



# Computação científica com Python

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Programando Ciência

# Por que Python?

**Por que não uma  
linguagem compilada?**

**C / C++: menos produtivo**

## **Fortran:**

- **menos produtivo**
- **restrito**

**Por que não outra  
linguagem interpretada?**

## **Matlab / Octave:**

- **restrito**
- **lento**
- **caro**

**R: curva de aprendizado maior**



**Julia:** poucos pacotes

**“Every tool has its place”.**

## **Python:**

- **adoção confusa (2 ou 3?)**
- **lento**

**Vamos falar de Python :)**

**Qual a melhor distribuição?**

**sudo VS**



```
$ sudo apt-get install idle-  
python3.4 ipython3-qtconsole  
python3-numpy python3-scipy  
python3-matplotlib mayavi2  
python3-pandas python3-  
skimage python3-simpy...
```

```
$ bash Anaconda3-2.3.0-Linux-  
x86_64.sh
```



```
$ conda install  
<pacote>
```



**KEEP  
CALM  
AND  
CONDA  
INSTALL**

**E o ambiente?**

IP[y] ?



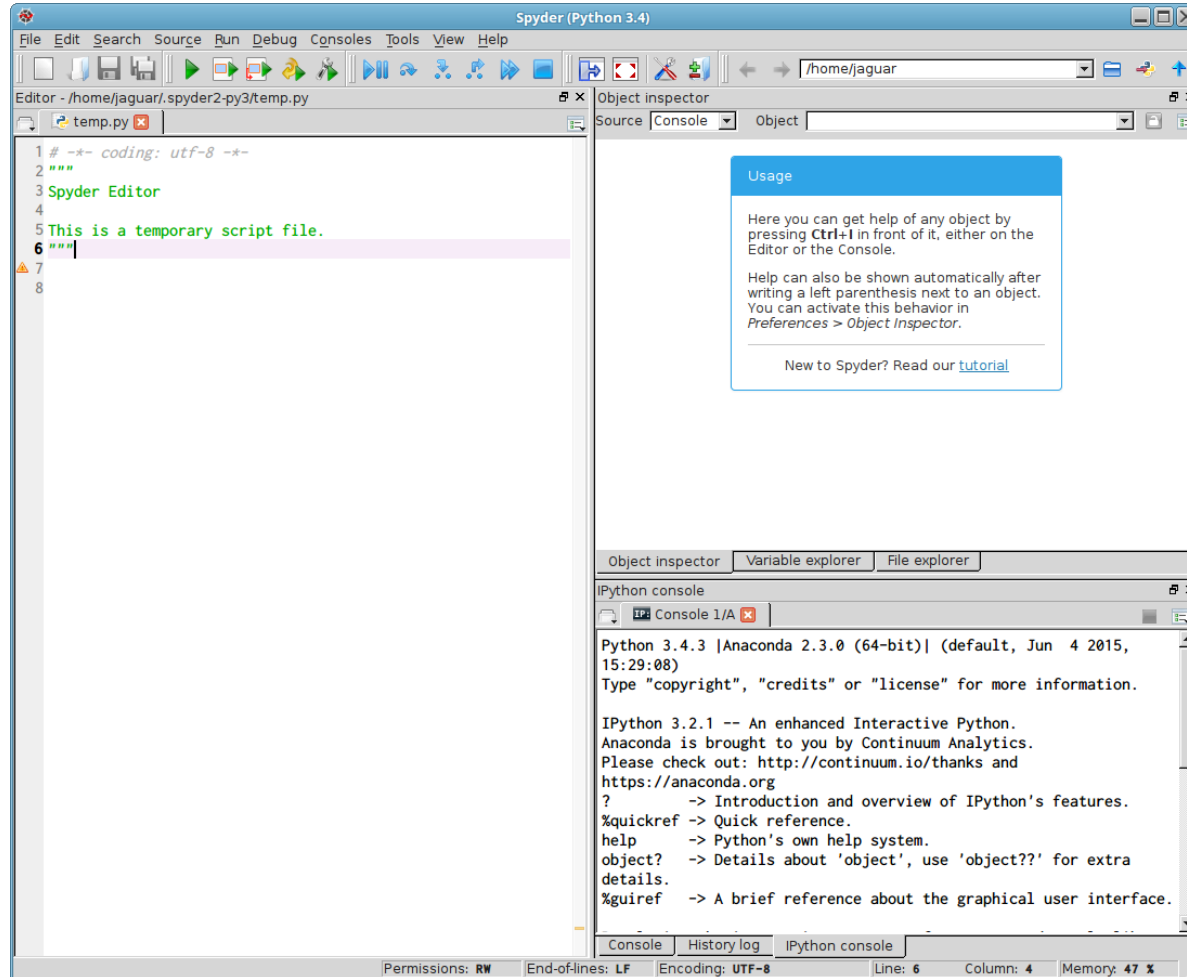
idle?



