算法整体流程

将数据集提取 sift 特征,转化为 15 个特征集对每个特征集进行 kmeans 聚类根据聚类结果生成每张训练集的特征向量根据特征向量训练线性核 SVM 将测试集转换为特征向量,输入 SVM 得到分类结果

函数功能说明

def trainSet2featureSet():

对训练集中每张图片提取 sift 特征,分类别保存在 15 个文件内

def feature2vector(features, centers):

features 特征集 centers kmeans 聚类得到的中心集合 将特征集转换为特征向量

def learnVocabulary():

对 trainSet2featureSet()输出的特征集进行 kmeans 聚类,聚类结果保存在 15 个文件内

def trainSVM():

使用 feature2vector(features, centers)生成特征向量,根据特征向量训练线性 SVM, 训练结果保存在 svm. clf 中

def train():

整合所有训练步骤

def test():

读取 svm. clf,将训练集转换为特征向量输入 svm 得到分类结果,分类结果输出为混淆矩阵,并保存为文件

def check result():

读取并输出 test()保存的混淆矩阵文件

输入参数说明

#提取的 sift 特征数目
sift_num=200
#k
wordCnt = 50
#kmeans 结束精度要求
eps=0.1
#kmeans 最大迭代次数
max_iter=20
#kmeans 重复次数
re_kmeans=3

混淆矩阵 (参数如上)

[54, 1, 0, 2, 2, 0, 0, 0, 0, 0, 0, 0, 1, 1, 5] [0, 75, 5, 2, 1, 0, 0, 1, 2, 0, 0, 0, 1, 1, 3][6, 3, 121, 3, 9, 0, 0, 0, 3, 0, 4, 6, 3, 1, 2][2, 2, 1, 47, 2, 0, 0, 0, 0, 2, 0, 2, 1, 1, 0][2, 2, 5, 8, 95, 4, 0, 5, 5, 1, 1, 2, 2, 2, 5][0, 1, 1, 1, 0, 205, 0, 0, 0, 1, 0, 1, 0, 0][2, 0, 0, 1, 1, 0, 169, 0, 0, 0, 0, 0, 3, 1, 1][2, 1, 1, 4, 2, 2, 0, 89, 1, 1, 1, 3, 1, 1, 1] [2, 3, 4, 0, 7, 1, 0, 2, 135, 0, 0, 1, 3, 0, 0][1, 0, 0, 1, 0, 0, 2, 3, 1, 212, 1, 1, 1, 1, 0][15, 0, 3, 2, 9, 1, 0, 0, 1, 2, 223, 2, 1, 1, 0][5, 7, 1, 6, 16, 2, 1, 1, 1, 0, 0, 96, 2, 1, 3] [3, 5, 3, 9, 5, 0, 4, 8, 2, 1, 0, 3, 151, 1, 11][1, 0, 4, 5, 1, 0, 0, 2, 0, 0, 0, 0, 1, 51, 0][5, 6, 17, 2, 9, 0, 2, 6, 3, 0, 0, 6, 12, 2, 95]

性能影响分析

kmeans 环节耗时在训练中占比最大

re_kmeans 显著影响训练时间,取 20 时训练时间约 30min,取 3 时训练时间约 5min

以下是 re_kmeans=20, 其他参数不变时输出的混淆矩阵

[53, 1, 2, 2, 1, 0, 0, 2, 1, 0, 0, 0, 1, 3, 0]

[4, 77, 3, 0, 0, 0, 0, 0, 1, 0, 0, 2, 0, 4]

[1, 6, 128, 3, 4, 0, 0, 0, 5, 1, 1, 2, 0, 2, 8]

[1, 4, 2, 40, 6, 0, 0, 0, 0, 1, 0, 0, 1, 3, 2]

[5, 2, 6, 25, 74, 0, 0, 3, 1, 1, 5, 6, 2, 3, 6]

[3, 0, 0, 0, 0, 201, 0, 2, 0, 0, 0, 0, 1, 0, 3]

[0, 1, 1, 3, 0, 0, 168, 0, 2, 0, 0, 0, 0, 0, 3]

[0, 1, 1, 5, 2, 6, 0, 90, 0, 2, 0, 0, 1, 1, 1]

[2, 3, 6, 4, 4, 1, 1, 0, 127, 0, 1, 2, 0, 2, 5]

[1, 0, 5, 1, 3, 0, 0, 2, 0, 203, 0, 1, 5, 1, 2]

[1, 0, 5, 10, 5, 1, 0, 0, 0, 0, 231, 3, 2, 0, 2]

[1, 0, 3, 6, 7, 0, 0, 3, 2, 0, 0, 113, 0, 2, 5]

[18, 12, 8, 13, 5, 2, 3, 2, 6, 6, 4, 5, 107, 12, 3]

[0, 1, 2, 3, 4, 1, 0, 0, 0, 0, 0, 1, 1, 51, 1]

[2, 6, 20, 4, 6, 1, 3, 1, 0, 3, 2, 7, 5, 6, 99]

值得注意的是, kmeans 重复次数的增加对某些分类的分类准确度产生了负面影响。

max_iter 也会显著影响训练时间,取 30 时训练时间约 10min,取 20 时训练时间约 5min

以下是 max iter=30 时的混淆矩阵

[50, 2, 4, 2, 4, 0, 0, 1, 1, 0, 0, 2, 0, 0, 0]

