800000	88.680000	1.950000
PCA=75, Softmax (ep=100, lr=0.01, a=0.1, b=64)	88.760000	88.640000	2.400000
PCA=75, Softmax (ep=100, lr=0.05, a=0.1, b=128)	88.760000	88.630000	1.840000
PCA=75, Softmax (ep=200, lr=0.01, a=0.1, b=128)	88.680000	88.560000	3.330000
PCA=75, Softmax (ep=200, lr=0.05, a=0.1, b=128)	88.640000	88.500000	4.400000
PCA=225, Softmax (ep=100, lr=0.05, a=0.1, b=64)	88.640000	88.480000	5.400000
PCA=75, Softmax (ep=200, lr=0.01, a=0.1, b=64)	88.600000	88.470000	4.530000
PCA=30, Softmax (ep=100, lr=0.01, a=0.01, b=128)	88.560000	88.400000	0.760000
PCA=75, Softmax (ep=200, lr=0.05, a=0.1, b=64)	88.560000	88.430000	5.470000
PCA=175, Softmax (ep=100, lr=0.05, a=0.1, b=64)	88.480000	88.310000	3.830000
PCA=30, Softmax (ep=200, lr=0.05, a=0.01, b=128)	88.360000	88.190000	2.650000
PCA=30, Softmax (ep=200, lr=0.05, a=0.01, b=64)	88.320000	88.160000	2.200000
PCA=50, Softmax (ep=100, lr=0.01, a=0.1, b=64)	88.280000	88.140000	1.180000
PCA=50, Softmax (ep=100, lr=0.05, a=0.1, b=128)	88.280000	88.130000	1.320000
PCA=50, Softmax (ep=100, lr=0.05, a=0.1, b=64)	88.160000	88.030000	2.140000
PCA=50, Softmax (ep=200, lr=0.01, a=0.1, b=128)	88.160000	88.030000	3.020000
PCA=50, Softmax (ep=100, lr=0.01, a=0.1, b=128)	88.120000	87.980000	1.190000
PCA=50, Softmax (ep=200, lr=0.05, a=0.1, b=128)	88.000000	87.860000	3.010000
PCA=50, Softmax (ep=200, lr=0.01, a=0.1, b=64)	87.960000	87.830000	3.840000
PCA=50, Softmax (ep=200, lr=0.05, a=0.1, b=64)	87.920000	87.750000	3.580000
PCA=30, Softmax (ep=200, lr=0.01, a=0.1, b=64)	87.430000	87.330000	3.620000
PCA=30, Softmax (ep=100, lr=0.01, a=0.1, b=64)	87.310000	87.200000	1.100000
PCA=30, Softmax (ep=200, lr=0.01, a=0.1, b=128)	87.310000	87.210000	2.640000
PCA=30, Softmax (ep=100, lr=0.01, a=0.1, b=128)	87.270000	87.150000	1.040000
PCA=30, Softmax (ep=100, lr=0.05, a=0.1, b=128)	86.990000	86.840000	0.760000
PCA=30, Softmax (ep=200, lr=0.05, a=0.1, b=64)	86.750000	86.600000	3.640000
PCA=30, Softmax (ep=200, lr=0.05, a=0.1, b=128)	86.750000	86.610000	2.680000
PCA=30, Softmax (ep=100, lr=0.05, a=0.1, b=64)	86.710000	86.540000	1.000000

Table 3: Models on LDA data Performance Comparison

Model	Accuracy (%)	F1-Score (%)	Fit (s)
LDA + KNN (k=11, w='distance')	88.920000	88.720000	0.000000
LDA + KNN (k=7, w='distance')	88.880000	88.660000	0.000000
LDA + KNN (k=11, w='uniform')	88.880000	88.670000	0.000000
LDA + KNN (k=13, w='distance')	88.840000	88.630000	0.000000
LDA + KNN (k=9, w='distance')	88.800000	88.600000	0.000000
LDA + KNN (k=7, w='uniform')	88.720000	88.510000	0.000000
LDA + KNN (k=9, w='uniform')	88.720000	88.510000	0.000000