?

```
$ conda install  
spyder
```

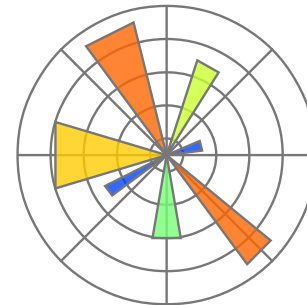
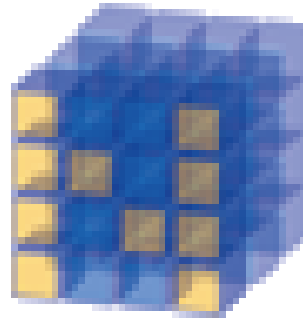




<https://github.com/spyder-ide/spyder>

**E o que posso fazer com isso?**

- o básico



```
import matplotlib.pyplot as plt
import numpy as np
```

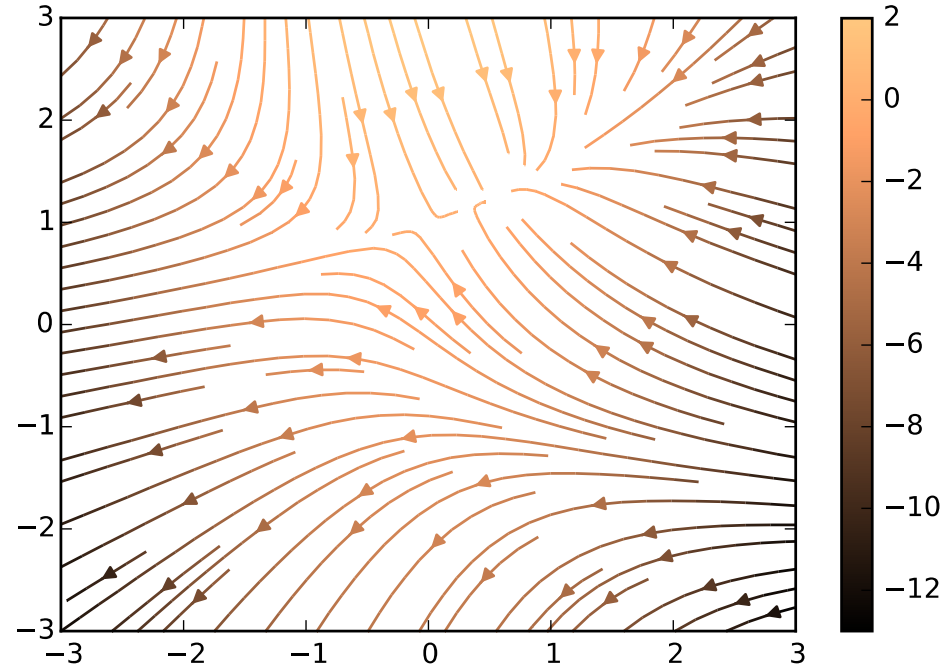
```
Y, X = np.mgrid[-3:3:100j,  
-3:3:100j]
```

```
U = -1 - X**2 + Y
```

```
V = 1 + X - Y**2
```

```
speed = np.sqrt(U*U + V*V)
```

