GenerateFakeClassificationData

June 26, 2019

• The Python package rpy2 has an RMagic module that facilitates using R inside a Jupyter notebook running Python

Any Jupyter notebook cell that contains R code must be prefaced by the %%R "magic" statement.

```
In [2]: %%R
        library(tidyverse)
        library(GGally)
R[write to console]: Registered S3 methods overwritten by 'ggplot2':
 method
                from
  [.quosures
                rlang
  c.quosures
                rlang
 print.quosures rlang
R[write to console]: Registered S3 method overwritten by 'rvest':
 method
                   from
  read_xml.response xml2
R[write to console]: Attaching packages tidyverse 1.2.1
R[write to console]: ggplot2 3.1.1
                                        purrr
                                                0.3.2
 tibble 2.1.1
                   dplyr 0.8.1
 tidyr 0.8.3
                   stringr 1.4.0
readr 1.3.1
                  forcats 0.4.0
R[write to console]: Conflicts tidyverse_conflicts()
 dplyr::filter() masks stats::filter()
 dplyr::lag()
                masks stats::lag()
R[write to console]: Registered S3 method overwritten by 'GGally':
 method from
        ggplot2
  +.gg
```

```
R[write to console]:
Attaching package: GGally

R[write to console]: The following object is masked from package:dplyr:

nasa
```

1 Make funky data

- We'll use the Python package Scikit-learn
- The module sklearn.datasets has a function called make_classification()
- Creates clusters of points
- Points are normally distributed about vertices in N-dimensional hypercube
 - N is the number of informative dimensions
 - Cube side length = 2 * class_sep

```
In [3]: from sklearn.datasets import make_classification, make_moons, make_circles
        import pandas as pd
In [4]: Xgauss, Y = make_classification(
            n_samples=200,
            n_features=2,
            n_informative=2,
            n_redundant=0,
            n_repeated=0,
            n_classes=2,
            n_clusters_per_class=1,
            weights=None,
            flip_y=0.00,
            class_sep=0.6,
            hypercube=True,
            shift=0.0,
            scale=1.0,
            shuffle=False,
            random_state=None,)
```

1.1 Create a Python DataFrame

1.2 Pass the data to R and make a scatterplot matrix

R[write to console]: `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.

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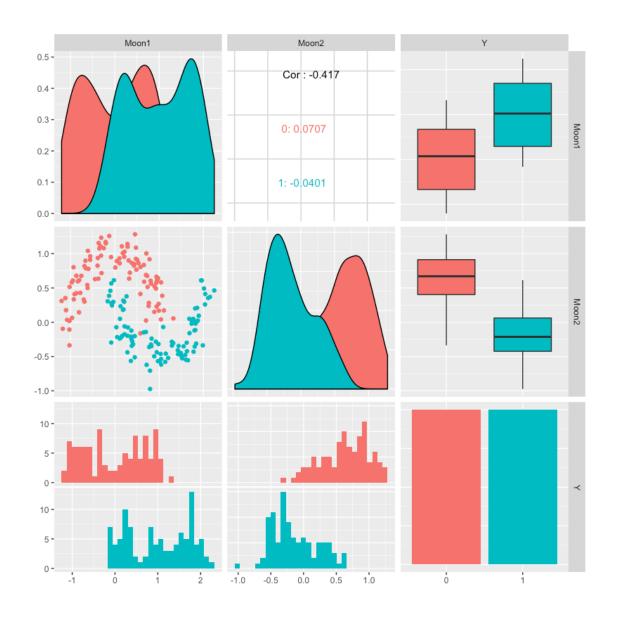


In [7]: data1.drop(columns='Y', inplace=True)

```
In [8]: Xmoon, Y = make_moons( n_samples=200, shuffle=False, noise=0.15, random_state=None)
```

