

```
In [1]: # 機械学習 p. 64~73  
  
# 2.5 機械学習アルゴリズムが処理しやすいようにデータを準備する
```

```
In [2]: import pandas as pd  
import numpy as np  
import matplotlib.pyplot as plt  
  
from sklearn.model_selection import StratifiedShuffleSplit  
  
housing = pd.read_csv("datasets/housing/housing.csv")  
  
housing[“income_cat”] = pd.cut(housing[“median_income”],  
                               bins=[0.0, 1.5, 3.0, 4.5, 6.0, np.inf], labels=[1, 2, 3, 4, 5])  
split = StratifiedShuffleSplit(n_splits=1, test_size=0.2, random_state=42)  
  
for train_index, test_index in split.split(housing, housing[“income_cat”]):  
    strat_train_set = housing.loc[train_index]  
    strat_test_set = housing.loc[test_index]  
  
for set_ in (strat_train_set, strat_test_set):  
    set_.drop(“income_cat”, axis=1, inplace=True)  
  
housing = strat_train_set.drop(“median_house_value”, axis=1)  
housing_labels = strat_train_set[“median_house_value”].copy()
```

```
In [3]: # longitude latitude housing_median_age total_rooms total_bedrooms households median_income median_house_value ocean_proximity  
# 経度 緯度 築年数の中央値 部屋数 寝室数 世帯数 収入の中央値 住宅価格の中央値 海との位置関係
```

```
In [4]: # scikit-learn の SimpleImputer による欠損値の埋め込み  
  
housing_num = housing.drop(“ocean_proximity”, axis=1) # テキスト属性の除去  
  
from sklearn.impute import SimpleImputer # クラスのインポート  
  
imputer = SimpleImputer(strategy=“median”) # 「中央値埋め込む」インスタンスの作成  
  
imputer.fit(housing_num) # データにインスタンスを適合  
x = imputer.transform(housing_num) # 欠損値の埋め込み  
housing_num = pd.DataFrame(x, columns=housing_num.columns, index=housing_num.index) # DataFrameに戻す
```

```
In [5]: # 埋め込み実験（テストデータ）
test = pd.read_csv("datasets/test/test.csv")
test
```

```
Out[5]:   term-A  term-B  term-C  term-D  term-E
0       1      10    15.0     100   1000
1       2      20    NaN      200   2000
2       3      30    NaN      300   3000
3       4      40    NaN      400   4000
4       5      50    55.0     500   5000
```

```
In [6]: imputer.fit(test)
temp = imputer.transform(test)
test = pd.DataFrame(temp, columns=test.columns, index=test.index)
test
```

```
Out[6]:   term-A  term-B  term-C  term-D  term-E
0       1.0     10.0    15.0    100.0  1000.0
1       2.0     20.0    35.0    200.0  2000.0
2       3.0     30.0    35.0    300.0  3000.0
3       4.0     40.0    35.0    400.0  4000.0
4       5.0     50.0    55.0    500.0  5000.0
```

```
In [7]: # テキスト／カテゴリ属性の処理
housing_cat = housing[["ocean_proximity"]]
housing_cat.head(10)
```

```
Out[7]: ocean_proximity
17606 <1H OCEAN
18632 <1H OCEAN
14650 NEAR OCEAN
```

ocean_proximity

3230	INLAND
3555	<1H OCEAN
19480	INLAND
8879	<1H OCEAN
13685	INLAND
4937	<1H OCEAN
4861	<1H OCEAN

In [8]: # OrdinalEncoder

```
from sklearn.preprocessing import OrdinalEncoder
ordinal_encoder = OrdinalEncoder()
housing_cat_encoded = ordinal_encoder.fit_transform(housing_cat)
housing_cat_encoded[:10]
```

Out[8]: array([[0.],
... [0.],
... [4.],
... [1.],
... [0.],
... [1.],
... [0.],
... [1.],
... [0.],
... [0.]])