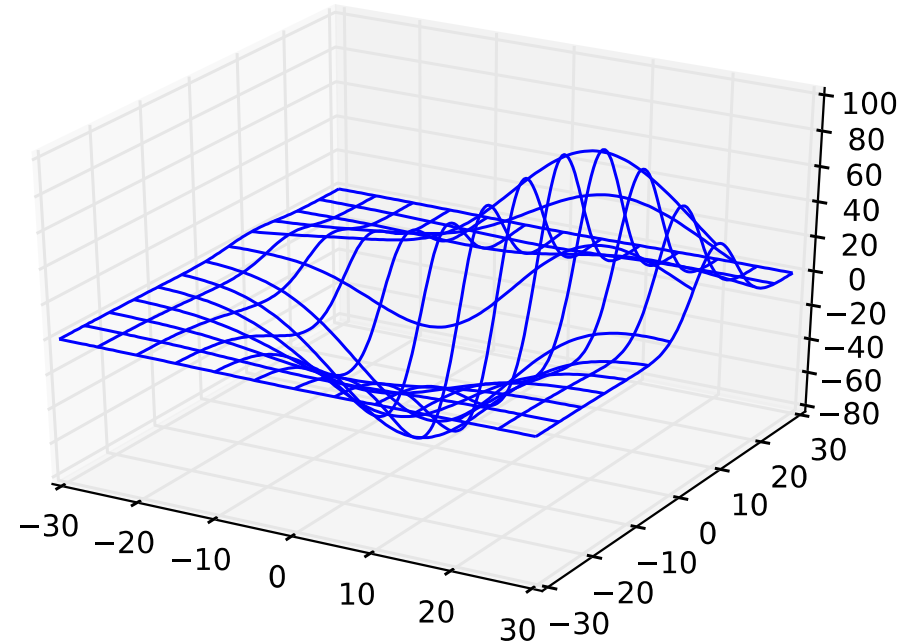
```
plt.streamplot(X, Y, U, V, color=U,  
cmap='copper')  
plt.colorbar()  
plt.show()
```





```
from mpl_toolkits.mplot3d import  
axes3d  
import matplotlib.pyplot as plt  
import numpy as np
```

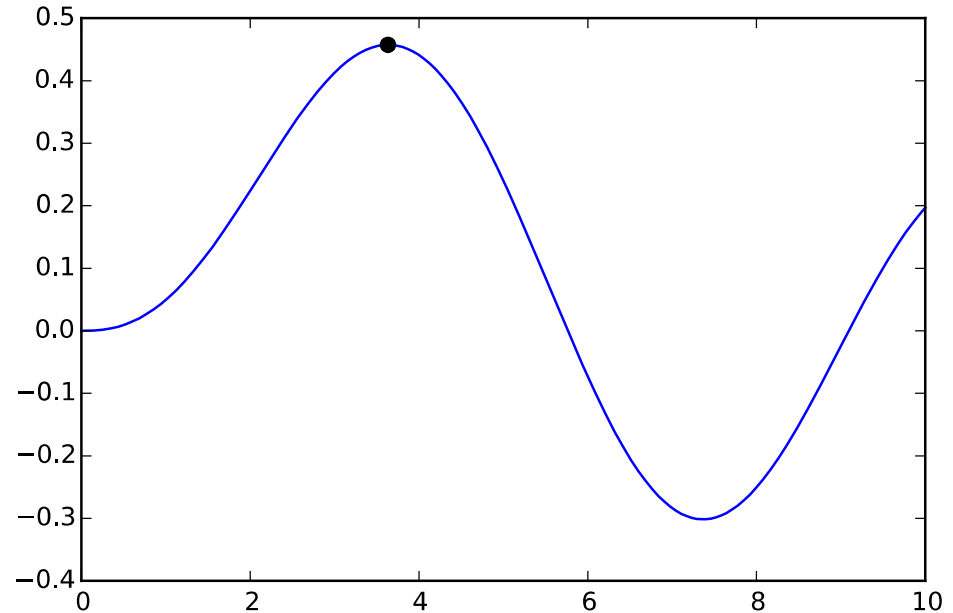
```
fig = plt.figure()  
ax = fig.add_subplot(111,  
projection='3d')  
X, Y, Z =  
axes3d.get_test_data(0.05)  
ax.plot_wireframe(X, Y, Z,  
rstride=10, cstride=10)  
  
plt.show()
```



```
from scipy import optimize, special
import matplotlib.pyplot as plt
import numpy as np
```

```
def f(x):
    output = -special.jv(2.5, x)
    return output
```

```
sol = optimize.minimize(f, 1.0)
x = np.linspace(0, 10, 5000)
plt.plot(x, special.jv(2.5, x),
        '-', sol.x, -sol.fun, 'ko')
plt.show()
```