LDA + KNN (k=5, w='distance')	88.680000	88.470000	0.000000
LDA + KNN (k=13, w='uniform')	88.680000	88.470000	0.000000
LDA + KNN (k=15, w='uniform')	88.600000	88.390000	0.000000
LDA + KNN (k=15, w='distance')	88.600000	88.400000	0.000000
LDA + KNN (k=5, w='uniform')	88.520000	88.290000	0.000000
LDA + KNN (k=17, w='uniform')	88.520000	88.310000	0.000000
LDA + KNN (k=17, w='distance')	88.480000	88.280000	0.000000
LDA + KNN (k=19, w='distance')	88.360000	88.160000	0.000000
LDA + KNN (k=23, w='distance')	88.240000	88.040000	0.000000
LDA + KNN (k=19, w='uniform')	88.160000	87.940000	0.000000
LDA + KNN (k=21, w='distance')	88.160000	87.960000	0.000000
LDA + KNN (k=3, w='distance')	88.080000	87.860000	0.000000
LDA + KNN (k=21, w='uniform')	88.080000	87.870000	0.000000
LDA + KNN (k=23, w='uniform')	87.960000	87.760000	0.000000
LDA + KNN (k=3, w='uniform')	87.760000	87.530000	0.000000
LDA + KNN (k=1, w='uniform')	86.430000	86.220000	0.000000
LDA + KNN (k=1, w='distance')	86.430000	86.220000	0.000000
LDA + Softmax (ep=200, lr=0.08, a=0.01, b=32)	86.190000	86.020000	5.150000
LDA + Softmax (ep=150, lr=0.01, a=0.01, b=64)	86.150000	85.960000	2.700000
LDA + Softmax (ep=200, lr=0.05, a=0.01, b=32)	86.150000	85.980000	4.970000
LDA + Softmax (ep=200, lr=0.06, a=0.01, b=32)	86.150000	85.980000	6.000000
LDA + Softmax (ep=200, lr=0.07, a=0.01, b=64)	86.150000	85.970000	3.510000
LDA + Softmax (ep=200, lr=0.09, a=0.01, b=32)	86.150000	85.980000	4.940000
LDA + Softmax (ep=200, lr=0.09, a=0.01, b=64)	86.150000	85.980000	3.280000
LDA + Softmax (ep=200, lr=0.1, a=0.01, b=32)	86.150000	85.980000	4.850000
LDA + Softmax (ep=150, lr=0.03, a=0.01, b=32)	86.110000	85.930000	3.670000
LDA + Softmax (ep=150, lr=0.04, a=0.01, b=32)	86.110000	85.930000	4.230000
LDA + Softmax (ep=200, lr=0.03, a=0.01, b=32)	86.110000	85.930000	5.310000
LDA + Softmax (ep=200, lr=0.04, a=0.01, b=32)	86.110000	85.940000	4.850000
LDA + Softmax (ep=200, lr=0.05, a=0.01, b=64)	86.110000	85.930000	3.350000
LDA + Softmax (ep=200, lr=0.06, a=0.01, b=64)	86.110000	85.930000	3.130000
LDA + Softmax (ep=200, lr=0.07, a=0.01, b=32)	86.110000	85.940000	5.480000
LDA + Softmax (ep=200, lr=0.08, a=0.01, b=64)	86.110000	85.940000	3.370000
LDA + Softmax (ep=200, lr=0.1, a=0.01, b=64)	86.110000	85.940000	3.220000
LDA + Softmax (ep=250, lr=0.01, a=0.01, b=64)	86.110000	85.920000	4.180000
LDA + Softmax (ep=100, lr=0.01, a=0.01, b=32)	86.070000	85.890000	2.560000
LDA + Softmax (ep=150, lr=0.01, a=0.01, b=32)	86.070000	85.890000	3.850000
LDA + Softmax (ep=150, lr=0.02, a=0.01, b=32)	86.070000	85.890000	3.650000
LDA + Softmax (ep=150, lr=0.02, a=0.01, b=64)	86.070000	85.890000	2.310000

