

16-03-2023

COMPUTER VISION

- multicourse interdisciplinary
- not from scratch → Policy Perspective
- Assignment 3 - Topic Modelling
- Assignment 4 - ~~in class~~ Essay
includes Coding.

Def:

It is field in AI that derive meaningful information from the digital image.

- Computer vision task

- image recognition
 - Object detection
 - boundary image segmentation
 - Semantic segmentations.
 - Captionisation
- meaning from the images Geometrically, image Processing.
before ML.
- Tools → open CV & MATLAB.

Timeline

1970's

- digital image processing
- Pictorial structure
- Generalized cylinders
- optimal flows.

1980's {
→ Markov random fields.
→ texture & focus.
→ structure from motion.
→ Physically based Modelling.

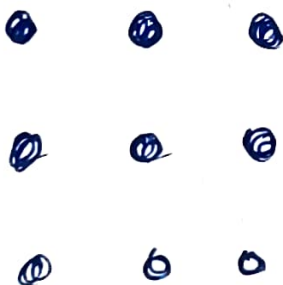
1990's {
→ graph cut
→ physics based vision
→ factorization

2000's {
→ inference algorithm
→ learning. (ML)

Standard task

→ image processing
→ feature extractions.
→ Segmentation
→ Captioning.

Edge detections → Edge is where pixel values suddenly changes.



0 -1 0

-1 4 -1

0 -1 0

→
sobel's filter

Canny Edge detection

$$\begin{bmatrix} -1 & 0 & 1 \\ -2 & 0 & 2 \\ -1 & 0 & 1 \end{bmatrix}$$

Horizontal

$$\begin{bmatrix} 1 & 2 & 1 \\ 0 & 0 & 0 \\ -1 & -2 & -1 \end{bmatrix}$$

Vertical

Resources

- google colab
- open CV
- Gimp, etc.

Things to do in coding in Next class.

- ① google colab use case
- ② open CV use case
- ③ MNIST Better.

zoom out — Easy
zoom in — Hard

Problem → In photograph find the tree edges.

Salt and pepper noise → Cause by sharp and sudden disturbance in image.

Simple computer vision problem MNSIT Dataset

```
In [2]: # To plot graoh
import matplotlib.pyplot as plt

# use svm
from sklearn import datasets,svm ,metrics

# to split dataset into train and test
from sklearn.model_selection import train_test_split
```

```
In [3]: # for confusion matrix and classification report
from sklearn.metrics import confusion_matrix
from sklearn.metrics import classification_report
```

```
In [4]: # load MNSIT() dataset
data = datasets.load_digits()
```

```
In [5]: # resize data matric
n_samples = len(data.images)
d = data.images.reshape((n_samples,-1))
```

```
In [6]: target = data.target
```

```
In [7]: # splitting into train and test
X_train,X_test,y_train,y_test = train_test_split(d,target,test_size=0.3)
```

```
In [8]: # train model
model = svm.SVC(gamma = 0.001)
model.fit(X_train,y_train)
```

```
Out[8]: ▼      SVC
        SVC(gamma=0.001)
```

```
In [9]: y_predict=model.predict(X_test)
```

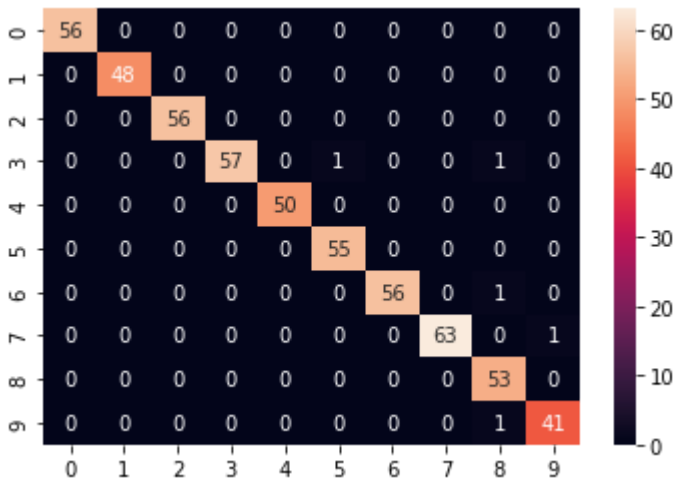
```
In [10]: # accuracy and F1 score
print(classification_report(y_test,y_predict))
```

	precision	recall	f1-score	support
0	1.00	1.00	1.00	56
1	1.00	1.00	1.00	48
2	1.00	1.00	1.00	56
3	1.00	0.97	0.98	59
4	1.00	1.00	1.00	50
5	0.98	1.00	0.99	55
6	1.00	0.98	0.99	57
7	1.00	0.98	0.99	64
8	0.95	1.00	0.97	53
9	0.98	0.98	0.98	42
accuracy			0.99	540
macro avg	0.99	0.99	0.99	540
weighted avg	0.99	0.99	0.99	540

```
In [11]: cf_matrix=confusion_matrix(y_test,y_predict)
```

```
In [12]: # for confusion matrix
import seaborn as sns
sns.heatmap(cf_matrix, annot=True)
```

Out[12]: <AxesSubplot:>



```
In [ ]:
```