Discrete Mathematics

Mid-term Report

Clustering

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- 簡介
- 應用
- 實做範例

簡介

Clustering (群聚分析)

- · Cluster 群聚: 一群 data objects
- - 在同一群內相當相似
 - 在不同群內非常不相似
- Cluster analysis
- · 把資料依相似性分群

- 簡介
- 應用
- 實做範例

• Marketing:

• Help marketers discover distinct groups in their customer bases, and then use this knowledge to develop targeted marketing programs

• Insurance:

· Identifying groups of motor insurance policy holders with a high average claim cost

· City Planning:

· Identifying groups of houses according to their house type, value, and geographical location

- 簡介
- 應用
- 實做範例

• 實作範例

- Reference material
- Clustering Problem (Training data)
- Implementation
- Algorithm : K-means

• Reference material

tf.argmin(input, dimension, name=None)

函數解說:沿著需要的維度找尋最小值的索引值,最小由0開始

tf.reduce sum()

reduce_sum()就是求和,由於求和的對象是tensor,所以是沿着tensor的某些维度求和。reduction_indices是指沿tensor的哪些维度求和。https://www.zhihu.com/question/51325408

tf.reshape(tensor,shape, name=None)

函数的作用是將tensor變換為參數shape的形式。

其中shape為一个列表形式,特殊的一點是列表中可以存在-1。-1代表的含義是不用我們自己指定這一维的大小,函数会自動计算,但列表中只能存在一個-1好了我想说的重点还有一个就是根据shape如何变换矩阵。其实简单的想就是,

reshape (t, shape) => reshape(t, [-1]) => reshape(t, shape)

首先将矩阵t变为一维矩阵,然后再对矩阵的形式更改就可以了。

http://blog.csdn.net/zeuseign/article/details/72742559

tile() 平鋪之意,用於在同一维度上的複製

http://blog.csdn.net/xwd18280820053/article/details/72867818

tf.reduce_any() 計算tensor中各個元素的邏輯或運算

http://blog.csdn.net/lhanchao/article/details/51442182

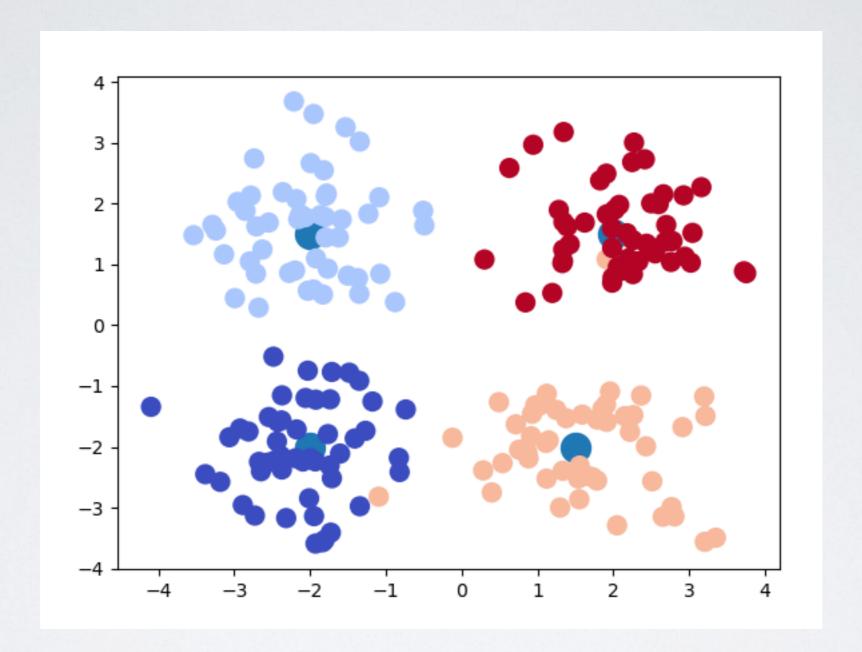
tf.unsorted segment sum(data, segment ids, num segments, name=None)