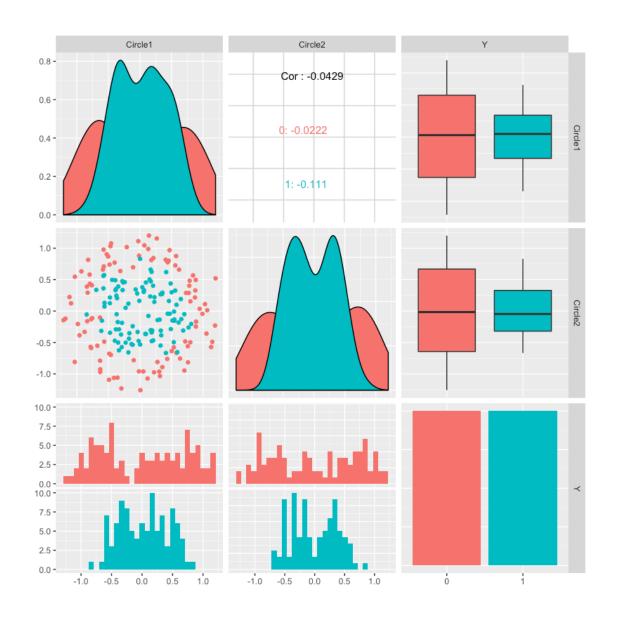
R[write to console]: `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.

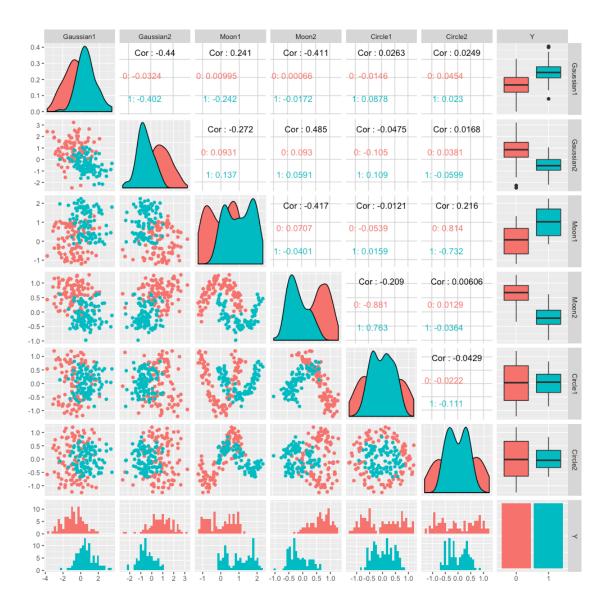
R[write to console]: `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.



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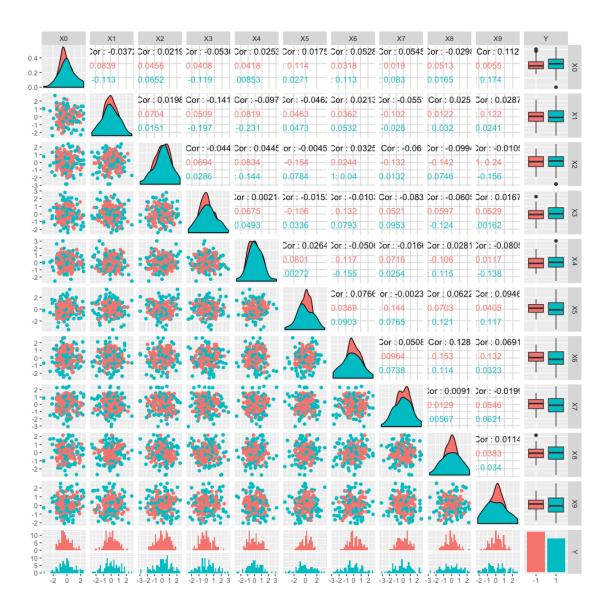




In [19]: data.sample(frac=1).to_csv('funkydata.csv', index=False)

2 Make Unequal variance data

```
In [20]: from sklearn.datasets import make_hastie_10_2
In [21]: Xhastie, y = make_hastie_10_2( 200 )
In [22]: Xhastie.shape
Out[22]: (200, 10)
In [23]: data4 = pd.DataFrame( Xhastie )
```



In [26]: data4.sample(frac=1).to_csv('unequal_variance_data.csv', index=False)