Final Project Report

Veracity

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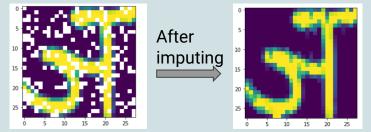
Problem Statement

- The objective is to classify handwritten characters of Devanagari script. Each character belongs to a class.
- Training set of 1000 images belonging to 10 classes were provided where every training sample has missing pixels which are to be imputed.
- Test set comprises of 1000 images belonging to 10 classes.

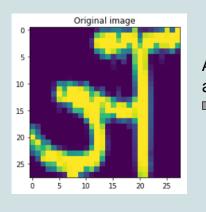
Solution approaches

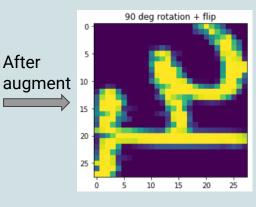
Preprocessing

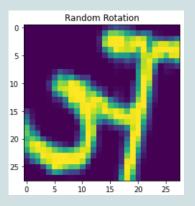
Missing values filled using interpolatation and KNNImputer(). Tuned parameter n_neighbors.

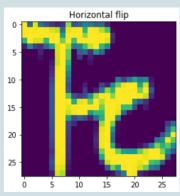


• Image Augmentation: Rotated the image by 90 degrees and flipped it vertically. Random rotation is also imposed on the original images along with horizontal flipping.









Solution approaches

Motivation:

To analyze the effect of image augmentation and compare the performances of SVM, kNN and Random Forest (RF) with and without PCA.

Approach 1

- Training set: Training set without augmentation.
- Model: kNN with cross validation and gridsearch to tune hyperparameters.

Approach 2

- Training set: Training set without augmentation.
- Model: SVM with cross validation and gridsearch to tune hyperparameters.

Approach 3

- Training set: Training set without augmentation.
- Model: RF with cross validation and gridsearch to tune hyperparameters.

Solution approaches (contd..)

Approach 4

- Training set: Training set with all the augmentations.
- Model: PCA with number of components that explain atleast 75% of the variance + kNN with cross validation and gridsearch to tune hyperparameters.

Approach 5

- Training set: Training set with all the augmentations.
- Model: PCA with number of components that explain atleast 75% of the variance + SVM with cross validation and gridsearch to tune hyperparameters.

Approach 6

- Training set: Training set with all the augmentations.
- Model: PCA with number of components that explain atleast 75% of the variance + RF with cross validation and gridsearch to tune hyperparameters.

Approach 7

- Training set: Training set with all the augmentations.
- Model: PCA with number of components that explain atleast 75% of the variance + Voting classifier of SVM, kNN and RF.

Best Solution

Data Preprocessing:

• Imputation of missing pixels was done with pandas *linear* interpolation followed by kNN Imputer (*n_neighbors* parameter was tuned and set to 3).

```
imputer = KNNImputer(n_neighbors=3)
for i in range(10):
    for j in tqdm(range(1, 1001)):
        path = './Training_Dataset/character_' + str(i) + '/' + str(j) + '.csv'
        img = np.loadtxt(path, delimiter=',')
        img = pd.DataFrame(img)
        img.interpolate(method='linear',inplace=True)
        img = imputer.fit_transform(img.to_numpy())
```

 Image augmentation with random, horizontal and vertical rotations and flips or a combination of both was performed to artificially expand training data.

```
def random_rotation(image_array: ndarray):
    random_degree = random.uniform(-25, 25)
    return sk.transform.rotate(image_array, random_degree)

def vertical_flip(image_array: ndarray):
    return sk.transform.rotate(image_array, 180)

def horizontal_flip(image_array: ndarray):|
    return image_array[:, ::-1]

def rotflip(image_array):
    return np.flip(np.rot90(image_array, axes=(1,0)), axis=1)
```

Best Solution (contd..)

Train Test Split, PCA and SVM:

- The original training data was shuffled and split into train and validation data in a stratified fashion in a ratio of 80:20.
- PCA (with components that explain atleast 75% variance) was fit on the train data and both train and validation data were transformed.
- SVM model with RBF kernel was trained and GridSearchCV with 3 fold cross validation was used to tune the following SVC parameters:

```
 \label{eq:continuous} \begin{tabular}{ll} & \class_{\tt weight':[None,'balanced'], 'C':[1,10,100]} \\ & \cla
```

We got the best parameters as:

```
{'gamma':'scale', 'shrinking':True, 'class_weight': 'balanced', 'C':10}
```

 Finally PCA(0.75) was fit on original train data and then transformed. SVC with the best parameters found earlier was trained on this data and saved as a pickle file.

Results

Approach	Score on test-set
no Augmentation (noAug) + kNN	0.216
noAug + SVM	0.233
noAug + Random Forest	0.228
PCA + Aug + kNN	0.968
PCA + Aug + SVM	0.971
PCA + Aug + RF	0.956
PCA + Aug + Voting Classifier	0.955

Conclusion

- Looked at how imputing is done by combining linear interpolations and kNN imputer.
- Compared the effect of image augmentation and dimensionality reduction on kNN, SVM and RF.
- Image augmentation and PCA improved performance.
- Using SVM as the classifier yielded the highest score on test set.

References

- 1. https://pandas.pydata.org/pandas-docs/stable/reference/api/pandas.DataFrame.interpolate.html
- 2. PCA, kNN, SVM and RF from Sklearn documentation.
- 3. https://medium.com/@thimblot/data-augmentation-boost-your-image-dataset-with-few-lines-of-python-155c2dc1baec