#### 算法 1 特征抽取

15: end function

输入: data 时间序列对应的数据, label 时间序列对应的标签, window\_len 窗口长度 输出:时间序列抽取的特征矩阵和标签 1: **function** EXTRACT\_FEATURES(data, label, window\_len)  $features \leftarrow 0$ 2:  $targets \leftarrow 0$ 3:  $i \leftarrow 0$ 4:  $num\_windows \leftarrow len(data)/(window\_len/2)$ 5: for  $i = 0 \rightarrow num\_windows$  do 6:  $target \leftarrow int(label[i \rightarrow i + window \ len].mode())$ 7:  $targets \leftarrow targets \cup target$ 8: for  $c = 0 \rightarrow data.columns$  do 9:  $features \leftarrow features \cup FEATURIZE(data[i \rightarrow i + window\_len])$ 10: end for 11:  $i \leftarrow i + window\_len/2$ 12: end for 13: 14: return features, targets

$$Precision = \frac{TP}{(TP + FP)}$$
 
$$Recall = \frac{TP}{(TP + FN)}$$
 
$$F1 - Score = \frac{2 * Precision * Recall}{(Recall + Precision)}$$
 
$$Accuracy = \frac{TP}{(TP + FP + FN + TN)}$$

#### 算法 2 FEATURIZE

```
1: rms\_val = \sqrt{\frac{1}{n}\sum_{i=1}^{n}x_{i}^{2}}

2: min\_max\_mean = \frac{1}{len}\sum_{i=1}^{len}|max_{i} - min_{i}| 其中 len = \min(max, min) ; min 和 max 为序列中的 极大值和极小值点

3: peak = \max max - \min min

4: peaknum = len(max) + len(min)

5: mean = \frac{1}{n}\sum_{i=1}^{n}x_{i}

6: standarddeviation = \sqrt{\frac{1}{n}\sum_{i=1}^{n}x_{i} - mean}

7: coefficientsofvariation = \frac{standarddeviation}{mean}

8: Skewness = \frac{\frac{1}{n}\sum_{i=1}^{n}(x_{i} - mean)^{2}}{(\frac{1}{n}\sum_{i=1}^{n}(x_{i} - mean)^{2})^{\frac{3}{2}}}

9: Kurtosis = \frac{\frac{1}{n}\sum_{i=1}^{n}(x_{i} - mean)^{2}}{(\frac{1}{n}\sum_{i=1}^{n}(x_{i} - mean)^{2})^{3}} - 3

10: log - energy = \frac{1}{n}\sum_{i=1}^{n}log(x_{i}^{2})
```

#### 算法 3 单一活动类别识别

```
1: function Task1
        features \leftarrow 0
 2:
        targets \leftarrow 0
3:
        for i = 0 \rightarrow 12 do
4:
            data \leftarrow data \cup PREPROCESSING(i)
 5:
            for i = 1 \rightarrow 7 do
 6:
                d \leftarrow data[label = i]
 7:
                 feature, target = \text{EXTRACT\_FEATURES}(d[x, y...], d[label], 520)
                 features \leftarrow features \cup feature
9:
                targets \leftarrow targets \cup target
10:
            end for
11:
        end for
12:
        classifiers = Train(features, targets)
13:
        for i = 13 \to 15 do
14:
            data \leftarrow data \cup PREPROCESSING(i)
15:
            for i = 1 \rightarrow 7 do
16:
                 feature, target = \text{EXTRACT\_FEATURES}(data[label = i][x, y...], data[label = i][label], 520)
17:
                 Test (classifiers, feature, target)
18:
            end for
19:
        end for
20:
21: end function
```

### 算法 4 多活动类别识别

```
1: function Task3
        features \leftarrow 0
 2:
3:
        targets \leftarrow 0
        for i = 0 \rightarrow 12 do
 4:
            data \leftarrow data \cup PREPROCESSING(i)
5:
            feature, target = \text{EXTRACT} \quad \text{FEATURES}(data[x, y...], data[label], 3)
 6:
            features \leftarrow features \cup feature
 7:
            targets \leftarrow targets \cup target
 8:
        end for
9:
        classifiers = Train(features, targets)
10:
        for i = 13 \rightarrow 15 do
11:
            data \leftarrow data \cup PREPROCESSING(i)
12:
             feature, target = \text{EXTRACT\_FEATURES}(data[x, y...], data[label], 3)
13:
            Test'(classifiers, feature, target)
14:
        end for
15:
16: end function
```

#### 算法 5 多活动类别识别

```
1: function Task3
        features \leftarrow 0
2:
        targets \leftarrow 0
3:
        for i = 0 \rightarrow 12 do
 4:
            data \leftarrow data \cup PREPROCESSING(i)
 5:
            feature, target = \text{EXTRACT\_FEATURES}(data[x, y...], data[label], 3)
 6:
            features \leftarrow features \cup feature
 7:
 8:
            targets \leftarrow targets \cup target
9:
        classifiers = Train(features, targets)
10:
        for i = 13 \rightarrow 15 do
11:
            data \leftarrow data \cup PREPROCESSING(i)
12:
            feature, target = \text{EXTRACT\_FEATURES}(data[x, y...], data[label], 3)
13:
            Test'(classifiers, feature, target)
14:
        end for
15:
16: end function
```

# 算法 6 梯度下降法

```
输入: data 数据矩阵, label 标签
```

## 输出: 权重

- 1: **function** GRADASCENT(data, label)
- 2:  $alpha \leftarrow 0.001$
- 3:  $maxCycles \leftarrow 500$
- 4:  $weights \leftarrow ones((n,1))$
- 5: **for**  $k = 0 \rightarrow maxCycles$  **do**
- 6:  $h \leftarrow SIGMOID(data * weights)$
- 7:  $error \leftarrow (label h)$
- 8:  $weights \leftarrow weights + alpha * data * error$
- 9: end for
- 10: **return** weights
- 11: end function

#### 算法 7 决策树

**输入:** data 数据矩阵, Target\_attribute 要预测的目标属性, Attributes 除目标属性外供学习到的决策树测试属性列表

输出: root 一颗能正确分类给定 data 的决策树

- 1: **function** DecisionTreeClassifier(data, label)
- 2: 创建 root 节点
- 3: **if** 所有 data 都为正 **then**
- 4: **return** label = + 的单节点树 root
- 5: end if
- 6: **if** 所有 data 都为负 **then**
- 7: **return** label = -的单节点树 root
- 8: end if
- 9: **if** attributes 为空 **then**
- 10: **return** 单节点树 root, label = data 中最普遍的 Target\_attributes
- 11: end if
- 12:  $A \leftarrow attributes$  中分类 data 能力最好的属性
- 13: root 的决策属性  $\leftarrow A$
- 14: **for**  $i \rightarrow len(A)$  **do**
- 15:  $a \leftarrow A[i]$
- 16: 在 root 下加一个新的分支对应测试 A=a
- 17: 令 data(a) 为 data 中满足 A 属性值为 a 的子集
- 18: 在这个新分支下一个子树 DECISIONTREECLASSIFIER(data(a), Target\_attribute, Attributes A)
- 19: end for
- 20: **return** root
- 21: end function