Continued on next page

Model	Accuracy (%)	F1-Score (%)	Fit (s)
LDA + Softmax (ep=200, lr=0.01, a=0.01, b=64)	86.070000	85.890000	3.270000
LDA + Softmax (ep=200, lr=0.02, a=0.01, b=32)	86.070000	85.890000	5.210000
LDA + Softmax (ep=200, lr=0.02, a=0.01, b=64)	86.070000	85.890000	3.370000
LDA + Softmax (ep=200, lr=0.03, a=0.01, b=64)	86.070000	85.890000	3.250000
LDA + Softmax (ep=200, lr=0.04, a=0.01, b=64)	86.070000	85.890000	3.220000
LDA + Softmax (ep=250, lr=0.01, a=0.01, b=32)	86.070000	85.890000	7.280000
LDA + Softmax (ep=250, lr=0.02, a=0.01, b=32)	86.070000	85.890000	7.500000
LDA + Softmax (ep=250, lr=0.02, a=0.01, b=64)	86.070000	85.890000	4.330000
LDA + Softmax (ep=250, lr=0.03, a=0.01, b=64)	86.070000	85.890000	3.540000
LDA + Softmax (ep=100, lr=0.01, a=0.01, b=64)	86.030000	85.840000	1.590000
LDA + Softmax (ep=100, lr=0.02, a=0.01, b=32)	86.030000	85.850000	2.550000
LDA + Softmax (ep=100, lr=0.02, a=0.01, b=64)	86.030000	85.850000	1.690000
LDA + Softmax (ep=100, lr=0.03, a=0.01, b=64)	86.030000	85.850000	1.810000
LDA + Softmax (ep=100, lr=0.04, a=0.01, b=64)	86.030000	85.850000	1.740000
LDA + Softmax (ep=100, lr=0.09, a=0.01, b=64)	86.030000	85.860000	1.870000
LDA + Softmax (ep=150, lr=0.03, a=0.01, b=64)	86.030000	85.850000	2.300000
LDA + Softmax (ep=150, lr=0.04, a=0.01, b=64)	86.030000	85.850000	2.330000
LDA + Softmax (ep=150, lr=0.05, a=0.01, b=32)	86.030000	85.850000	4.370000
LDA + Softmax (ep=150, lr=0.05, a=0.01, b=64)	86.030000	85.850000	2.400000
LDA + Softmax (ep=150, lr=0.06, a=0.01, b=64)	86.030000	85.850000	2.810000
LDA + Softmax (ep=150, lr=0.07, a=0.01, b=64)	86.030000	85.850000	2.470000
LDA + Softmax (ep=150, lr=0.08, a=0.01, b=64)	86.030000	85.850000	2.500000
LDA + Softmax (ep=200, lr=0.01, a=0.01, b=32)	86.030000	85.850000	5.370000
LDA + Softmax (ep=250, lr=0.03, a=0.01, b=32)	86.030000	85.850000	6.630000
LDA + Softmax (ep=250, lr=0.04, a=0.01, b=64)	86.030000	85.850000	4.120000
LDA + Softmax (ep=250, lr=0.05, a=0.01, b=64)	86.030000	85.850000	4.010000
LDA + Softmax (ep=250, lr=0.06, a=0.01, b=64)	86.030000	85.850000	4.510000
LDA + Softmax (ep=100, lr=0.05, a=0.01, b=64)	85.990000	85.810000	1.740000
LDA + Softmax (ep=100, lr=0.07, a=0.01, b=32)	85.990000	85.830000	2.650000
LDA + Softmax (ep=100, lr=0.1, a=0.01, b=64)	85.990000	85.820000	1.800000
LDA + Softmax (ep=150, lr=0.09, a=0.01, b=32)	85.990000	85.810000	4.040000
LDA + Softmax (ep=150, lr=0.09, a=0.01, b=64)	85.990000	85.810000	2.290000
LDA + Softmax (ep=150, lr=0.1, a=0.01, b=64)	85.990000	85.810000	2.620000
LDA + Softmax (ep=250, lr=0.05, a=0.01, b=32)	85.990000	85.810000	7.410000
