```
!pip install pyspark
    Looking in indexes: <a href="https://pypi.org/simple">https://us-python.pkg.dev/colab-w</a>
    Collecting pyspark
      Downloading pyspark-3.2.1.tar.gz (281.4 MB)
                             281.4 MB 34 kB/s
    Collecting py4j==0.10.9.3
      Downloading py4j-0.10.9.3-py2.py3-none-any.whl (198 kB)
                       198 kB 53.7 MB/s
    Building wheels for collected packages: pyspark
      Building wheel for pyspark (setup.py) ... done
      Created wheel for pyspark: filename=pyspark-3.2.1-py2.py3-none-any.whl size=
      Stored in directory: /root/.cache/pip/wheels/9f/f5/07/7cd8017084dce4e93e84e9
    Successfully built pyspark
    Installing collected packages: py4j, pyspark
    Successfully installed py4j-0.10.9.3 pyspark-3.2.1
from pyspark.sql import SparkSession
spark=SparkSession.builder.appName('SparkML').getOrCreate()
spark
    SparkSession - in-memory
    SparkContext
 Saved successfully!
         v3.2.1
    Master
         local[*]
    AppName
         SparkML
df = spark.read.csv('IRIS.csv',header=True,inferSchema=True)
df.printSchema()
    root.
     |-- sepal length: double (nullable = true)
     |-- sepal width: double (nullable = true)
     |-- petal_length: double (nullable = true)
     |-- petal width: double (nullable = true)
     |-- species: string (nullable = true)
df.show(5)
    |sepal_length|sepal_width|petal_length|petal_width|
        -----+
              5.1
                         3.5
                                      1.4
                                                  0.2 | Iris-setosa |
```

3.0

1.4

0.2 | Iris-setosa |

4.9

```
#Preprocessing steps
from pyspark.ml.feature import StringIndexer, OneHotEncoder
# create object of StringIndexer class and specify input and output column
SI species = StringIndexer(inputCol='species',outputCol='species Index')
# transform the data
df = SI species.fit(df).transform(df)
df.tail(5)
    [Row(sepal length=6.7, sepal width=3.0, petal length=5.2, petal width=2.3, spe
     Row(sepal length=6.3, sepal width=2.5, petal length=5.0, petal width=1.9, spe
     Row(sepal length=6.5, sepal width=3.0, petal length=5.2, petal width=2.0, spe
     Row(sepal length=6.2, sepal width=3.4, petal length=5.4, petal width=2.3, spe
     Row(senal length=5.9 senal width=3.0, petal length=5.1, petal width=1.8, spe
 Saved successfully!
ar.printschema()
    root
     |-- sepal length: double (nullable = true)
     |-- sepal width: double (nullable = true)
     |-- petal length: double (nullable = true)
     |-- petal width: double (nullable = true)
     |-- species: string (nullable = true)
     |-- species Index: double (nullable = false)
from pyspark.ml.feature import VectorAssembler
fa1=VectorAssembler(inputCols=['sepal length', 'sepal width', 'petal length', 'petal w
df1=fa1.transform(df)
df1.show()
    +----+
    |sepal_length|sepal_width|petal_length|petal_width| species|species_Index|
```

_____+

1.4

1.4

1.3|

1.5

1.4

0.2|Iris-setosa|

0.2|Iris-setosa|

0.2|Iris-setosa|

0.2|Iris-setosa|

0.2|Iris-setosa|

3.5

3.0

3.2

3.1

3.6

5.1

4.9

4.7

4.6

5.0

1 0.0

] | 0.0

1 0.0

1 0.0

] | 0.0

4.9 5.4 4.8 4.8 4.3 5.8 5.7 5.4 5.1 5.7	3.1 3.7 3.4 3.0 3.0 4.0 4.4 3.9 3.5 3.8 3.8	1.5 1.6 1.4 1.1 1.2 1.5 1.3 1.4	0.2 0.2 0.1 0.1 0.2 0.4 0.4 0.3	Iris-setosa Iris-setosa Iris-setosa Iris-setosa Iris-setosa Iris-setosa Iris-setosa Iris-setosa Iris-setosa Iris-setosa] 0.0] 0.0] 0.0] 0.0] 0.0] 0.0] 0.0
!		1.4 1.5 1.4 1.5	0.3 0.2 0.2 0.1 0.1	•] 0.0] 0.0] 0.0] 0.0] 0.0

