# 基于Swift语 言的iOS应用 开发

时间的信息: 时序模型

# 本节概要

- 前情回顾: 单一时刻的识别
  - 目标检测 (Object Detection)
  - 语义分割 (Semantic Segmentation)
- 本节内容: 多时次的识别

### 何为多时次识别?

基于最原始的数据直接进行表示学习



饼干去哪里了? 故事书、连环画剧情... 需要依赖前几帧图像来判断后面的内容

### 一些需考虑时序的任务

- 视频内容识别
- 语音识别:语音转文字,歌曲转曲谱
- 文本语义识别
- 其他: 气象相关、股市相关等等

# 本堂课关注的内容:人体行为识别

- 利用传感器数据进行用户的行为识别
  - 和iOS相关密切: 每日步数、Apple Watch的环...
- 如何收集传感器数据
- 利用TuriCreate训练人体行为识别模型
- 利用训练好的模型,识别用户动作

# 目标场景: 节奏达人游戏

- App按节奏给出用户需完成的姿势提示,用户手持设备完成姿势
  - 挥砍
  - 抖动
  - 。 转动方向盘
- 可能所需的传感器
  - 加速度、陀螺仪、计步器、磁力计、气压高度计、G可能PS等...

# 构造数据集

如何构造? 既然用传感器数据做识别, 那自然需要通过传感器采集数据。

- 开发一个传感器数据采集的app
  - 收集、存储用户做动作时的传感器数据
  - 。 手动的为不同的动作打上标签

### **Gesture Data Recorder**

### App功能

- 支持输入用户id
- 选择需录制的动作
- 选择录制的片段数
- 开始录制

基础功能我们都提供了.. 需要大家完成: 动作采集(Core Motion) 相关代码



# 获取传感器数据的权限

通过Core Motion获取iPhone/iPod的传感器数据,在ViewController.swift内添加

import CoreMotion

还需在Info.plist中添加相关内容,允许App使用某些传感器详见Privacy - Motion Usage Description

创建获取传感器数据的对象CMMotionManager 创建Queue,通过异步方式获取传感器数据

```
let motionManager = CMMotionManager()
let queue = OperationQueue()
```

#### 设置每秒获取数据的数量

```
static let samplesPerSecond = 25.0
```

创建数据存储的变量

```
var activityData: [String] = []
```

#### 存储数据到文件

```
do {
try self.activityData.appendLinesToURL(fileURL: dataURL)
print("Data appended to \(dataURL)")
} catch {
print("Error appending data: \(error)")
}
```

```
let activity = isRecording ? currendActivity : .none
let sample = """
activityData.append(sample)
```

```
motionManager.deviceMotionUpdateInterval = Config.samplesPerSecond
activityData = []
motionManager.startDeviceMotionUpdates(
using: .xArbitraryZVertical,
to: queue,
withHandler: { [weak self] motionData, error in
quard let self = self, let motionData = motionData else {
let errorText = error?.localizedDescription ?? "Unknown"
print("Device motion update error: \((errorText)\)")
self.process(data: motionData)
```

# 开启、关闭session

```
case Utterances.sessionStart:
// TODO: enable Core Motion
enableMotionUpdates()
queueNextActivity()
```

```
func disableMotionUpdates() {
motionManager.stopDeviceMotionUpdates()
}
```

```
case Utterances.sessionComplete:
disableMotionUpdates()
```

# 构造数据集

通过App获取用户行为的传感器数据

- 数据存放在手机,可通过Files等app打开
  - Info.plist设置
    - Application supports iTunes file sharing: YES
    - Supports opening documents inplace: YES

手动构造训练集、验证集、测试集

# 数据分析和处理

```
%matplotlib inline
import turicreate as tc
import activity_detector_utils as utils
```

#### 读取数据集

```
train_sf = utils.sframe_from_folder("data/train")
valid_sf = utils.sframe_from_folder("data/valid")
test_sf = utils.sframe_from_folder("data/test")
```