LDA + Softmax (ep=250, lr=0.06, a=0.01, b=32)	85.990000	85.810000	3.910000
LDA + Softmax (ep=250, lr=0.07, a=0.01, b=32)	85.990000	85.820000	5.580000
LDA + Softmax (ep=250, lr=0.08, a=0.01, b=32)	85.990000	85.820000	5.690000
LDA + Softmax (ep=250, lr=0.08, a=0.01, b=64)	85.990000	85.810000	2.850000
LDA + Softmax (ep=250, lr=0.09, a=0.01, b=64)	85.990000	85.810000	3.350000
LDA + Softmax (ep=250, lr=0.1, a=0.01, b=64)	85.990000	85.810000	3.640000
LDA + Softmax (ep=100, lr=0.03, a=0.01, b=32)	85.950000	85.780000	2.650000
LDA + Softmax (ep=100, lr=0.04, a=0.01, b=32)	85.950000	85.780000	2.800000
LDA + Softmax (ep=100, lr=0.05, a=0.01, b=32)	85.950000	85.780000	2.500000
LDA + Softmax (ep=100, lr=0.06, a=0.01, b=32)	85.950000	85.780000	2.650000
LDA + Softmax (ep=100, lr=0.06, a=0.01, b=64)	85.950000	85.780000	1.630000
LDA + Softmax (ep=100, lr=0.07, a=0.01, b=64)	85.950000	85.780000	1.660000
LDA + Softmax (ep=100, lr=0.08, a=0.01, b=64)	85.950000	85.780000	1.550000
LDA + Softmax (ep=150, lr=0.06, a=0.01, b=32)	85.950000	85.770000	3.650000

Continued on next page

Model	Accuracy (%)	F1-Score (%)	Fit (s)
LDA + Softmax (ep=150, lr=0.07, a=0.01, b=32)	85.950000	85.770000	3.820000
LDA + Softmax (ep=150, lr=0.08, a=0.01, b=32)	85.950000	85.770000	3.790000
LDA + Softmax (ep=150, lr=0.1, a=0.01, b=32)	85.950000	85.770000	3.760000
LDA + Softmax (ep=250, lr=0.04, a=0.01, b=32)	85.950000	85.770000	7.210000
LDA + Softmax (ep=250, lr=0.07, a=0.01, b=64)	85.950000	85.770000	2.650000
LDA + Softmax (ep=250, lr=0.09, a=0.01, b=32)	85.910000	85.740000	6.140000
LDA + Softmax (ep=100, lr=0.08, a=0.01, b=32)	85.870000	85.710000	2.900000
LDA + Softmax (ep=100, lr=0.09, a=0.01, b=32)	85.870000	85.710000	2.930000
LDA + Softmax (ep=100, lr=0.1, a=0.01, b=32)	85.870000	85.710000	2.890000
LDA + Softmax (ep=250, lr=0.1, a=0.01, b=32)	85.870000	85.700000	6.720000
LDA + Softmax (ep=200, lr=0.06, a=0.1, b=64)	84.990000	84.790000	3.370000
LDA + Softmax (ep=200, lr=0.03, a=0.1, b=32)	84.950000	84.750000	5.680000
LDA + Softmax (ep=200, lr=0.05, a=0.1, b=64)	84.950000	84.750000	3.540000
LDA + Softmax (ep=200, lr=0.04, a=0.1, b=32)	84.910000	84.720000	5.250000
LDA + Softmax (ep=200, lr=0.07, a=0.1, b=64)	84.910000	84.710000	3.330000
LDA + Softmax (ep=200, lr=0.1, a=0.1, b=64)	84.910000	84.710000	3.400000
LDA + Softmax (ep=100, lr=0.08, a=0.1, b=64)	84.870000	84.670000	1.710000
LDA + Softmax (ep=100, lr=0.09, a=0.1, b=64)	84.870000	84.670000	1.630000
LDA + Softmax (ep=200, lr=0.02, a=0.1, b=32)	84.870000	84.680000	5.160000
LDA + Softmax (ep=200, lr=0.04, a=0.1, b=64)	84.870000	84.670000	3.490000