only showing top 20 rows

```
df2=df1.select("newic1","species_Index")
```

```
df2.show()
```

```
newic1|species Index|
  15 1 2 5 1 1 0 21
                            0.0
                            0.0
Saved successfully!
                            0.0
   [4.6,3.1,1.5,0.2]
                           0.0
   [5.0,3.6,1.4,0.2]
                            0.0
   [5.4,3.9,1.7,0.4]
                            0.0
   [4.6,3.4,1.4,0.3]
                            0.0
   |[5.0,3.4,1.5,0.2]|
                            0.0
   [4.4,2.9,1.4,0.2]
                            0.0
   [4.9,3.1,1.5,0.1]
                            0.0
   [5.4,3.7,1.5,0.2]
                            0.0
   [4.8,3.4,1.6,0.2]
                            0.0
   [4.8,3.0,1.4,0.1]
                            0.0
   [4.3,3.0,1.1,0.1]
                            0.0
   [5.8,4.0,1.2,0.2]
                            0.0
   [5.7,4.4,1.5,0.4]
                            0.0
   [5.4,3.9,1.3,0.4]
                            0.0
   [5.1,3.5,1.4,0.3]
                            0.0
   [5.7,3.8,1.7,0.3]
                            0.0
   [5.1,3.8,1.5,0.3]
                            0.0
  +----+
```

only showing top 20 rows

```
from pyspark.sql.functions import *
print(df.stat.corr('pg','Outcome'))
print(df.stat.corr('g','Outcome'))
print(df.stat.corr('bp','Outcome'))
print(df.stat.corr('st','Outcome'))
```

```
0.22189815303398636

0.4665813983068737

0.06506835955033274

0.07475223191831945

0.13054795488404794

0.2926946626444454

0.17384406565296

0.23835598302719757

1.0
```

from pyspark.ml.classification import RandomForestClassifier
train_data,test_data=df2.randomSplit([0.75,0.25])
applyml=RandomForestClassifier(featuresCol='newic1', labelCol='species_Index')
applyml=applyml.fit(train_data)

predict=applyml.evaluate(test_data)

predict.predictions.show()

/usr/local/lib/python3.7/dist-packages/pyspark/sql/context.py:127: FutureWarni FutureWarning

newic1	species_Index	rawPrediction	probability	prediction
[4.4,3.0,1.3,0.2]	0.0	[20.0,0.0,0.0]	[1.0,0.0,0.0]	0.0
	0.0	[20.0,0.0,0.0]	[1.0,0.0,0.0]	0.0
ed successfully!	× 0.0	[20.0,0.0,0.0]	[1.0,0.0,0.0]	0.0
,	1.0	[0.0,20.0,0.0]	[0.0,1.0,0.0]	1.0
[5.0,3.2,1.2,0.2]	0.0	[20.0,0.0,0.0]	[1.0,0.0,0.0]	0.0
[5.0,3.5,1.6,0.6]	0.0	[17.0,3.0,0.0]	[0.85,0.15,0.0]	0.0
[5.1,3.8,1.5,0.3]	0.0	[20.0,0.0,0.0]	[1.0,0.0,0.0]	0.0
[5.2,4.1,1.5,0.1]	0.0	[20.0,0.0,0.0]	[1.0,0.0,0.0]	0.0
[5.3,3.7,1.5,0.2]	0.0	[20.0,0.0,0.0]	[1.0,0.0,0.0]	0.0
[5.4,3.4,1.5,0.4]	0.0	[18.0,2.0,0.0]	[0.9,0.1,0.0]	0.0
[5.4,3.9,1.3,0.4]	0.0	[18.0,2.0,0.0]	[0.9,0.1,0.0]	0.0
[5.5,3.5,1.3,0.2]	0.0	[19.0,1.0,0.0]	[0.95,0.05,0.0]	0.0
[5.6,3.0,4.1,1.3]	1.0	[0.0,19.0,1.0]	[0.0,0.95,0.05]	1.0
[5.7,2.5,5.0,2.0]	2.0	[0.0,3.5,16.5]	[0.0,0.175,0.825]	2.0
[5.7,3.0,4.2,1.2]	1.0	[0.0,20.0,0.0]	[0.0,1.0,0.0]	1.0
[5.7,3.8,1.7,0.3]	0.0	[12.0,8.0,0.0]	[0.6,0.4,0.0]	0.0
[5.7,4.4,1.5,0.4]	0.0	[12.0,8.0,0.0]	[0.6,0.4,0.0]	0.0
[5.8,2.6,4.0,1.2]	1.0	[0.0,20.0,0.0]	[0.0,1.0,0.0]	1.0
[5.8,2.8,5.1,2.4]	2.0	[0.0,1.5,18.5]	[0.0,0.075,0.925]	2.0
[5.9,3.0,4.2,1.5]	1.0	[0.0,20.0,0.0]	[0.0,1.0,0.0]	1.0

only showing top 20 rows

predict.accuracy

0.9347826086956522