- **matemática  
simbólica**



```
In [30]: from sympy import *
```

```
In [31]: f, g = symbols('f g',  
cls=Function)
```

```
In [32]: f(x)  
Out[32]: f(x)
```

```
In [33]: f(x).diff(x)  
Out[33]: Derivative(f(x), x)
```

```
In [34]: diffeq =  
Eq(f(x).diff(x, x) -  
2*f(x).diff(x) + f(x), sin(x))
```

```
In [35]: diffeq  
Out[35]: f(x) - 2*Derivative(f(x),  
x) + Derivative(f(x), x, x) ==  
sin(x)
```

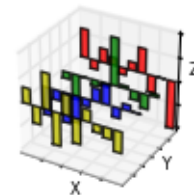
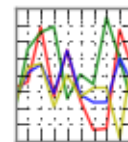
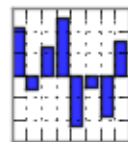
```
In [36]: dsolve(diffeq, f(x))  
Out[36]: f(x) == (C1 + C2*x)*exp(x)  
+ cos(x)/2
```

<http://docs.sympy.org/latest/tutorial/solvers.html#solving-differential-equations>

- **estadística**

pandas

$$y_{it} = \beta' x_{it} + \mu_i + \epsilon_{it}$$

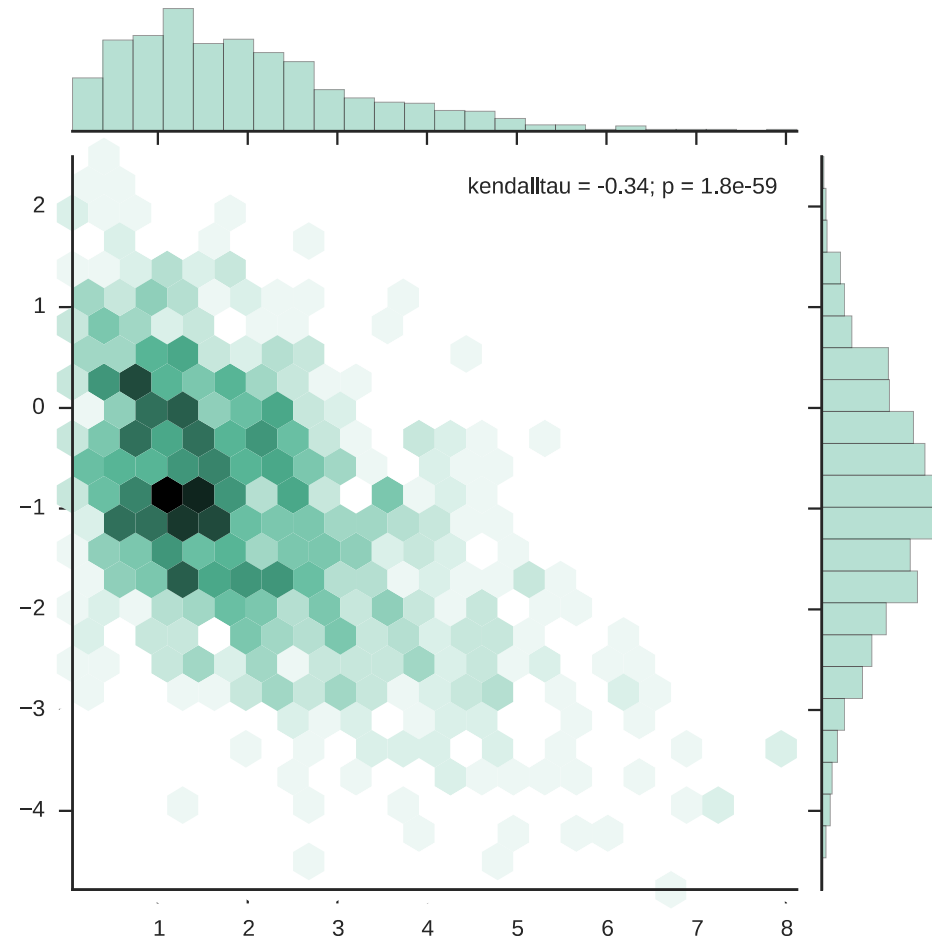


seaborn

```
from scipy.stats import kendalltau
import numpy as np
import seaborn as sns
sns.set(style="ticks")
```

```
rs = np.random.RandomState(11)
x = rs.gamma(2, size=1000)
y = -.5 * x + rs.normal(size=1000)
```

```
sns.jointplot(x, y, kind="hex",
stat_func=kendalltau,
color="#4CB391")
```