LDA + Softmax (ep=200, lr=0.05, a=0.1, b=32)	84.870000	84.670000	5.400000
LDA + Softmax (ep=200, lr=0.08, a=0.1, b=64)	84.870000	84.670000	3.010000
LDA + Softmax (ep=200, lr=0.09, a=0.1, b=64)	84.870000	84.670000	3.620000
LDA + Softmax (ep=100, lr=0.05, a=0.1, b=32)	84.830000	84.630000	2.310000
LDA + Softmax (ep=100, lr=0.07, a=0.1, b=64)	84.830000	84.630000	1.620000
LDA + Softmax (ep=100, lr=0.1, a=0.1, b=64)	84.830000	84.630000	1.580000
LDA + Softmax (ep=150, lr=0.04, a=0.1, b=64)	84.830000	84.650000	2.500000
LDA + Softmax (ep=150, lr=0.05, a=0.1, b=64)	84.830000	84.650000	2.360000
LDA + Softmax (ep=200, lr=0.03, a=0.1, b=64)	84.830000	84.640000	3.190000
LDA + Softmax (ep=200, lr=0.06, a=0.1, b=32)	84.830000	84.630000	5.390000
LDA + Softmax (ep=200, lr=0.1, a=0.1, b=32)	84.830000	84.610000	5.110000
LDA + Softmax (ep=100, lr=0.04, a=0.1, b=32)	84.790000	84.590000	2.670000
LDA + Softmax (ep=100, lr=0.06, a=0.1, b=64)	84.790000	84.590000	1.760000
LDA + Softmax (ep=100, lr=0.07, a=0.1, b=32)	84.790000	84.580000	2.810000
LDA + Softmax (ep=100, lr=0.09, a=0.1, b=32)	84.790000	84.560000	2.470000
LDA + Softmax (ep=150, lr=0.02, a=0.1, b=64)	84.790000	84.600000	2.570000
LDA + Softmax (ep=150, lr=0.04, a=0.1, b=32)	84.790000	84.600000	3.860000
LDA + Softmax (ep=150, lr=0.06, a=0.1, b=64)	84.790000	84.600000	2.240000
LDA + Softmax (ep=200, lr=0.01, a=0.1, b=32)	84.790000	84.590000	5.220000
LDA + Softmax (ep=200, lr=0.01, a=0.1, b=64)	84.790000	84.590000	3.090000
LDA + Softmax (ep=200, lr=0.02, a=0.1, b=64)	84.790000	84.590000	3.050000
LDA + Softmax (ep=200, lr=0.07, a=0.1, b=32)	84.790000	84.590000	4.830000
LDA + Softmax (ep=200, lr=0.09, a=0.1, b=32)	84.790000	84.570000	5.500000
LDA + Softmax (ep=100, lr=0.02, a=0.1, b=32)	84.750000	84.540000	2.690000
LDA + Softmax (ep=100, lr=0.04, a=0.1, b=64)	84.750000	84.540000	1.640000
LDA + Softmax (ep=100, lr=0.05, a=0.1, b=64)	84.750000	84.550000	1.630000
LDA + Softmax (ep=100, lr=0.06, a=0.1, b=32)	84.750000	84.550000	2.560000
LDA + Softmax (ep=100, lr=0.08, a=0.1, b=32)	84.750000	84.530000	2.390000

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Model	Accuracy (%)	F1-Score (%)	Fit (s)
LDA + Softmax (ep=150, lr=0.01, a=0.1, b=32)	84.750000	84.560000	4.220000
LDA + Softmax (ep=150, lr=0.01, a=0.1, b=64)	84.750000	84.550000	2.350000
LDA + Softmax (ep=150, lr=0.03, a=0.1, b=32)	84.750000	84.560000	3.710000
LDA + Softmax (ep=150, lr=0.03, a=0.1, b=64)	84.750000	84.560000	2.560000
LDA + Softmax (ep=150, lr=0.07, a=0.1, b=64)	84.750000	84.560000	2.200000
LDA + Softmax (ep=100, lr=0.01, a=0.1, b=64)	84.710000	84.510000	1.740000