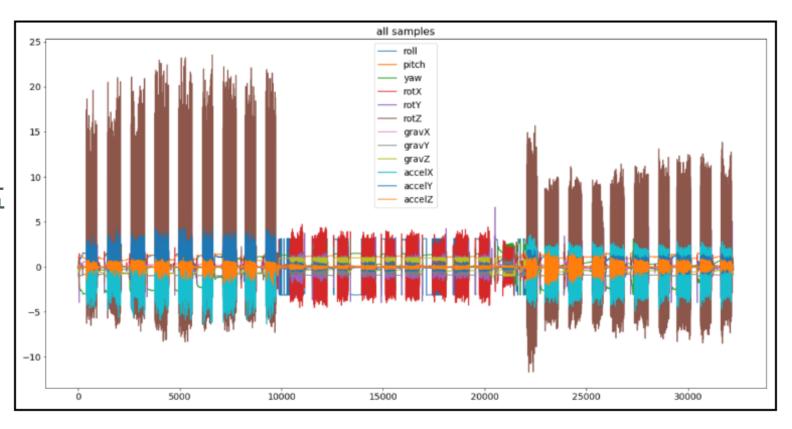
# 分析数据

总要看看数据集好不好

- 因此是手工标记,标记是否存在错误?
- 传感器放置错误,没有严格按照要求采集数据等
- 传感器数据有误,出现坏数据
- 三种行为以外的数据

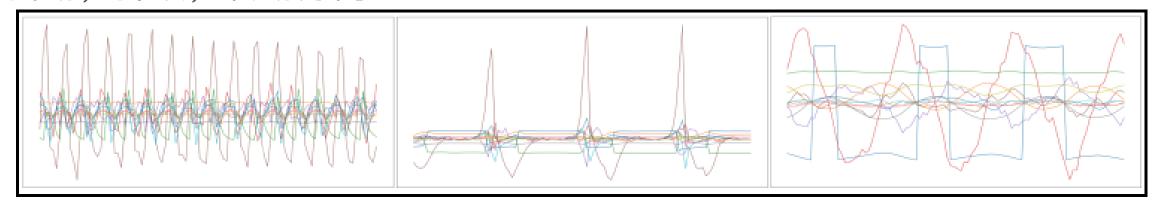
# 数据可视化

抖动 挥砍 转动方向盘



# 数据可视化

抖动,挥砍,转动方向盘



# 平衡类别

#### 按类别统计数量

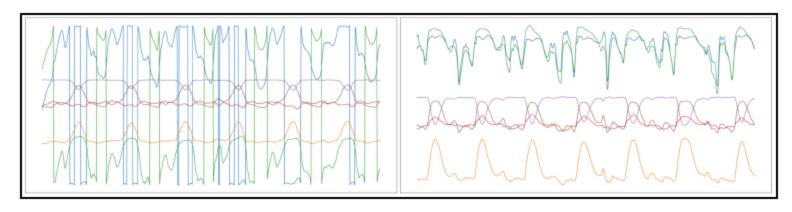
```
utils.count_activities(train_sf)
utils.count_activities(valid_sf)
utils.count_activities(test_sf)
```

++								
activity	userId	Count	activity	userId	Count	activity	userId	Count
+	++	+·	+	+	+	+ +	+	++
chop_it	u_01	36	chop_it	u_03	4	chop_it	u_05	9
chop_it	u_02	36	chop_it	u_04	4	drive_it	u_05	9
drive_it	u_01	36	drive_it	u_03	4	rest_it	u_05	27
drive_it	u_02	36	drive_it	u_04	4	shake_it	u_05	9
rest_it	u_01	108	rest_it	u_03	12	++		
rest_it	u_02	108	rest_it	u_04	12	[4 rows x 3	columns]	
shake_it	u_01	36	shake_it	u_03	4			
shake_it	u_02	36	shake_it	u_04	4			
++								
[8 rows x 3 columns] [8 rows x 3 columns]								

# 构造模型

模型的输入数据

• 一次采样? 二次采用? 四次? 八次?



不同用户在同一行为下的数据大相径庭、需根据不同用户做不同的切分

# 构造模型

#### 利用Ture Create创建模型

```
model = tc.activity_classifier.create(
dataset=train_sf, session_id='sessionId', target='activity',
features=[
"rotX", "rotY", "rotZ", "accelX", "accelY", "accelZ"],
prediction_window=20, validation_set=valid_sf,
max_iterations=20)
```

# 模型训练结果

```
0.985
  50
                                            0.043
                                                                   0.952
Training complete
Total Time Spent: 29.5252s
▶ ▶\  MI
   model.summary()
Class
                                        : ActivityClassifier
Schema
Number of examples
                                        : 235057
Number of sessions
                                        : 216
Number of classes
                                        : 4
Number of feature columns
                                        : 6
Prediction window
                                        : 20
Training summary
Log-likelihood
                                        : 0.0434
Training time (sec)
                                        : 30.0827
```