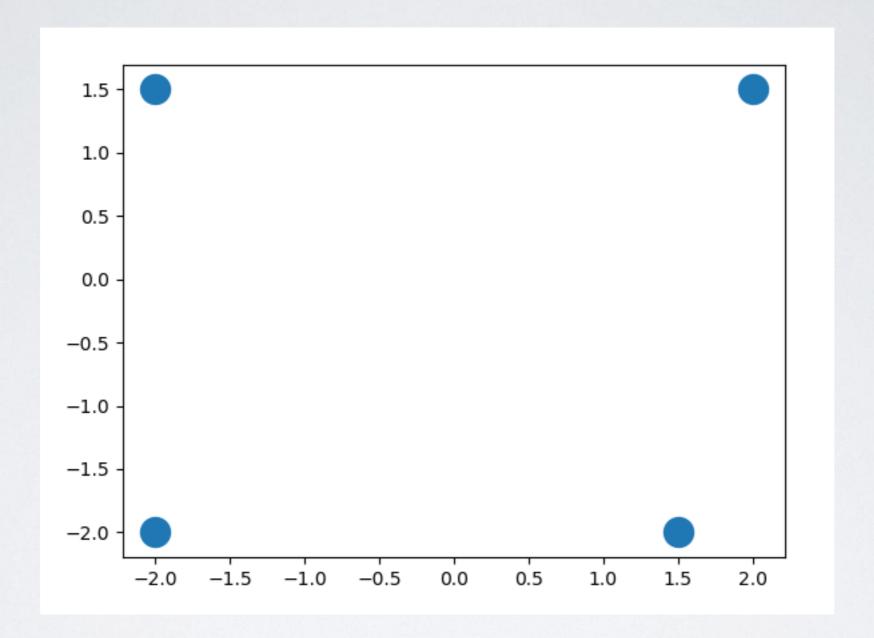
這個函數的作用是沿着segment ids指定的维度,分割張量data中的值,並且返回累加值。

計算公式為:

其中,segment_ids[j] == i。这个API和SegmentSum最大的區别是,這個API不需要從0到k有序排列,可以亂序排列,並且該API不需要包含從0到k。 如果對於给定的分割區間ID i,output[i] = 0。那麼,num_segmetns應該等於不同的段ID的數量。 http://www.jianshu.com/p/4daafdbcdddf

• Clustering Problem (Training data)





• Implementation

- Using Machine Learning (unsupervised learning)
- Framework: tensorflow

- Window 10
- Python 3.6
- GPU: GTX950

• Using Machine Learning (unsupervised learning)

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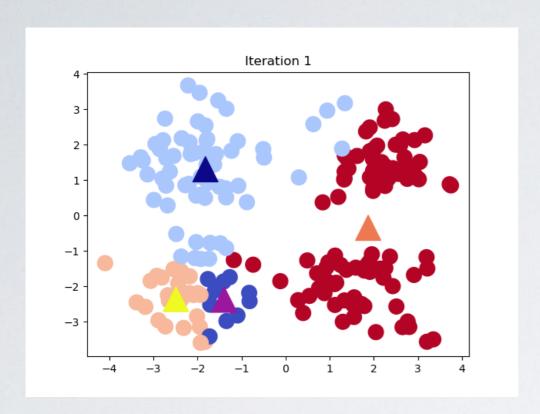
- Supervised Learning:
- Training the model with "Labeled" data, which is a common way in implementation
- Advantage: Higher precision
- Disadvantage: More cost

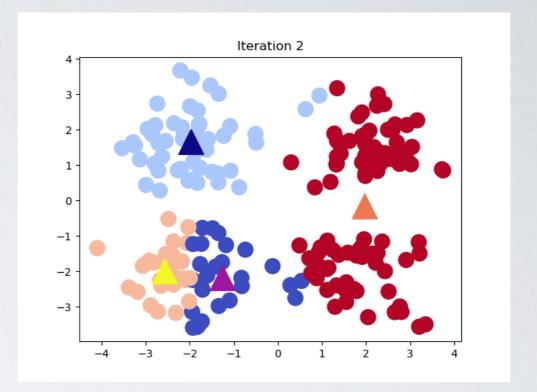
- Unsupervised Learning:
- Training with "Un-Labeled" data, common in clustering, but pure unsupervised learning is not common in implementation
- Advantage : Less cost

