## fisheriris

```
1 % 加载数据集
 2
   clear; clc;
 3
   load fisheriris;
 4
   X = meas:
 5
   y = species;
 6
 7
   % 将类别标签转换为数值型
 8
   y = grp2idx(y);
 9
   % 设置重复次数
10
11
   numRepeats = 10;
12
13 % 初始化精度数组
14
   linearAccuracies = zeros(numRepeats, 1);
   rbfAccuracies = zeros(numRepeats, 1);
15
   bpAccuracies = zeros(numRepeats, 1);
16
17
18
   % 进行多次训练和测试
   for i = 1:numRepeats
19
       % 分割数据为训练集和测试集(70%训练,30%测试)
20
       cv = cvpartition(y, 'HoldOut', 0.3);
21
22
       X_train = X(training(cv), :);
23
       y_train = y(training(cv));
       X_{\text{test}} = X(\text{test}(cv), :);
24
25
       y_test = y(test(cv));
26
27
       % 标准化特征
28
       X_train = zscore(X_train);
29
       X_test = zscore(X_test);
30
31
       % 训练线性核SVM
32
       linearSVMModel = fitcecoc(X_train, y_train, 'Learners',
   templateSVM('KernelFunction', 'linear'));
33
34
       % 训练高斯核SVM
35
       rbfSVMModel = fitcecoc(X_train, y_train, 'Learners',
   templateSVM('KernelFunction', 'rbf'));
36
37
       % 训练BP神经网络
       bpModel = feedforwardnet([10, 10]); % 增加隐藏层数量,两个隐藏层,每层10个神
38
   经元
39
       bpModel.trainParam.epochs = 200; % 增加训练迭代次数
40
       bpModel.trainParam.lr = 1; % 设置学习率
41
       bpModel = train(bpModel, X_train', dummyvar(y_train)'); %利用dummyvar变y
   回one hot类型
42
43
       % 预测线性核SVM
```

```
linearPredictions = predict(linearSVMModel, X_test);
44
45
       linearAccuracies(i) = mean(linearPredictions = y_test);
46
       % 预测高斯核SVM
47
48
       rbfPredictions = predict(rbfSVMModel, X_test);
49
       rbfAccuracies(i) = mean(rbfPredictions = y_test);
50
       % 预测BP神经网络
51
52
       bpPredictions = bpModel(X_test');
       [~, bpPredictions] = max(bpPredictions, [], 1); % 将输出的概率转化为类别标签
53
       bpAccuracies(i) = mean(bpPredictions' = y_test);
54
55
   end
56
57
   % 计算平均精度
58
   meanLinearAccuracy = mean(linearAccuracies);
59
   meanRbfAccuracy = mean(rbfAccuracies);
   meanBpAccuracy = mean(bpAccuracies);
60
61
62
   % 输出精度
   fprintf('线性核SVM的平均精度: %.2f%%\n', meanLinearAccuracy * 100);
63
64
   fprintf('高斯核SVM的平均精度: %.2f%%\n', meanRbfAccuracy * 100);
   fprintf('BP神经网络的平均精度: %.2f%%\n', meanBpAccuracy * 100);
65
66
67
   % 创建表格数据
   Iteration = (1:numRepeats)';
68
69
   LinearAccuracy = linearAccuracies * 100;
70
   RBFAccuracy = rbfAccuracies * 100;
71
   BPAccuracy = bpAccuracies * 100;
72
73
   % 创建表格并添加平均精度行
74
   resultsTable = table(Iteration, LinearAccuracy, RBFAccuracy, BPAccuracy);
75
   averageRow = table(numRepeats+1, meanLinearAccuracy * 100, meanRbfAccuracy *
   100, meanBpAccuracy * 100, 'VariableNames', {'Iteration', 'LinearAccuracy',
   'RBFAccuracy', 'BPAccuracy'});
76
   resultsTable = [resultsTable; averageRow];
77
78 % 修改表格行名称
   resultsTable.Properties.RowNames = [cellstr(num2str((1:numRepeats)'));
79
   'Average'];
80
81 % 显示表格
82 disp(resultsTable);
```

线性核SVM的平均精度: 97.78% 高斯核SVM的平均精度: 95.78% BP神经网络的平均精度: 95.33% Iteration LinearAccuracy RBFAccuracy **BPAccuracy** 97.778 1 100 91.111 1 2 2 100 97.778 97.778 3 3 100 95.556 97.778 4 95.556 95.556 91.111 5 5 95.556 93.333 91.111 6 6 97.778 93.333 93.333 7 7 100 100 100 8 8 95.556 95.556 97.778 9 9 93.333 93.333 91.111 10 100 100 97.778 Average 11 97.778 95.778 95.333

