

# RK2

June 5, 2019

1 2

2 2. .

..5-24

.  
· / ·  
:

Adjusted Rand index

Adjusted Mutual Information

Homogeneity, completeness, V-measure

3

```
In [2]: import numpy as np
```

```
from PIL import Image
```

```
from sklearn.decomposition import PCA
```

```
from sklearn.preprocessing import scale
```

```
from sklearn import metrics
```

```
from sklearn.cluster import KMeans
```

```
from IPython.display import display
```

```
from tqdm import tqdm_notebook as tqdm
```

```
import matplotlib.pyplot as plt
```

```
import matplotlib.cm as cm
```

```
from scipy import ndimage as ndi
```

```
from skimage.morphology import medial_axis
```

```
from scipy import ndimage as ndi
```

```
from skimage.morphology import medial_axis
```

```

from skimage.morphology import skeletonize

from scipy.spatial import Delaunay

from sklearn.model_selection import train_test_split

In [7]: data = np.load("hiragana.npz")['arr_0']

In [8]: X = []
        Y = []
        for index, letter in enumerate(data):
            for variant in letter:
                X.append(variant)
                Y.append(index)

In [38]: def moment(array, m1, m2):
            moment = 0
            for y, ver in enumerate(array):
                for x, hor in enumerate(ver):
                    moment += pow(x, m1) * pow(y, m2) * hor
            return moment

        def center(array):
            x = moment(array, 1, 0) / moment(array, 0, 0)
            y = moment(array, 0, 1) / moment(array, 0, 0)
            return (x, y)

        def translate(array, x, y):
            buffer = np.roll(array, -x, axis=1)
            buffer = np.roll(buffer, -y, axis=0)
            return buffer

        def centeredarray(array):
            buffer = []
            for pic in tqdm(array):
                shape = pic.shape
                centroid = center(pic)
                delta_x = -shape[1] / 2 + centroid[0]
                delta_y = -shape[0] / 2 + centroid[1]

                buffer += [translate(pic, int(delta_x), int(delta_y))]
            return buffer

In [39]: test = np.array(centeredarray(X[:640]))

HBox(children=(IntProgress(value=0, max=640), HTML(value='')))

```

```

In [5]: dataL = np.load("hirag.npz")['arr_0']

In [9]: data = np.array(dataL[:1599])
        y = np.array(Y[:1599])

In [14]: datax = data.reshape(data.shape[0], data.shape[1]*data.shape[2])

In [12]: from sklearn.cluster import KMeans

In [25]: kmeans = KMeans(n_clusters=10, random_state=0).fit(datax)

In [17]: from sklearn.metrics.cluster import adjusted_rand_score

In [26]: kmLabel = kmeans.labels_

In [27]: adjusted_rand_score(y, kmLabel)

Out[27]: 0.3869665701454988

In [28]: from sklearn import metrics

In [31]: metrics.silhouette_score(datax, kmLabel, metric='euclidean')

Out[31]: -0.016696665638428632

In [32]: metrics.homogeneity_score(y, kmLabel)

Out[32]: 0.5605506711832837

In [33]: metrics.adjusted_mutual_info_score(y, kmLabel)

/home/hexagramg/exp/venv/lib/python3.6/site-packages/sklearn/metrics/cluster/supervised.py:732
FutureWarning)

Out[33]: 0.5555767757562253

In [36]: from sklearn.cluster import AgglomerativeClustering

In [41]: agc = AgglomerativeClustering(n_clusters=10).fit(datax)

In [42]: agLabels = agc.labels_

In [43]: adjusted_rand_score(y, agLabels)

Out[43]: 0.47397686046328175

In [44]: metrics.silhouette_score(datax, agLabels, metric='euclidean')

Out[44]: -0.02925792626053541

In [45]: metrics.homogeneity_score(y, agLabels)

```

```
Out [45]: 0.6467588865151238
```

```
In [46]: metrics.adjusted_mutual_info_score(y, agLabels)
```

```
/home/hexagramg/exp/venv/lib/python3.6/site-packages/sklearn/metrics/cluster/supervised.py:732  
FutureWarning)
```

```
Out [46]: 0.6427628965554346
```

```
In [47]: from sklearn.cluster import MeanShift
```

```
In [48]: clustering = MeanShift().fit(datax)
```

```
In [49]: mshLabels = clustering.labels_
```

```
In [50]: adjusted_rand_score(y,mshLabels)
```

```
Out [50]: 0.0020420433861530374
```

```
In [51]: metrics.silhouette_score(datax,mshLabels, metric='euclidean')
```

```
Out [51]: 0.0805389671008504
```

```
In [52]: metrics.homogeneity_score(y, mshLabels)
```

```
Out [52]: 0.10042434295351106
```

```
In [53]: metrics.adjusted_mutual_info_score(y, mshLabels)
```

```
/home/hexagramg/exp/venv/lib/python3.6/site-packages/sklearn/metrics/cluster/supervised.py:732  
FutureWarning)
```

```
Out [53]: 0.003904343710372022
```