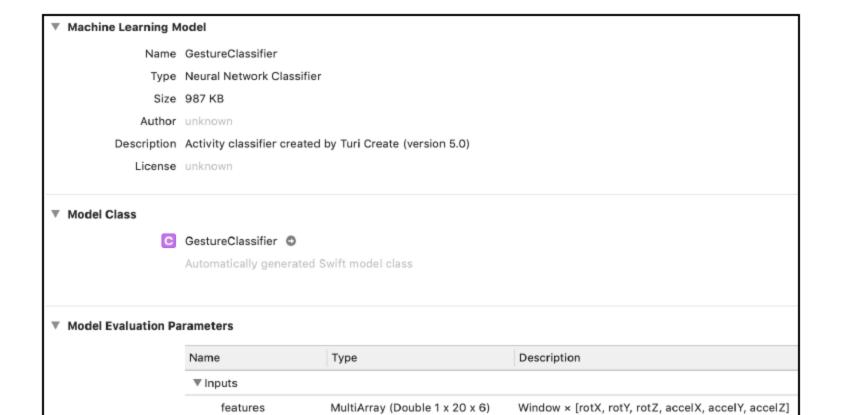
# 验证模型

```
metrics = model.evaluate(test_sf)
print(metrics)
```

```
('accuracy': 0.9734649489710557, 'auc': 0.9966916821874753, 'precision': 0.976516169734427, 'recall': 0.968202640497
4962, 'f1 score': 0.9720395581549871, 'log loss': 0.12406052127540747, 'confusion matrix': Columns:
       target label
       predicted label str
       count int
Rows: 11
Data:
 target_label | predicted_label | count
   shake_it
                    rest_it
                                    96
                    drive it
   rest it
                                   17
   drive_it
                    drive it
                                   5443
   chop_it
                    rest_it
                                   181
   chop_it
                    shake it
                                   176
   rest_it
                    rest_it
                                  13164
   chop_it
                    chop_it
                                   5118
   drive it
                    rest it
                                   184
   shake it
                    shake_it
                                   5367
                    chop_it
   rest_it
                                    62
```

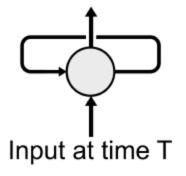
# 保存模型

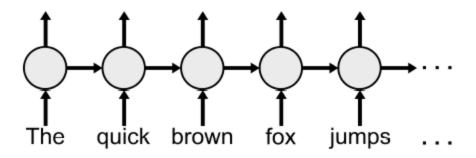
model.export\_coreml("GestureClassifier.mlmodel")
model.save("GestureClassifier")



# 时序模型概览

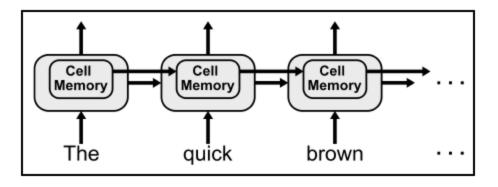
RNN模型





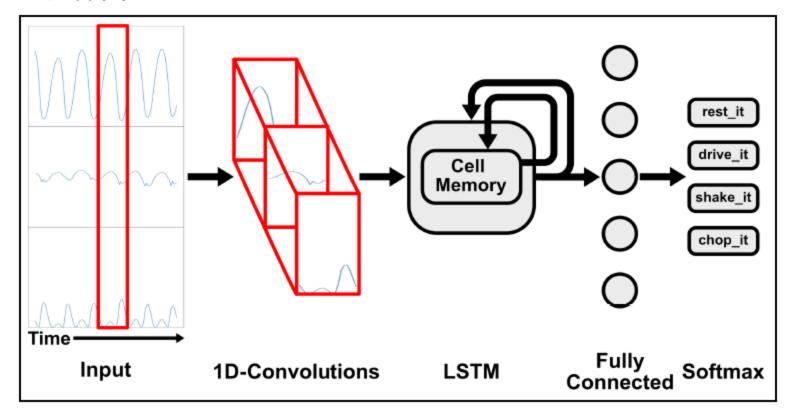
# 长短时记忆

Long short-term memory



# Turi Create的行为识别模型

#### CNN和RNN的结合



# 最终任务: 节奏达人游戏

- App按节奏给出用户需完成的姿势提示,用户手持设备在规定时间内 完成姿势
  - 挥砍
  - 抖动
  - 。 转动方向盘

# 行为识别模型的基本信息

```
static let samplesPerSecond = 25.0
static let numberOfFeatures = 6
static let windowSize = 20
```

# 获取传感器采样数据

```
static private func makeMLMultiArray(numberOfSamples: Int) ->
MLMultiArray? {
try? MLMultiArray(
shape: [1, numberOfSamples, Config.numberOfFeatures] as [NSNumber],
dataType: .double)
}
```