## Diabetes

```
1 % 加载 Diabetes 数据集
   clear; clc;
 3
   load diabetes;
 4
   X = Feature;
   y = Class;
 5
 6
 7
   % 将 one-hot 编码转换为类别标签
   [\sim, y] = \max(y, [], 2);
 8
 9
   % 设置重复次数
10
11
   numRepeats = 10;
12
   % 初始化精度数组
13
14
   linearAccuracies = zeros(numRepeats, 1);
   rbfAccuracies = zeros(numRepeats, 1);
15
   bpAccuracies = zeros(numRepeats, 1);
16
17
   % 进行多次训练和测试
18
   for i = 1:numRepeats
19
20
       % 分割数据为训练集和测试集(70%训练,30%测试)
       cv = cvpartition(y, 'HoldOut', 0.3);
21
22
       X_train = X(training(cv), :);
23
       y_train = y(training(cv));
24
       X_{\text{test}} = X(\text{test}(cv), :);
25
       y_test = y(test(cv));
26
27
       % 标准化特征
28
       X_train = zscore(X_train);
29
       X_test = zscore(X_test);
30
31
       % 训练线性核SVM
```

```
32
       linearSVMModel = fitcecoc(X_train, y_train, 'Learners',
   templateSVM('KernelFunction', 'linear'));
33
34
       % 训练高斯核SVM
35
       rbfSVMModel = fitcecoc(X_train, y_train, 'Learners',
   templateSVM('KernelFunction', 'rbf'));
36
37
38
       % 训练BP神经网络
       bpModel = feedforwardnet([10, 10]); % 增加隐藏层数量,两个隐藏层,每层10个神
39
   经元
       bpModel.trainParam.epochs = 200; % 增加训练迭代次数
40
41
       bpModel.trainParam.lr = 1; % 设置学习率
42
       bpModel = train(bpModel, X_train', dummyvar(y_train)'); %利用dummyvar变y
   回one hot类型
43
       % 预测线性核SVM
44
45
       linearPredictions = predict(linearSVMModel, X_test);
46
       linearAccuracies(i) = mean(linearPredictions = y_test);
47
48
       % 预测高斯核SVM
49
       rbfPredictions = predict(rbfSVMModel, X_test);
50
       rbfAccuracies(i) = mean(rbfPredictions = y_test);
51
52
       % 预测BP神经网络
53
       bpPredictions = bpModel(X_test');
       [~, bpPredictions] = max(bpPredictions, [], 1); % 将输出的概率转化为类别标签
54
       bpAccuracies(i) = mean(bpPredictions' = y_test);
55
   end
56
57
   % 计算平均精度
58
59
   meanLinearAccuracy = mean(linearAccuracies);
60
   meanRbfAccuracy = mean(rbfAccuracies);
61
   meanBpAccuracy = mean(bpAccuracies);
62
63
   % 输出精度
   fprintf('线性核SVM的平均精度: %.2f%\n', meanLinearAccuracy * 100);
64
   fprintf('高斯核SVM的平均精度: %.2f%%\n', meanRbfAccuracy * 100);
65
   fprintf('BP神经网络的平均精度: %.2f%%\n', meanBpAccuracy * 100);
66
67
   % 创建表格数据
68
69
   Iteration = (1:numRepeats)';
70
   LinearAccuracy = linearAccuracies * 100;
71
   RBFAccuracy = rbfAccuracies * 100;
72
   BPAccuracy = bpAccuracies * 100;
73
74
   % 创建表格并添加平均精度行
   resultsTable = table(Iteration, LinearAccuracy, RBFAccuracy, BPAccuracy);
75
   averageRow = table(numRepeats+1, meanLinearAccuracy * 100, meanRbfAccuracy *
76
   100, meanBpAccuracy * 100, 'VariableNames', {'Iteration', 'LinearAccuracy',
   'RBFAccuracy', 'BPAccuracy'});
```

线性核SVM的平均精度: 76.61% 高斯核SVM的平均精度: 70.00% BP神经网络的平均精度: 74.52%

	Iteration	LinearAccuracy	RBFAccuracy	BPAccuracy
1	1	74.783	66.087	72.609
2	2	77.391	70.87	76.522
3	3	76.087	66.522	72.609
4	4	76.087	67.391	70
5	5	75.217	70.87	76.522
6	6	77.391	71.304	74.783
7	7	75.217	73.478	73.913
8	8	78.261	73.913	78.696
9	9	78.261	68.696	76.087
10	10	77.391	70.87	73.478
Average	11	76.609	70	74.522