

# 总结

关键: [ row , col ]

对于Dataframe, 如果相应**操控列col**, 直接df [ 'col' 名 ], 如果要**操作行row**, 则df.loc[ 'row'名 ]

iloc, 则是对matrix的index做处理。

## loc 与条件

还是关键 [ row , col ]

df.loc[ df.column\_name 条件 ]的意思是, 选择column\_name符合某些条件的row数据。

还可以基于多个列条件来筛选row数据。

## Numpy的局限

不能处理混合数据类型的数据, 这也是为什么我们要使用Pandas。

NumPy arrays are designed for numerical computation and **cannot easily handle collections of data that contain a mix of types** (e.g., strings, integers, and floats) in a single array.

A typical example of this limitation is a student attendance sheet. Each record might include a student's name (a string), their marks (an integer), and their student ID (which serves as a label).

NumPy arrays aren't well-suited for such cases because they can't store multiple types of data in the same array or use labels for indexing.

This is where **pandas** shines. pandas provides flexible and powerful tools to handle such **mixed-type datasets** and allows you to work with **labelled data**. It can store data in **DataFrames**, which are similar to tables in a database or Excel spreadsheet, making it ideal for data manipulation, analysis, and visualization tasks.

## Series and Dataframe

### Series

- 构造Series

### 字典传入

# pandas与Numpy的配合

## slicing

与Numpy一样

## series filtering

## DataFrame

### 构造

传入DataFrame的是一个array，或字典

传入字典，相当于定义col name。

字典对应的value必须得是array-like object

### index

### df取得指定值

### df 数据类型转换

传入字典，指定数据类型

### 常用的操作

- 更换index与赋值
- 数据类型转换

## Dataframe Operation

### NanN Not a number

缺失数据，统一用np.nan

原来可以直接放Series数据进series，并且缺失的数据，会自动以nan来填充。

### isnull()

若为nan，则为true.

### unique()

### value\_counts()

### isin()

倒过来读：**the feature\_name is in df**，若有则true，若无，则False

```
1 df.isin(['feature_name'])
```

配合使用any，**只要有一个匹配为true，则返回true**

### 用法

取特定数值的rows.

### sum()

## DataFrame Select/filtering data

.values：取值

.loc: 通过row与col选取数据

.iloc: 通过index 来选取数据

## loc

主要以选取row为主

记住一点, `loc [ row , col ], loc[ df[ 'col' ] < 某条件 ]`

若单纯取列, 则直接`[ ]`取即可。

Out[69]:

	a	b	c
one	1	4	7
two	2	5	8
three	3	6	9

```
In [76]: #取列  
df3['a']
```

Out[76]:

one	1
two	2
three	3

Name: a, dtype: int64

```
In [71]: #选取多行  
df3.loc[['one', 'three']]
```

Out[71]:

	a	b	c
one	1	4	7
three	3	6	9

## loc选取并赋值

```
In [81]: #选取元素，并赋值
df3.loc['one', 'c'] = 100
df3
```

Out[81]:

	a	b	c
one	1	4	100
two	2	5	8
three	3	6	9

## iloc

通过行列的index来获取对应的值。

```
In [45]: dataframe_11.iloc[0,1]='rainbow'
dataframe_11
```

Out[45]:

	object	colour
one	ball	rainbow
two	pen	green
three	pencil	yellow
four	paper	red
five	mug	black

## 选取特定的列 [[]]

```
In [36]: #多选几列
df_pop[['cities', 'density']]
```

Out[36]:

	cities	density
0	Sydney	4627345
1	Melbourne	4246375
2	Brisbane	2189878
3	Perth	1896548
4	Adelaide	1225235

## slicing

### loc, slicing with condition

筛选出，某列符合某条件的row

筛选符合条件的特定的列

### 例子2

### 筛选的例子

获得每一列的数据类型 dtype()

计算某列中各项数据的比例

# 替换数据

需要配合使用numpy

# 题目例子用法