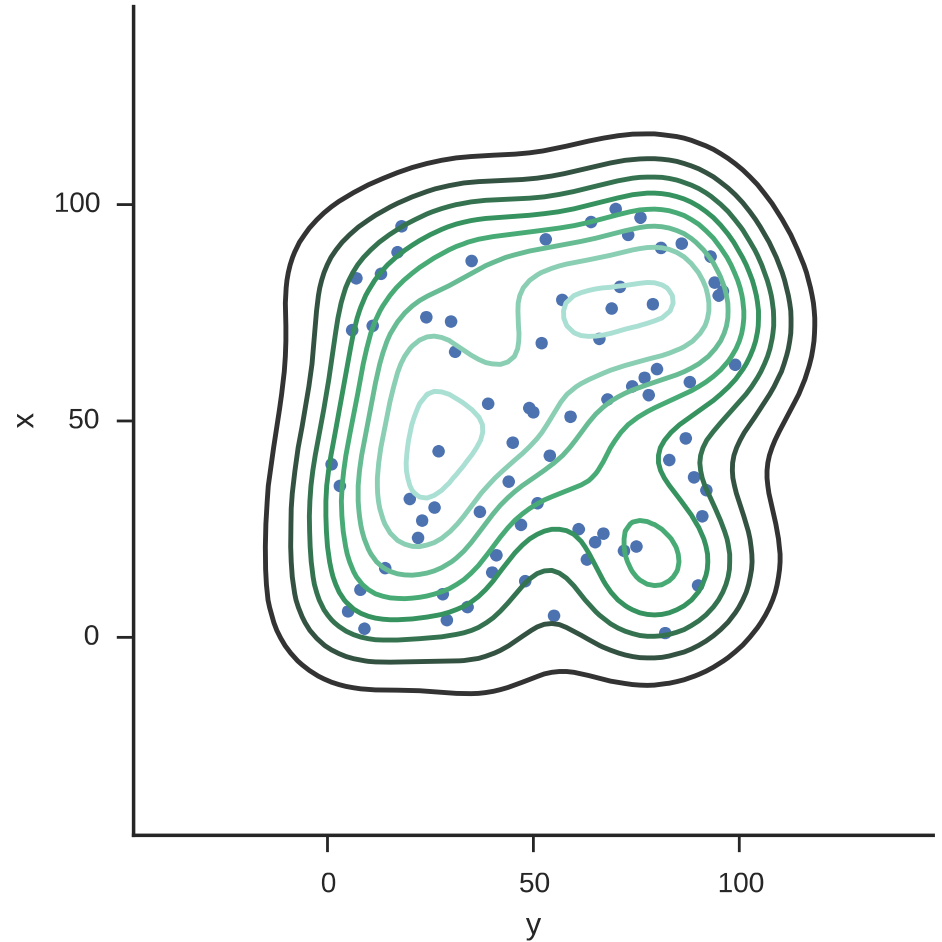


[http://stanford.edu/~mwaskom/software/seaborn/examples/hexbin\\_marginals.html](http://stanford.edu/~mwaskom/software/seaborn/examples/hexbin_marginals.html)

```
import pandas as pd
import random as rd
import seaborn as sns

df = pd.DataFrame()
df['x'] = rd.sample(range(1, 100),
75)
df['y'] = rd.sample(range(1, 100),
75)

sns.lmplot('x', 'y', data=df,
fit_reg=False)
sns.kdeplot(df.y, df.x)
```



- **processamento de imagens**





```

from skimage import data
from skimage.filters import threshold_otsu,
threshold_adaptive
import matplotlib.pyplot as plt

image = data.page()
global_thresh = threshold_otsu(image)
binary_global = image > global_thresh
block_size = 40
binary_adaptive = threshold_adaptive(image,
block_size, offset=10)
fig, axes = plt.subplots(nrows=3, figsize=(7,
8))
ax0, ax1, ax2 = axes
plt.gray()

ax0.imshow(image)
ax1.imshow(binary_global)
ax2.imshow(binary_adaptive)
for ax in axes:
    ax.axis('off')

plt.show()

```

