Data transformation using .groupby().transform

OPTIMIZING PYTHON CODE WITH PANDAS





The restaurant dataset

```
restaurant_grouped = restaurant.groupby('smoker')
print(restaurant_grouped.count())
```



Data transformation

```
zscore = lambda x: (x - x.mean()) / x.std()
restaurant_grouped = restaurant.groupby('time')
restaurant_transformed = restaurant_grouped.transform(zscore)
restaurant_transformed.head()
  total_bill tip
                          size
   -0.416446 -1.457045 -0.692873
   -1.143855 -1.004475 0.405737
    0.023282 0.276645 0.405737
```



Comparison with native methods

```
restaurant.groupby('sex').transform(zscore)

mean_female = restaurant.groupby('sex').mean()['total_bill']['Female']
mean_male = restaurant.groupby('sex').mean()['total_bill']['Male']
std_female = restaurant.groupby('sex').std()['total_bill']['Female']
std_male = restaurant.groupby('sex').std()['total_bill']['Male']

for i in range(len(restaurant)):
    if restaurant.iloc[i][2] == 'Female':
        restaurant.iloc[i][0] = (restaurant.iloc[i][0] - mean_female)/std_female
    else:
        restaurant.iloc[i][0] = (restaurant.iloc[i][0] - mean_female)/std_female
```

```
Results from the first operation calculated in 0.016291141 seconds
Results from the second operation calculated in 3.937326908 seconds

Difference in time: 24,068.5145%
```



Let's practice!



Missing value imputation using transform()

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Counting missing values

Name: total_bill, dtype: int64

```
prior_counts = restaurant.groupby('time')
['total_bill'].count()
missing_counts = restaurant_nan.groupby('time')
['total_bill'].count()
print(prior_counts - missing_counts)
time
Dinner
          32
          13
Lunch
```

Missing value imputation

```
missing_trans = lambda x: x.fillna(x.mean())
restaurant_nan_grouped = restaurant_nan.groupby('time')['total_bill']
restaurant_nan_grouped.transform(missing_trans)
Results from the above operation calculated in 0.00368881225586 seconds
     20.676573
     10.340000
     21.010000
     23.680000
     24.590000
    25.290000
     20.676573
Name: total_bill, dtype: float64
```



Comparison with native methods

```
start_time = time.time()
mean_din = restaurant_nan.loc[restaurant_nan.time ==
'Dinner']['total_bill'].mean()
mean_lun = restaurant_nan.loc[restaurant_nan.time ==
'Lunch']['total_bill'].mean()

for row in range(len(restaurant_nan)):
    if restaurant_nan.iloc[row]['time'] == 'Dinner':
        restaurant_nan.loc[row, 'total_time'] = mean_din
    else:
        restaurant_nan.loc[row, 'total_time'] = mean_lun
print("Results from the above operation calculated in %s seconds" % (time.time() - start_time))
```

Results from the above operation calculated in 0.172566890717 seconds

```
Difference in time: 4,578.115%
```



Let's do it!



Data filtration using the filter() function

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Purpose of filter()

Limit results based on an aggregate feature

- Number of missing values
- Mean of a specific feature
- Number of occurrences of the group

Filter using groupby().filter()

```
restaurant_grouped = restaurant.groupby('day')
filter_trans = lambda x : x['total_bill'].mean() > 20
restaurant_filtered = restaurant_grouped.filter(filter_trans)
```

Results from the above operation calculated in 0.00414085388184 seconds

```
print(restaurant_filtered['tip'].mean())
```

3.11527607362

```
print(restaurant['<mark>tip'].me</mark>an())
```

2.9982786885245902



Comparison with native methods

```
t=[restaurant.loc[df['day'] == i]['tip'] for i in restaurant['day'].unique()
   if restaurant.loc[df['day'] == i]['total_bill'].mean()>20]
restaurant_filtered = t[0]
for j in t[1:]:
    restaurant_filtered=restaurant_filtered.append(j,ignore_index=True)
```

Results from the above operation calculated in 0.00663900375366 seconds

```
print(restaurant_filtered.mean())
```

3.11527607362

Difference in time: 60.329341317157024%



Let's do it!



Congratulations!

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What you have learned

- Why and how to time operations
- Select targeted rows and columns efficiently
- Select random rows and columns efficiently
- Replace values of a DataFrame efficiently using replace()
 - Replace multiple values using lists
 - Replace multiple values using dictionaries

What you have learned

- Iterate on a DataFrame using the .iterrows() function
- Iterate on a DataFrame using the .apply() function
- Iterate on a DataFrame using pandas optimization
- Iterate on a DataFrame using numpy optimization
- Comparison of the groupby() function compared to native python code
 - When transforming the data group-wise
 - When imputing missing values group-wise
 - When filtering groups with specific characteristics

Congratulations!