[2, 77, 1, 1, 1, 0, 1, 0, 0, 0, 0, 0, 0, 7, 1]

[8, 1, 134, 5, 1, 1, 0, 2, 2, 3, 0, 0, 1, 2, 1]

[6, 3, 2, 38, 2, 0, 1, 1, 0, 1, 0, 1, 0, 4, 1]

[7, 4, 7, 5, 84, 4, 0, 0, 7, 0, 4, 3, 6, 5, 3]

[5, 0, 0, 0, 0, 204, 0, 0, 0, 0, 0, 0, 0, 0, 1]

[0, 0, 0, 0, 1, 0, 174, 0, 1, 0, 0, 0, 1, 0, 1]

[3, 3, 0, 4, 4, 2, 0, 88, 2, 0, 0, 1, 1, 1, 1]

[2, 1, 5, 0, 8, 1, 0, 1, 135, 0, 1, 2, 2, 0, 0]

[0, 1, 2, 0, 1, 1, 0, 0, 1, 214, 1, 0, 3, 0, 0]

[5, 1, 0, 2, 9, 1, 0, 4, 3, 1, 225, 2, 4, 1, 2]

[16, 12, 8, 8, 15, 0, 1, 1, 2, 0, 3, 67, 5, 0, 4]

[10, 5, 8, 6, 18, 2, 1, 5, 6, 2, 3, 8, 115, 4, 13]

[1, 4, 0, 6, 1, 0, 0, 1, 0, 0, 0, 0, 3, 47, 2]

[15, 4, 8, 3, 4, 0, 3, 1, 12, 0, 1, 6, 9, 3, 96]

调参结果和 re_k means 增加的效果很相似,对某些类别的分类结果产生了负面影响

wordCnt 影响训练时间的同时比较明确的的影响分类准确率以下是 wordCnt=30 时的混淆矩阵

[44, 2, 2, 1, 3, 2, 0, 3, 0, 1, 0, 2, 2, 0, 4][2, 60, 5, 0, 1, 2, 0, 2, 5, 2, 0, 6, 2, 2, 2][14, 5, 81, 26, 5, 0, 0, 8, 5, 6, 0, 6, 3, 1, 1][4, 0, 1, 47, 1, 0, 0, 0, 1, 1, 0, 1, 2, 0, 2][30, 3, 17, 4, 65, 1, 1, 0, 2, 1, 0, 9, 0, 2, 4][1, 2, 1, 0, 0, 199, 0, 2, 0, 1, 0, 1, 1, 2, 0][1, 0, 0, 1, 1, 0, 164, 0, 0, 0, 0, 0, 1, 2, 8][6, 1, 6, 4, 5, 1, 0, 82, 0, 2, 1, 0, 1, 0, 1][2, 6, 0, 1, 2, 1, 0, 4, 138, 0, 1, 0, 0, 0, 3][3, 2, 24, 0, 5, 3, 1, 13, 9, 155, 2, 3, 3, 0, 1][1, 4, 5, 0, 3, 10, 1, 1, 1, 0, 224, 1, 3, 1, 5][10, 14, 6, 0, 23, 0, 0, 0, 1, 0, 0, 77, 4, 6, 1][24, 16, 34, 13, 8, 7, 6, 7, 5, 5, 6, 8, 59, 4, 4][1, 0, 0, 4, 2, 4, 0, 0, 1, 0, 2, 1, 1, 48, 1][11, 0, 0, 7, 17, 2, 1, 8, 12, 4, 9, 3, 2, 3, 86]wordCnt 下降时可以明显观察到大部分类别分类准确率的下降

sift_num 显著影响分类准确率

以下是 sift_num=100 产生的混淆矩阵

[35, 0, 0, 12, 5, 0, 0, 1, 2, 1, 2, 1, 5, 2, 0]

[3, 66, 2, 1, 4, 0, 0, 1, 2, 0, 1, 7, 2, 0, 2]

[6, 5, 111, 10, 5, 0, 0, 2, 2, 0, 0, 2, 5, 2, 11]

[9, 1, 0, 32, 2, 1, 0, 0, 4, 1, 1, 2, 3, 2, 2]

[15, 9, 4, 14, 71, 1, 2, 2, 2, 1, 2, 5, 7, 2, 2]

[5, 1, 2, 3, 5, 188, 1, 1, 0, 0, 0, 0, 1, 2, 1]

[3, 2, 1, 1, 12, 0, 154, 1, 3, 0, 0, 1, 0, 0, 0]

[3, 1, 1, 1, 3, 0, 0, 90, 0, 1, 1, 5, 0, 2, 2]

[1, 7, 2, 12, 15, 3, 1, 3, 87, 3, 0, 7, 7, 7, 3]

[1, 0, 2, 4, 4, 0, 0, 3, 2, 202, 0, 0, 2, 3, 1]

[4, 1, 9, 8, 13, 1, 1, 2, 0, 1, 203, 5, 5, 3, 4]

[1, 4, 1, 10, 4, 1, 2, 2, 7, 2, 4, 93, 3, 3, 5]

[11, 7, 14, 11, 12, 6, 4, 3, 7, 8, 3, 7, 97, 7, 9]

[3, 3, 1, 3, 2, 3, 0, 0, 3, 2, 1, 1, 6, 36, 1]

[5, 1, 1, 11, 14, 0, 0, 4, 3, 4, 13, 12, 12, 2, 83]

明显观察到各类别分类准确率大幅下降