[http://scikit-image.org/docs/dev/auto\\_examples/plot\\_threshold\\_adaptive.html](http://scikit-image.org/docs/dev/auto_examples/plot_threshold_adaptive.html)

## Region-based segmentation

Let us first determine markers of the coins and the background. These markers are pixels that we can label unambiguously as either object or background. Here, the markers are found at the two extreme parts of the histogram of grey values:

```
>>> markers = np.zeros_like(coins)
```

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```

```
from skimage import data
from skimage.filters.rank import
entropy
from skimage.morphology import disk
from skimage.util import img_as_ubyte
import matplotlib.pyplot as plt

image = img_as_ubyte(data.camera())
fig, (ax0, ax1) = plt.subplots(nrows=2,
figsize=(4, 8))
```

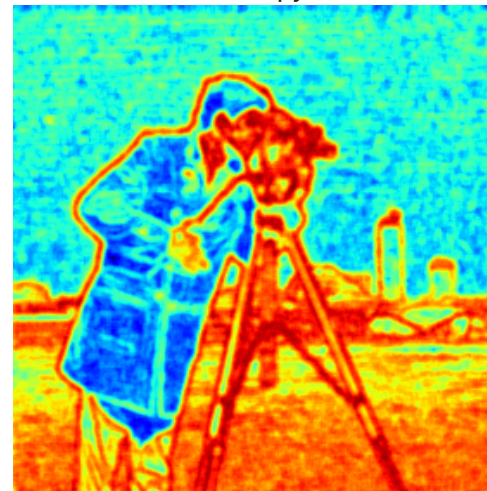
```
img0 = ax0.imshow(image,
cmap=plt.cm.gray)
ax0.set_title('Image')
ax0.axis('off')
img1 = ax1.imshow(entropy(image,
disk(5)), cmap=plt.cm.jet)
ax1.set_title('Entropy')
ax1.axis('off')
```

```
plt.show()
```