LDA + Softmax (ep=100, lr=0.03, a=0.1, b=32)	84.710000	84.510000	2.850000
LDA + Softmax (ep=100, lr=0.1, a=0.1, b=32)	84.710000	84.480000	2.800000
LDA + Softmax (ep=150, lr=0.02, a=0.1, b=32)	84.710000	84.520000	4.000000
LDA + Softmax (ep=150, lr=0.06, a=0.1, b=32)	84.710000	84.530000	4.100000
LDA + Softmax (ep=150, lr=0.08, a=0.1, b=64)	84.710000	84.520000	2.530000
LDA + Softmax (ep=150, lr=0.09, a=0.1, b=64)	84.710000	84.530000	2.470000
LDA + Softmax (ep=200, lr=0.08, a=0.1, b=32)	84.710000	84.500000	5.690000
LDA + Softmax (ep=250, lr=0.01, a=0.1, b=64)	84.710000	84.510000	4.330000
LDA + Softmax (ep=250, lr=0.02, a=0.1, b=32)	84.710000	84.510000	6.880000
LDA + Softmax (ep=250, lr=0.03, a=0.1, b=32)	84.710000	84.510000	5.440000
LDA + Softmax (ep=250, lr=0.04, a=0.1, b=32)	84.710000	84.500000	5.660000
LDA + Softmax (ep=250, lr=0.05, a=0.1, b=32)	84.710000	84.510000	5.510000
LDA + Softmax (ep=100, lr=0.03, a=0.1, b=64)	84.670000	84.460000	1.620000
LDA + Softmax (ep=150, lr=0.05, a=0.1, b=32)	84.670000	84.480000	3.820000
LDA + Softmax (ep=150, lr=0.08, a=0.1, b=32)	84.670000	84.490000	3.840000
LDA + Softmax (ep=150, lr=0.1, a=0.1, b=64)	84.670000	84.490000	2.490000
LDA + Softmax (ep=250, lr=0.01, a=0.1, b=32)	84.670000	84.470000	6.650000
LDA + Softmax (ep=250, lr=0.02, a=0.1, b=64)	84.670000	84.470000	4.910000
LDA + Softmax (ep=250, lr=0.03, a=0.1, b=64)	84.670000	84.470000	4.210000
LDA + Softmax (ep=250, lr=0.04, a=0.1, b=64)	84.670000	84.470000	3.310000
LDA + Softmax (ep=250, lr=0.05, a=0.1, b=64)	84.670000	84.470000	3.400000
LDA + Softmax (ep=250, lr=0.06, a=0.1, b=32)	84.670000	84.470000	6.850000
LDA + Softmax (ep=250, lr=0.06, a=0.1, b=64)	84.670000	84.470000	3.710000
LDA + Softmax (ep=100, lr=0.01, a=0.1, b=32)	84.630000	84.430000	2.900000
LDA + Softmax (ep=100, lr=0.02, a=0.1, b=64)	84.630000	84.430000	1.820000
LDA + Softmax (ep=150, lr=0.07, a=0.1, b=32)	84.630000	84.440000	4.460000
LDA + Softmax (ep=250, lr=0.07, a=0.1, b=32)	84.630000	84.430000	7.260000
LDA + Softmax (ep=250, lr=0.07, a=0.1, b=64)	84.630000	84.430000	3.630000
LDA + Softmax (ep=250, lr=0.08, a=0.1, b=64)	84.630000	84.430000	4.490000
LDA + Softmax (ep=250, lr=0.08, a=0.1, b=32)	84.590000	84.390000	3.560000
LDA + Softmax (ep=250, lr=0.09, a=0.1, b=64)	84.590000	84.390000	2.260000
LDA + Softmax (ep=250, lr=0.1, a=0.1, b=64)	84.590000	84.390000	2.650000
LDA + Softmax (ep=150, lr=0.09, a=0.1, b=32)	84.550000	84.370000	3.900000
LDA + Softmax (ep=150, lr=0.1, a=0.1, b=32)	84.550000	84.380000	3.850000
LDA + Softmax (ep=250, lr=0.09, a=0.1, b=32)	84.550000	84.350000	4.850000
LDA + Softmax (ep=250, lr=0.1, a=0.1, b=32)	84.470000	84.270000	5.420000