In [9]: ordinal_encoder.categories_

Out[9]: [array(['<1H OCEAN', 'INLAND', 'ISLAND', 'NEAR BAY', 'NEAR OCEAN'],
... dtype=object)]

In [10]: # OneHotEncoder

```
from sklearn.preprocessing import OneHotEncoder
cat_encoder = OneHotEncoder()
housing_cat_1hot = cat_encoder.fit_transform(housing_cat)
housing_cat_1hot
```

```
Out[10]: <16512x5 sparse matrix of type '<class 'numpy.float64'>'  
with 16512 stored elements in Compressed Sparse Row format>
```

```
In [11]: housing_cat_1hot.toarray()
```

```
Out[11]: array([[1., 0., 0., 0., 0.],  
... [1., 0., 0., 0., 0.],  
... [0., 0., 0., 0., 1.],  
... [0., 1., 0., 0., 0.],  
... [1., 0., 0., 0., 0.],  
... [0., 0., 0., 1., 0.]])
```

```
In [12]: # カスタム変換器 BaseEstimator, TransformerMixin
```

```
from sklearn.base import BaseEstimator, TransformerMixin  
  
rooms_ix, bedrooms_ix, population_ix, households_ix = 3, 4, 5, 6  
  
class CombinedAttributesAdder(BaseEstimator, TransformerMixin):  
  
    def __init__(self, add_bedrooms_per_room = True): # *args, **kargs なし  
        self.add_bedrooms_per_room = add_bedrooms_per_room  
  
    def fit(self, X, y=None):  
        return self # 特になし  
  
    def transform(self, X):  
        rooms_per_household = X[:, rooms_ix] / X[:, households_ix]  
        population_per_household = X[:, population_ix] / X[:, households_ix]  
        if self.add_bedrooms_per_room:  
            bedrooms_per_room = X[:, bedrooms_ix] / X[:, rooms_ix]  
            return np.c_[X, rooms_per_household, population_per_household, bedrooms_per_room]  
        else:  
            return np.c_[X, rooms_per_household, population_per_household]
```

```
In [13]: # 特徴量のスケーリング MinMaxScaler:最小最大スケーリング、StandardScaler:標準化
```

```
In [14]: # 変換パイプライン
```

```
from sklearn.pipeline import Pipeline  
from sklearn.preprocessing import StandardScaler  
  
num_pipeline = Pipeline([
```

```

('imputer', SimpleImputer(strategy="median")),
('attribs_adder', CombinedAttributesAdder()),
('std_scaler', StandardScaler()),
])

housing_num_tr = num_pipeline.fit_transform(housing_num)
housing_num_tr

```

```
Out[14]: array([[-1.15604281,  0.77194962,  0.74333089, ... , -0.31205452,
   ... , -0.08649871,  0.15531753],
   ... , [-1.17602483,  0.6596948 , -1.1653172 , ... ,  0.21768338,
   ... , -0.03353391, -0.83628902],
   ... , [ 1.18684903, -1.34218285,  0.18664186, ... , -0.46531516,
   ... , -0.09240499,  0.4222004 ],
   ... ,
   ... , [ 1.58648943, -0.72478134, -1.56295222, ... ,  0.3469342 ,
   ... , -0.03055414, -0.52177644],
   ... , [ 0.78221312, -0.85106801,  0.18664186, ... ,  0.02499488,
   ... ,  0.06150916, -0.30340741],
   ... , [-1.43579109,  0.99645926,  1.85670895, ... , -0.22852947,
   ... , -0.09586294,  0.10180567]])
```

```
In [15]: # ColumnTransformer

from sklearn.compose import ColumnTransformer

num_attribs = list(housing_num)
cat_attribs = ["ocean_proximity"]

full_pipeline = ColumnTransformer([
    ("num", num_pipeline, num_attribs),
    ("cat", OneHotEncoder(), cat_attribs),
])
housing_prepared = full_pipeline.fit_transform(housing)
housing_prepared
```

```
Out[15]: array([[-1.15604281,  0.77194962,  0.74333089, ... , 0. .... ,
   ... , 0. .... ],
   ... , [-1.17602483,  0.6596948 , -1.1653172 , ... , 0. .... ,
   ... , 0. .... ],
   ... , [ 1.18684903, -1.34218285,  0.18664186, ... , 0. .... ,
   ... , 0. .... ],
   ... ,
   ... , [ 1.58648943, -0.72478134, -1.56295222, ... , 0. .... ,
   ... , 0. .... ],
   ... , [ 0.78221312, -0.85106801,  0.18664186, ... , 0. .... ,
   ... , 0. .... ],
```

```
[ -1.43579109,  0.99645926,  1.85670895,  ...,  0.          ,
  1.          ,  0.          ]])
```