Image



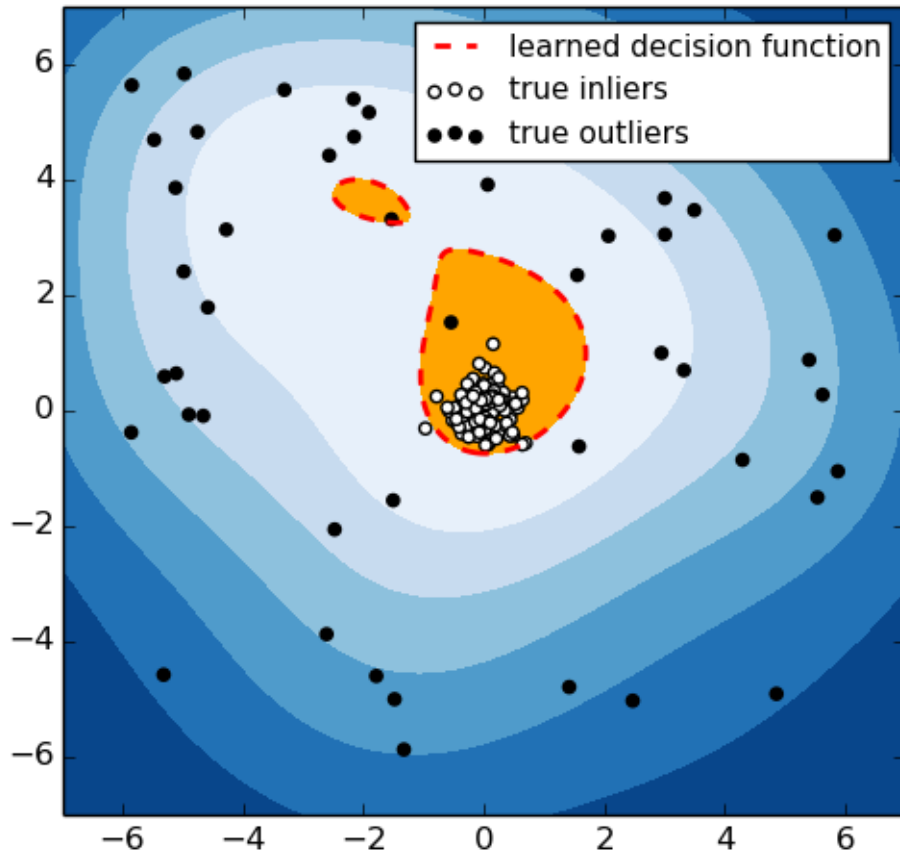
Entropy



- **aprendizagem  
de máquina**

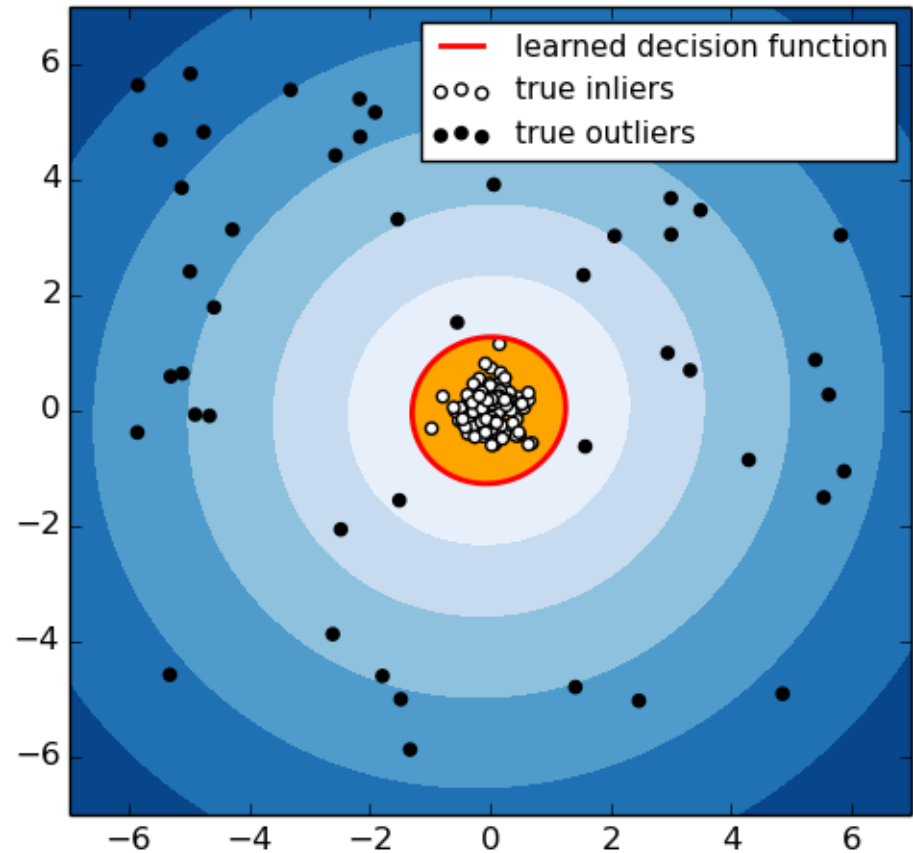


Outlier detection



1. One-Class SVM (errors: 4)

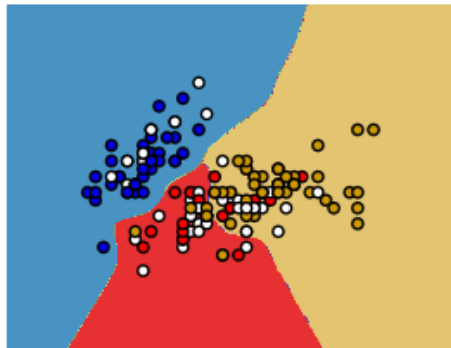
Outlier detection



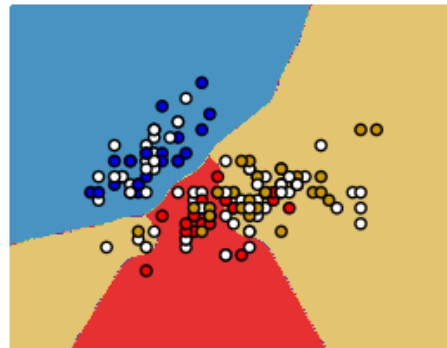
2. robust covariance estimator (errors: 0)

[http://scikit-learn.org/stable/auto\\_examples/covariance/plot\\_outlier\\_detection.html](http://scikit-learn.org/stable/auto_examples/covariance/plot_outlier_detection.html)

