Spark逻辑回归(LR)模型使用介绍

```
In
   [2]:
              #小案例学习spark LR模型的使用
           2
              from pyspark.ml.feature import VectorAssembler
           3
              import pandas as pd
           4
              sample dataset = [
                   (0, "male", 37, 10, "no", 3, 18, 7, 4),
           5
           6
                   (0, "female", 27, 4, "no", 4, 14, 6, 4),
           7
                   (0, "female", 32, 15, "yes", 1, 12, 1, 4),
                   (0, "male", 57, 15, "yes", 5, 18, 6, 5),
           8
                   (0, "male", 22, 0.75, "no", 2, 17, 6, 3),
           9
                   (0, "female", 32, 1.5, "no", 2, 17, 5, 5),
          10
          11
                   (0, "female", 22, 0.75, "no", 2, 12, 1, 3),
                   (0, "male", 57, 15, "yes", 2, 14, 4, 4),
          12
          13
                   (0, "female", 32, 15, "yes", 4, 16, 1, 2),
                  (0, "male", 22, 1.5, "no", 4, 14, 4, 5),
          14
                  (0, "male", 37, 15, "yes", 2, 20, 7, 2),
          15
                  (0, "male", 27, 4, "yes", 4, 18, 6, 4),
          16
                   (0, "male", 47, 15, "yes", 5, 17, 6, 4),
          17
          18
                   (0, "female", 22, 1.5, "no", 2, 17, 5, 4),
                  (0, "female", 27, 4, "no", 4, 14, 5, 4),
          19
                   (0, "female", 37, 15, "yes", 1, 17, 5, 5),
          20
          21
                   (0, "female", 37, 15, "yes", 2, 18, 4, 3),
          22
                   (0, "female", 22, 0.75, "no", 3, 16, 5, 4),
                   (0, "female", 22, 1.5, "no", 2, 16, 5, 5),
          23
                  (0, "female", 27, 10, "yes", 2, 14, 1, 5),
          24
          25
                   (1, "female", 32, 15, "yes", 3, 14, 3, 2),
                  (1, "female", 27, 7, "yes", 4, 16, 1, 2),
          26
                  (1, "male", 42, 15, "yes", 3, 18, 6, 2),
          27
          28
                  (1, "female", 42, 15, "yes", 2, 14, 3, 2),
                  (1, "male", 27, 7, "yes", 2, 17, 5, 4),
          29
          30
                  (1, "male", 32, 10, "yes", 4, 14, 4, 3),
                  (1, "male", 47, 15, "yes", 3, 16, 4, 2),
          31
          32
                  (0, "male", 37, 4, "yes", 2, 20, 6, 4)
          33
              7
              columns = ["affairs", "gender", "age", "label", "children", "religiousness", "education
          34
              # pandas构建dataframe,方便
              pdf = pd. DataFrame (sample dataset, columns=columns)
          37
              df = spark.createDataFrame(pdf)
          38
              #特征选取: affairs为目标值,其余为特征值
          39
              df2 = df. select("affairs", "age", "religiousness", "education", "occupation", "rating")
              colArray2 = ["age", "religiousness", "education", "occupation", "rating"]
          40
          41
              df3 = VectorAssembler().setInputCols(colArray2).setOutputCol('features').transform(df2
              print('数据集:')
          42
              df3. show(truncate=False)
          43
          44
              # 随机切分 训练集和测试集
              trainDF, testDF = df3. randomSplit([0.8, 0.2])
          45
          46
              print('训练集:')
          47
              trainDF. show(10, truncate=False)
              print('测试集:')
          48
              testDF. show(10, truncate=False)
```

数据集:

-	双师元:	L	L			L	L
	affairs	age	religiousness	education	occupation	rating	features
-	0	37	3	18	7	4	[37. 0, 3. 0, 18. 0, 7. 0, 4. 0]
	0	27	4	14	6	4	[27. 0, 4. 0, 14. 0, 6. 0, 4. 0]
	0	32	1	12	1	4	[32. 0, 1. 0, 12. 0, 1. 0, 4. 0]

0	57	5	18	6	5	[57. 0, 5. 0, 18. 0, 6. 0, 5. 0]
0	22	2	17	6	3	[22. 0, 2. 0, 17. 0, 6. 0, 3. 0]
0	32	2	17	5	5	[32. 0, 2. 0, 17. 0, 5. 0, 5. 0]
0	22	2	12	1	3	[22. 0, 2. 0, 12. 0, 1. 0, 3. 0]
0	57	2	14	4	4	[57. 0, 2. 0, 14. 0, 4. 0, 4. 0]
0	32	4	16	1	2	[32. 0, 4. 0, 16. 0, 1. 0, 2. 0]
0	22	4	14	4	5	[22. 0, 4. 0, 14. 0, 4. 0, 5. 0]
0	37	2	20	7	2	[37. 0, 2. 0, 20. 0, 7. 0, 2. 0]
0	27	4	18	6	4	[27. 0, 4. 0, 18. 0, 6. 0, 4. 0]
0	47	5	17	6	4	[47. 0, 5. 0, 17. 0, 6. 0, 4. 0]
0	22	2	17	5	4	[22. 0, 2. 0, 17. 0, 5. 0, 4. 0]
0	27	4	14	5	4	[27. 0, 4. 0, 14. 0, 5. 0, 4. 0]
0	37	1	17	5	5	[37. 0, 1. 0, 17. 0, 5. 0, 5. 0]
0	37	2	18	4	3	[37. 0, 2. 0, 18. 0, 4. 0, 3. 0]
0	22	3	16	5	4	[22. 0, 3. 0, 16. 0, 5. 0, 4. 0]
0	22	2	16	5	5	[22. 0, 2. 0, 16. 0, 5. 0, 5. 0]
0	27	2	14	1	5	[27. 0, 2. 0, 14. 0, 1. 0, 5. 0]

only showing top 20 rows

训练集:

