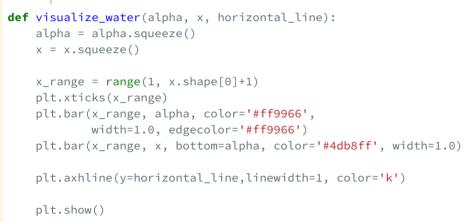
## 参考实现

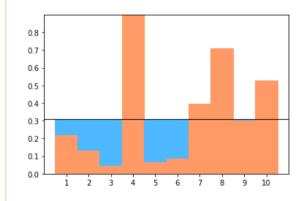
说明:这**只是一个参考**,你也可以按自己的想法进行实现(yep,你可以把这份文档拖进回收站,自己写),但需满足实验说明文档的相关要求。

```
In [1]:
import numpy as np
from matplotlib import pyplot as plt
# %matplotlib inline
# %config InlineBackend.figure_formats = ['svg']
In [2]:
def calculate_target(alpha, x):
    return -(np.log2(alpha+x).sum())
def fill_water(alpha, total_water, precision, track=False):
    # implement your algorithm here
    # ...
    # return your solution
In [3]:
# you can enlarge these parameters to check how efficient
# your algorithm is(not required by this lab),
# I limit it to be very small to help you to begin with.
# And note that small value is good for visualization.
alpha_range = [0.0, 1.0]
total_water = 1.0
dimension = 10
precision = 1e-6
alpha = np.random.uniform(low=alpha_range[0],
                           high=alpha_range[1],
                           size=(dimension, 1))
print(alpha)
 [[0.21889905]
  [0.13042259]
  [0.04632022]
  [0.89854737]
  [0.06499838]
  [0.08789615]
  [0.3975796]
  [0.71104049]
  [0.30180927]
  [0.52901178]]
```

```
In [4]:
x = fill_water(alpha=alpha,
           total_water=total_water,
           precision=precision)
print(x)
print(x.sum())
horizontal_line = (alpha + x).min()
print(horizontal_line)
 [[0.08949205]
  [0.17796851]
  [0.26207088]
  [0.24339272]
  [0.22049495]
  ΓΘ.
  [0.00658184]
  ΓΘ.
          ]]
 1.0000009362768902
 0.30839110058200936
In [5]:
def visualize_water(alpha, x, horizontal_line):
    alpha = alpha.squeeze()
    x = x.squeeze()
```



In [6]: |
visualize\_water(alpha, x, horizontal\_line)