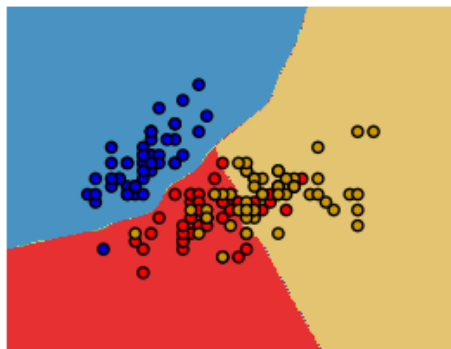
Label Spreading 30% data



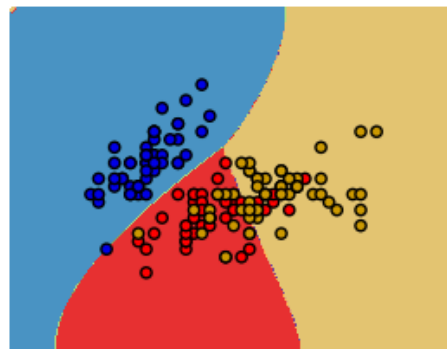
Label Spreading 50% data



Label Spreading 100% data

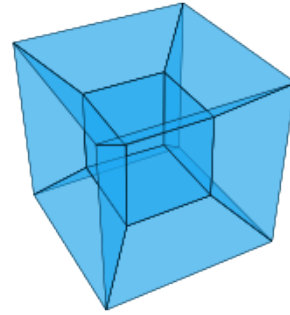


SVC with rbf kernel



Unlabeled points are colored white

mpld3



plotly



*PsychoPy*

Psychology software in Python

SCIKIT-BIO



SimPy

**e muitos outros!**

<https://pypi.python.org/pypi?:action=browse&c=385>

# Obrigado!

- contato: [www.programandociencia.com/sobre](http://www.programandociencia.com/sobre)