

TransactionEncoder: Convert item lists into transaction data for frequent itemset mining

Encoder class for transaction data in Python lists

```
from mlxtend.preprocessing import TransactionEncoder
```

Overview

Encodes database transaction data in form of a Python list of lists into a NumPy array.

Example 1

Suppose we have the following transaction data:

```
from mlxtend.preprocessing import TransactionEncoder

dataset = [['Apple', 'Beer', 'Rice', 'Chicken'],
           ['Apple', 'Beer', 'Rice'],
           ['Apple', 'Beer'],
           ['Apple', 'Bananas'],
           ['Milk', 'Beer', 'Rice', 'Chicken'],
           ['Milk', 'Beer', 'Rice'],
           ['Milk', 'Beer'],
           ['Apple', 'Bananas']]
```

Using and `TransactionEncoder` object, we can transform this dataset into an array format suitable for typical machine learning APIs. Via the `fit` method, the `TransactionEncoder` learns the unique labels in the dataset, and via the `transform` method, it transforms the input dataset (a Python list of lists) into a one-hot encoded NumPy boolean array:

```
te = TransactionEncoder()
te_ary = te.fit(dataset).transform(dataset)
te_ary
```

```
array([[ True, False,  True,  True, False,  True],
       [ True, False,  True, False, False,  True],
       [ True, False,  True, False, False, False],
       [ True,  True, False, False, False, False],
       [False, False,  True,  True,  True,  True],
       [False, False,  True, False,  True,  True],
       [False, False,  True, False,  True, False],
       [ True,  True, False, False, False, False]], dtype=bool)
```

The NumPy array is boolean for the sake of memory efficiency when working with large datasets. If a classic integer representation is desired instead, we can just convert the array to the appropriate type:

```
te_ary.astype("int")
```

```
array([[1, 0, 1, 1, 0, 1],
       [1, 0, 1, 0, 0, 1],
       [1, 0, 1, 0, 0, 0],
       [1, 1, 0, 0, 0, 0],
       [0, 0, 1, 1, 1, 1],
       [0, 0, 1, 0, 1, 1],
       [0, 0, 1, 0, 1, 0],
       [1, 1, 0, 0, 0, 0]])
```

After fitting, the unique column names that correspond to the data array shown above can be accessed via the `columns_` attribute:

```
te.columns_
```

```
['Apple', 'Bananas', 'Beer', 'Chicken', 'Milk', 'Rice']
```

For our convenience, we can turn the encoded array into a pandas `DataFrame` :

```
import pandas as pd

pd.DataFrame(te_ary, columns=te.columns_)
```

	Apple	Bananas	Beer	Chicken	Milk	Rice
0	True	False	True	True	False	True
1	True	False	True	False	False	True
2	True	False	True	False	False	False
3	True	True	False	False	False	False
4	False	False	True	True	True	True
5	False	False	True	False	True	True
6	False	False	True	False	True	False
7	True	True	False	False	False	False

If we desire, we can turn the one-hot encoded array back into a transaction list of lists via the `inverse_transform` function:

```
first4 = te_ary[:4]
te.inverse_transform(first4)
```

```
[['Apple', 'Beer', 'Chicken', 'Rice'],
 ['Apple', 'Beer', 'Rice'],
 ['Apple', 'Beer'],
 ['Apple', 'Bananas']]
```

API

TransactionEncoder()

Encoder class for transaction data in Python lists

Parameters

None

Attributes

`columns_`: list List of unique names in the `x` input list of lists

Examples

For usage examples, please see https://rasbt.github.io/mlxtend/user_guide/preprocessing/TransactionEncoder/

Methods

fit(X)

Learn unique column names from transaction DataFrame

Parameters

- `x` : list of lists

A python list of lists, where the outer list stores the n transactions and the inner list stores the items in each transaction.

For example, `[['Apple', 'Beer', 'Rice', 'Chicken'], ['Apple', 'Beer', 'Rice'], ['Apple', 'Beer'], ['Apple', 'Bananas'], ['Milk', 'Beer', 'Rice', 'Chicken'], ['Milk', 'Beer', 'Rice'], ['Milk', 'Beer'], ['Apple', 'Bananas']]`

fit_transform(X, sparse=False)

Fit a TransactionEncoder encoder and transform a dataset.

get_params(deep=True)

Get parameters for this estimator.

Parameters

- `deep` : boolean, optional

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If True, will return the parameters for this estimator and contained subobjects that are estimators.

Returns

- `params` : mapping of string to any

Parameter names mapped to their values.

inverse_transform(array)

Transforms an encoded NumPy array back into transactions.

Parameters

- `array` : NumPy array [n_transactions, n_unique_items]

The NumPy one-hot encoded boolean array of the input transactions, where the columns represent the unique items found in the input array in alphabetic order

For example,

```
array([[True , False, True , True , False, True ],
       [True , False, True , False, False, True ],
       [True , False, True , False, False, False],
       [True , True , False, False, False, False],
       [False, False, True , True , True , True ],
       [False, False, True , False, True , True ],
       [False, False, True , False, True , False],
       [True , True , False, False, False, False]])
```

The corresponding column labels are available as `self.columns_`, e.g., `['Apple', 'Bananas', 'Beer', 'Chicken', 'Milk', 'Rice']`

Returns

- `X` : list of lists

A python list of lists, where the outer list stores the n transactions and the inner list stores the items in each transaction.

For example,

```
[['Apple', 'Beer', 'Rice', 'Chicken'],
 ['Apple', 'Beer', 'Rice'],
 ['Apple', 'Beer'],
 ['Apple', 'Bananas'],
 ['Milk', 'Beer', 'Rice', 'Chicken'],
 ['Milk', 'Beer', 'Rice'],
 ['Milk', 'Beer'],
 ['Apple', 'Bananas']]
```

set_params(params)

Set the parameters of this estimator.

The method works on simple estimators as well as on nested objects (such as pipelines). The latter have parameters of the form `<component>__<parameter>` so that it's possible to update each component of a nested object.

Returns

self

transform(X, sparse=False)

Transform transactions into a one-hot encoded NumPy array.

Parameters

- `X` : list of lists

A python list of lists, where the outer list stores the n transactions and the inner list stores the items in each transaction.

For example, `[['Apple', 'Beer', 'Rice', 'Chicken'], ['Apple', 'Beer', 'Rice'], ['Apple', 'Beer'], ['Apple', 'Bananas'], ['Milk', 'Beer', 'Rice', 'Chicken'], ['Milk', 'Beer', 'Rice'], ['Milk', 'Beer'], ['Apple', 'Bananas']]`

`sparse`: bool (default=False) If True, transform will return Compressed Sparse Row matrix instead of the regular one.

Returns

- `array` : NumPy array [n_transactions, n_unique_items]

if `sparse=False` (default). Compressed Sparse Row matrix otherwise The one-hot encoded boolean array of the input transactions, where the columns represent the unique items found in the input array in alphabetic order. Exact representation depends on the `sparse` argument

For example, `array([[True , False, True , True , False, True], [True , False, True , False, False, True], [True , False, True , False, False, False], [True , True , False, False, False, False], [False, False, True , True , True , True], [False, False, True , False, True , True], [False, False, True , False, True , False], [True , True , False, False, False, False]])`

The corresponding column labels are available as `self.columns_`, e.g., `['Apple', 'Bananas', 'Beer', 'Chicken', 'Milk', 'Rice']`

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