				1		1
affairs	age	religiousness	education	occupation	rating	features
0	32	1	12	1	4	[32. 0, 1. 0, 12. 0, 1. 0, 4. 0]
0	37	3	18	7	4	[37. 0, 3. 0, 18. 0, 7. 0, 4. 0]
0	57	5	18	6	5	[57. 0, 5. 0, 18. 0, 6. 0, 5. 0]
0	32	2	17	5	5	[32. 0, 2. 0, 17. 0, 5. 0, 5. 0]
0	57	2	14	4	4	[57. 0, 2. 0, 14. 0, 4. 0, 4. 0]
0	22	4	14	4	5	[22. 0, 4. 0, 14. 0, 4. 0, 5. 0]
0	37	2	20	7	2	[37. 0, 2. 0, 20. 0, 7. 0, 2. 0]
0	22	2	17	5	4	[22. 0, 2. 0, 17. 0, 5. 0, 4. 0]
0	27	4	14	5	4	[27. 0, 4. 0, 14. 0, 5. 0, 4. 0]
0	37	1	17	5	5	[37. 0, 1. 0, 17. 0, 5. 0, 5. 0]
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only showing top 10 rows

测试集:

	affairs	age	religiousness	education	occupation	rating	features
	0	27	4	14	6	4	[27. 0, 4. 0, 14. 0, 6. 0, 4. 0]
	0	22	2	12	1	3	[22. 0, 2. 0, 12. 0, 1. 0, 3. 0]
	0	22	2	17	6	3	[22. 0, 2. 0, 17. 0, 6. 0, 3. 0]
	0	27	4	18	6	4	[27. 0, 4. 0, 18. 0, 6. 0, 4. 0]
	0	32	4	16	1	2	[32. 0, 4. 0, 16. 0, 1. 0, 2. 0]
	0	47	5	17	6	4	[47. 0, 5. 0, 17. 0, 6. 0, 4. 0]
	0	37	2	20	6	4	[37. 0, 2. 0, 20. 0, 6. 0, 4. 0]
	1	42	3	18	6	2	[42. 0, 3. 0, 18. 0, 6. 0, 2. 0]
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In [3]: # 逻辑回归训练模型
2 from pyspark.ml.classification import LogisticRegression
3 lr = LogisticRegression()
4 model = lr.setLabelCol('affairs').setFeaturesCol('features').fit(trainDF)
5 model.transform(testDF).show(truncate=False)
```

```
|affairs | age | religiousness | education | occupation | rating | features
                                                                   rawPredi
                         probability
                                                               prediction
ction
             |14  |6  |4  |[27, 0, 4, 0, 14, 0, 6, 0, 4, 0]|[1, 43701]
   |27 |4
28406197153, -1. 4370128406197153] | [0. 8079916464280155, 0. 19200835357198462] | 0. 0
             | 12 | | 3 | | [22. 0, 2. 0, 12. 0, 1. 0, 3. 0] | [-2. 6357
535531749416, 2. 6357535531749416] [0. 06687252951201492, 0. 9331274704879852] |1. 0
             |17  |6  |3  |[22. 0, 2. 0, 17. 0, 6. 0, 3. 0]|[0. 65492]
44841168171, -0. 6549244841168171] [0. 65811932430713, 0. 34188067569287] | 0. 0
              |18| |6| |4| |[27, 0, 4, 0, 18, 0, 6, 0, 4, 0]|[2, 37007]
      27 4
15957681684, -2. 3700715957681684] | [0. 91451645780968, 0. 08548354219031987] | 0. 0
     |32|4
              85738947233, 4. 628685738947233] | [0.009673104928627526, 0.9903268950713724] | 1.0
             | 17  | 6  | 4  | [47, 0, 5, 0, 17, 0, 6, 0, 4, 0] | [3, 05035]
      47 | 5
41549910054, -3. 0503541549910054] | [0. 954797813935877, 0. 04520218606412298] | 0. 0
             |20| |6| |4| |[37, 0, 2, 0, 20, 0, 6, 0, 4, 0]|[4, 93127]
      37 | 2
37927811945, -4. 9312737927811945] | [0. 992834412152514, 0. 0071655878474860945] | 0. 0
      42 | 3
               | 18 | 6 | 2 | [42. 0, 3. 0, 18. 0, 6. 0, 2. 0] | [-0. 5982
88626751545, 0. 598288626751545] | [0. 354735326530606, 0. 6452646734693941] | 1. 0
```