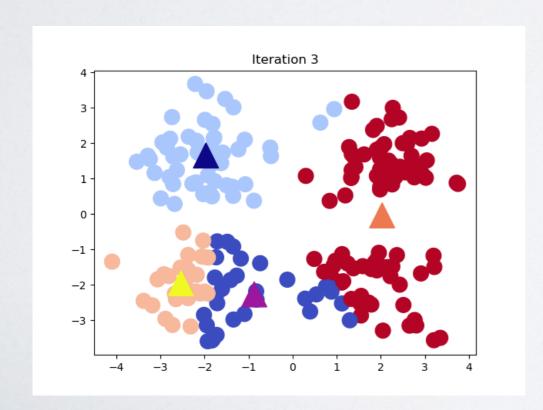
• Algorithm : K-means

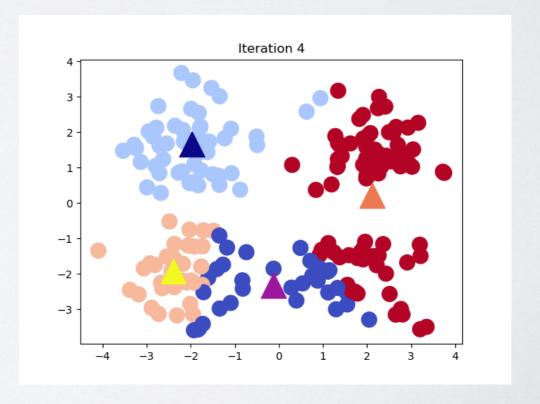
K-means Algorithm

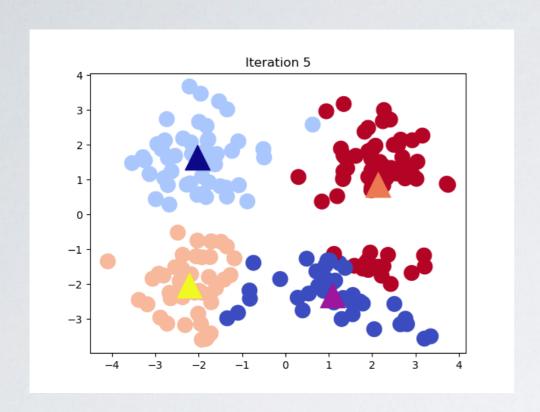
```
Input:
   Training set Training set: \{x^{(1)}, x^{(2)}, \dots, x^{(m)}\}
      x^{(i)} \in \mathbb{R}^n (drop x_0 = 1 convention, thus x^{(i)} 
otin \mathbb{R}^{n+1} )
   K: number of clusters
Randomly initialize K cluster centroids \mu_1, \mu_2, \ldots, \mu_K \in \mathbb{R}^n
Repeat {
  // cluster assignment step
   for i = 1 to m
     c^{(i)} := index (from 1 to K) of cluster centroid closest to x^{(i)}
      (或表示為 c^{(i)} := min_k ||x^{(i)} - \mu_k||^2)
   // move centroid step
   for k = 1 to K
      \mu_k := average (mean) of points assigned to cluster k
}
```