## 迭代过程查看

这部分不是必须的,但是当你的代码出现bug的时候,这些小技巧可能会有帮助。

```
In [7]:
x, targets, errors = fill_water(alpha=alpha,
                         total_water=total_water,
                         precision=precision,
                         track=True)
print(x)
print(targets)
print(errors)
# how much iterations your algorithm need to
# satisfy the precsion required?
print('iteration:', len(errors))
   [[0.08949205]
     [0.17796851]
     [0.26207088]
     [0.24339272]
     [0.22049495]
     ΓΘ.
     [0.
     [0.00658184]
    078580224445885, 13.078621287013183, 13.078637712094647, 13.078644282135958]
   \lceil 1.059111295205447,\ 0.3487903215987713,\ 0.13951612863950835,\ 0.05580645145580343,\ 0.022322580582321283,\ 0.00892903223291283,\ 0.00892903223291283,\ 0.00892903223291283,\ 0.00892903223291283,\ 0.00892903223291283,\ 0.00892903223291283,\ 0.00892903223291283,\ 0.00892903223291283,\ 0.00892903223291283,\ 0.00892903223291283,\ 0.00892903223291283,\ 0.00892903223291283,\ 0.00892903223291283,\ 0.00892903223291283,\ 0.00892903223291283,\ 0.00892903223291283,\ 0.00892903223291283,\ 0.00892903223291283,\ 0.00892903223291283,\ 0.00892903223291283,\ 0.00892903223291283,\ 0.00892903223291283,\ 0.00892903223291283,\ 0.00892903223291283,\ 0.00892903223291283,\ 0.00892903223291283,\ 0.00892903223291283,\ 0.00892903223291283,\ 0.00892903223291283,\ 0.00892903223291283,\ 0.00892903223291283,\ 0.00892903223291283,\ 0.00892903223291283,\ 0.00892903223291283,\ 0.00892903223291283,\ 0.00892903223291283,\ 0.00892903223291283,\ 0.00892903223291283,\ 0.00892903223291283,\ 0.00892903223291283,\ 0.00892903223291283,\ 0.00892903223291283,\ 0.00892903223291283,\ 0.00892903223291283,\ 0.00892903223291283,\ 0.00892903223291283,\ 0.00892903223291283,\ 0.00892903223291283,\ 0.00892903223291283,\ 0.00892903223291283,\ 0.00892903223291283,\ 0.008929032232291283,\ 0.008929032232291283,\ 0.0089290321283,\ 0.0089290321283,\ 0.0089290321283,\ 0.0089290321283,\ 0.0089290321283,\ 0.0089290321283,\ 0.0089290321283,\ 0.0089290321283,\ 0.0089290321283,\ 0.0089290321283,\ 0.0089290321283,\ 0.0089290321283,\ 0.0089290321283,\ 0.0089290321283,\ 0.0089290321283,\ 0.0089290321283,\ 0.0089290321283,\ 0.0089290321283,\ 0.0089290321283,\ 0.0089290321283,\ 0.0089290321283,\ 0.0089290321283,\ 0.0089290321283,\ 0.0089290321283,\ 0.0089290321283,\ 0.0089290321283,\ 0.0089290321283,\ 0.0089290321283,\ 0.0089290321283,\ 0.0089290321283,\ 0.0089290321283,\ 0.0089290321283,\ 0.0089290321283,\ 0.0089290321283,\ 0.0089290321283,\ 0.0089290321283,\ 0.0089290321283,\ 0.0089290321283,\ 0.0089290321283,\ 0.0089290321283,\ 0.0089290321283,\ 0.0089290321283,\ 0.00892903
   iteration: 16
In [8]:
def visualize_targets_and_errors(targets, errors):
         x = range(len(targets))
         plt.plot(x, targets, label='targets')
         plt.plot(x, errors, label='errors')
         plt.legend(loc='best')
         plt.show()
In [9]:
visualize_targets_and_errors(targets, errors)
    12
    10
     8
                                                                                       targets
                                                                                      errors
     4
     2
```

这个target在上升(或下降)是因为,迭代过程中,并不满足 $\mathbf{1}^T x = 1$ , 目前的算法直接迭代到最优解附近,但是在迭代过程中不满足 $\mathbf{1}^T x = 1$ 。

注: 这是我的实现, 你的实现可能不会出现这样的情况。

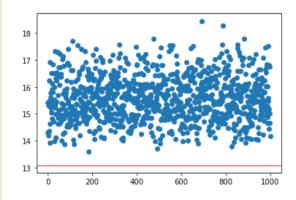
## 查看你的最优解

一顿操作之后,你可能想确认你的结果是不是正确的,你可能可以参考一些"标准答案", 但如果"标准答案"不好寻找的时 候,也可以简单的测试一下。

比如,我们的实现的算法应该比随机生成的解要好,我们来试试。

```
In [10]:
# let's play with monkey search, i.e., random search
def monkey search(alpha):
    # random return x s.t. 1^T x = 1 and x > 0
    while True:
        monkey_solution = np.random.dirichlet(np.ones(10),size=1).reshape(-1,1)
        if np.less(monkey_solution, 0).any():
            continue
        return monkey_solution
def visualize_monkey_search(alpha, monkey_amount, optimal):
    monkey_solutions = [calculate_target(alpha, monkey_search(alpha)) \
                        for _ in range(monkey_amount)]
    plt.scatter(range(monkey_amount), monkey_solutions)
    plt.axhline(y=optimal,linewidth=1, color='r')
    plt.show()
In [11]:
```

# you can verify that a monkey can never cross the optimal line optimal\_line = targets[len(targets)-1] visualize\_monkey\_search(alpha, 1000, optimal\_line)



1000只猴子敲不出莎士比亚。