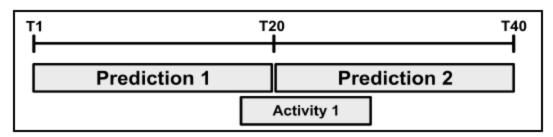
```
let modelInput: MLMultiArray! =
GameViewController.makeMLMultiArray(numberOfSamples:
Config.windowSize)
```

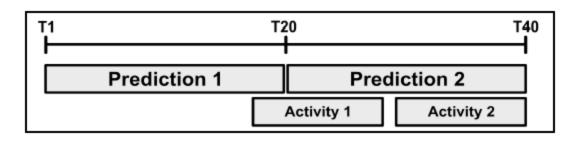
# 模型输入数据

窗口: 20帧

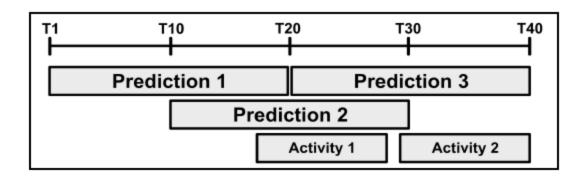


#### 实际运动姿势和预测窗口的叠加





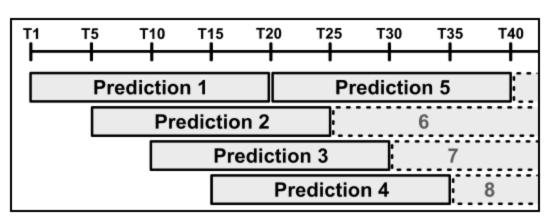
# 窗口的错位预测



#### 通过如下代码设置

```
static let windowOffset = 5
static let numberOfWindows = windowSize / windowOffset
```

# 窗口的错位预测



Gesture It's overlapping predictions — windowSize=20, windowOffset=5

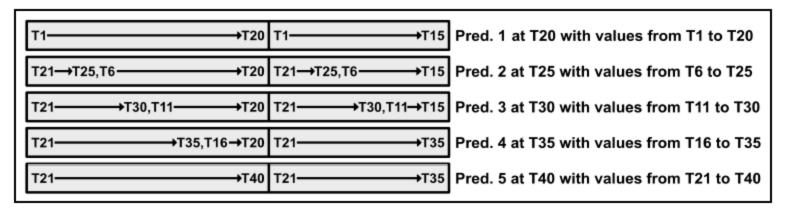
为了满足模型错位预测的需要,输入数据也需准备多份。通过buffer来实现(buffer长度35)

```
static let bufferSize =
windowSize + windowOffset * (numberOfWindows - 1)
```

### 窗口的错位预测

```
let dataBuffer: MLMultiArray! =
GameViewController.makeMLMultiArray(numberOfSamples:
Config.bufferSize)
var bufferIndex = 0
var isDataAvailable = false
```

#### 模型输入buffer示意图



### buffer数据填充

```
func buffer(motionData: CMDeviceMotion) {
for offset in [0, Config.windowSize] {
let index = bufferIndex + offset
if index >= Config.bufferSize {
addToBuffer(index, 0, motionData.rotationRate.x)
addToBuffer(index, 1, motionData.rotationRate.y)
addToBuffer(index, 2, motionData.rotationRate.z)
addToBuffer(index, 3, motionData.userAcceleration.x)
addToBuffer(index, 4, motionData.userAcceleration.y)
addToBuffer(index, 5, motionData.userAcceleration.z)
```

# 构建数据处理pipeline

• 向buffer存储传感器数据

```
buffer(motionData: motionData)
```

• 更新bufferIndex

```
bufferIndex = (bufferIndex + 1) % Config.windowSize
```

开始一次姿势的识别

### 开始一次姿势的识别

```
if isDataAvailable &&
bufferIndex % Config.windowOffset == 0 &&
bufferIndex + Config.windowOffset <= Config.windowSize {</pre>
let window = bufferIndex / Config.windowOffset
memcpy (modelInput.dataPointer,
dataBuffer.dataPointer.advanced(
by: window * Config.windowOffsetAsBytes),
Config.windowSizeAsBytes)
// <u>TODO:</u> predict the gesture
```