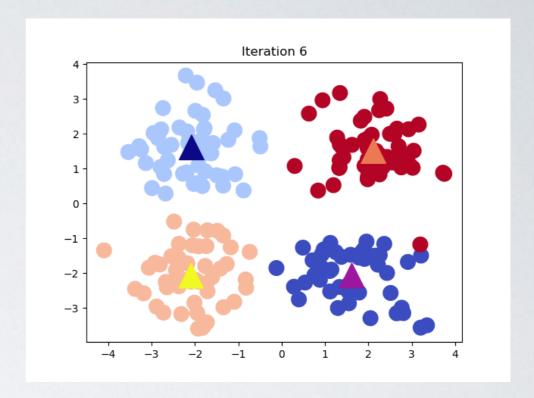


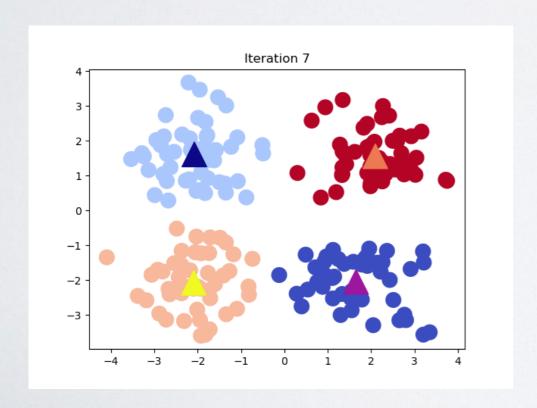


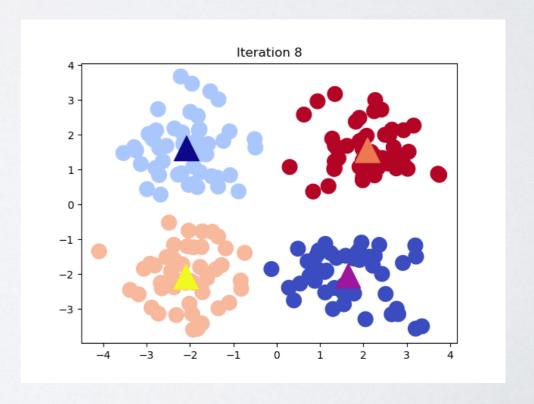






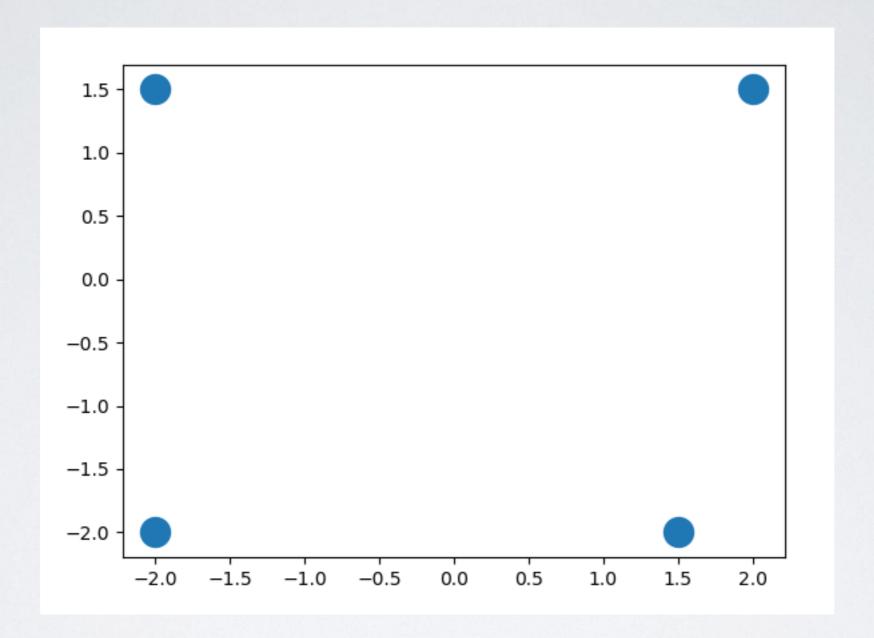






```
PS D:\Programing\machine learning> python .\k-means\k-means.py
2017-11-16 15:14:28.842545: I C:\tf jenkins\home\workspace\rel-win\M\windows-gpu\PY\36\tensorflow\core\platform\cpu feature guard.cc:137] Your CPU supports instructions that
t this TensorFlow binary was not compiled to use: AVX AVX2
2017-11-16 15:14:29.650957: I C:\tf jenkins\home\workspace\rel-win\M\windows-gpu\PY\36\tensorflow\core\common runtime\gpu\gpu device.cc:1030] Found device 0 with properties
name: GeForce GTX 950M major: 5 minor: 0 memoryClockRate(GHz): 1.124
pciBusID: 0000:01:00.0
totalMemory: 2.00GiB freeMemory: 1.65GiB
2017-11-16 15:14:29.651066: I C:\tf_jenkins\home\workspace\rel-win\M\windows-gpu\PY\36\tensorflow\core\common_runtime\gpu\gpu_device.cc:1120] Creating TensorFlow device (/d
evice:GPU:0) -> (device: 0, name: GeForce GTX 950M, pci bus id: 0000:01:00.0, compute capability: 5.0)
WARNING:tensorflow:From C:\Users\User\AppData\Local\Programs\Python\Python36\lib\site-packages\tensorflow\python\util\tf should use.py:107: initialize all variables (from t
ensorflow.python.ops.variables) is deprecated and will be removed after 2017-03-02.
Instructions for updating:
Use `tf.global_variables_initializer` instead.
Found in 20.09 seconds 8 iterations
Centroids:
[[ 1.65289262 -2.04643427]
[-2.0763623 1.61204964]
[-2.08862822 -2.07255306]
[ 2.09831502 1.55936014]]
3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3
```

PS D:\Programing\machine learning>



Source Code

Import all we need

import tensorflow as tf import numpy as np import time

#help us to graph
import matplotlib
import matplotlib.pyplot as plt

#import datasets we need by scikit-learn from sklearn.datasets.samples_generator import make_blobs from sklearn.datasets.samples_generator import make_circles

Set up Data set

#set up data type, here i choose blobs to make it simpler DATA_TYPE = "blobs"

#Set up Number of clusters in train data, if we choose circle,2 is enough

$$K = 4$$
if(DATA_TYPE == "circle"):
$$K = 2$$
else:
$$K = 4$$

Set up Data set

Using ski-learn for blobs:

n_samples: number of data, which means we have 200 points centers = [(-2, -2), (-2, 1.5), (1.5, -2), (2, 1.5)] centers = centers

n_features = dimmension, here we choose plane so = 2 cluster_std = std

shuffle: if we mix up samples, here I choose false random_state: random seed for circles:

noise: random noise data set up to the sample set factor: the ratio factor between circle data set

Conclusion

Using unsupervised learning takes advantage of Clustering.

As the result, the common way in implenmentation is

Clustering first before supervised learning which called "Semisupervised learning"

In this case, I choose an easy way to generate a data, and imply unsupervised learning which can help me to Label the data first, considered a pre-handling stage (different color represent different Lebel)