



A common problem when creating models to generate business value from data is that the datasets can be so large that it can take days for the model to generate predictions. Ensuring that your dataset is stored as efficiently as possible is crucial for allowing these models to run on a more reasonable timescale without having to reduce the size of the dataset.

You've been hired by a major online data science training provider called *Training Data Ltd.* to clean up one of their largest customer datasets. This dataset will eventually be used to predict whether their students are looking for a new job or not, information that they will then use to direct them to prospective recruiters.

You've been given access to `customer_train.csv`, which is a subset of their entire customer dataset, so you can create a proof-of-concept of a much more efficient storage solution. The dataset contains anonymized student information, and whether they were looking for a new job or not during training:

Column	Description
<code>student_id</code>	A unique ID for each student.
<code>city</code>	A code for the city the student lives in.
<code>city_development_index</code>	A scaled development index for the city.
<code>gender</code>	The student's gender.
<code>relevant_experience</code>	An indicator of the student's work relevant experience.
<code>enrolled_university</code>	The type of university course enrolled in (if any).
<code>education_level</code>	The student's education level.
<code>major_discipline</code>	The educational discipline of the student.
<code>experience</code>	The student's total work experience (in years).
<code>company_size</code>	The number of employees at the student's current employer.
<code>company_type</code>	The type of company employing the student.
<code>last_new_job</code>	The number of years between the student's current and previous jobs.
<code>training_hours</code>	The number of hours of training completed.
<code>job_change</code>	An indicator of whether the student is looking for a new job ( <code>1</code> ) or not ( <code>0</code> ).

```
# Import necessary libraries
import pandas as pd

# Load the dataset
ds_jobs = pd.read_csv("customer_train.csv")

# View the dataset
ds_jobs.head()
```

...	↑↓	s...	...	↑↓	...	↑↓	city_development_ind...	...	↑↓	...	↑↓	relevant_experience	...	↑↓	enrolled_universi...	...	↑↓	education...
	0		8949		city_103				0.92		Male	Has relevant experience			no_enrollment			Graduate
	1		29725		city_40				0.776		Male	No relevant experience			no_enrollment			Graduate
	2		11561		city_21				0.624		null	No relevant experience			Full time course			Graduate
	3		33241		city_115				0.789		null	No relevant experience			null			Graduate
	4		666		city_162				0.767		Male	Has relevant experience			no_enrollment			Masters

Rows: 5

Expand

```
# Create a copy of ds_jobs for transforming
ds_jobs_transformed = ds_jobs.copy()
```

## Store the data much more efficiently

You have to store the data in `ds_jobs_transformed`, following the requirements

1. Columns containing categories with two factors must be stores as booleans
2. Columns containing integers only must be store as int32
3. Columns containing floats must be stored as float16
4. Columns containing nominal categorical data must be stored as category dtype
5. Columns with ordinal data must be stored as an ordered category
6. Df should be filtered to only students with 10 or more years of experience at companies with at least 1000 employees.

```
ds_jobs_transformed.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 19158 entries, 0 to 19157
Data columns (total 14 columns):
 #   Column                Non-Null Count  Dtype
---  -
 0   student_id            19158 non-null  int64
 1   city                  19158 non-null  object
 2   city_development_index 19158 non-null  float64
 3   gender                14650 non-null  object
 4   relevant_experience    19158 non-null  object
 5   enrolled_university   18772 non-null  object
 6   education_level        18698 non-null  object
 7   major_discipline      16345 non-null  object
 8   experience             19093 non-null  object
 9   company_size           13220 non-null  object
10   company_type           13018 non-null  object
11   last_new_job           18735 non-null  object
12   training_hours         19158 non-null  int64
13   job_change             19158 non-null  float64
dtypes: float64(2), int64(2), object(10)
memory usage: 2.0+ MB
```

### Columns containing categories with two factors must be stores as booleans

The columns with just two categories are:

- `job_change`
- `relevant_experience`

```
ds_jobs_transformed['job_change'].value_counts(dropna = False)

# Converting the column to boolean

ds_jobs_transformed.loc[ds_jobs_transformed['job_change'] == 1, 'job_change'] = True
ds_jobs_transformed.loc[ds_jobs_transformed['job_change'] == 0, 'job_change'] = False

ds_jobs_transformed.loc[ds_jobs_transformed['relevant_experience'] == 'Has relevant experience', 'relevant_experience'] = True
ds_jobs_transformed.loc[ds_jobs_transformed['relevant_experience'] == 'No relevant experience', 'relevant_experience'] = False

ds_jobs_transformed['job_change'] = ds_jobs_transformed['job_change'].astype('bool')
ds_jobs_transformed['relevant_experience'] = ds_jobs_transformed['relevant_experience'].astype('bool')
```

```
ds_jobs_transformed['job_change'].value_counts(dropna = False)
```

...	↑↓	j...	...	↑↓
False		14381		
True		4777		

Rows: 2

[↗ Expand](#)

```
ds_jobs_transformed['relevant_experience'].value_counts(dropna = False)
```

...	↑↓	relevant_experie...	...	↑↓
True		13792		
False		5366		

Rows: 2

[↗ Expand](#)

## Columns containing integers only must be store as int32

The columns with integers are:

- student\_id
- training\_hours

```
# Converting the columns to dtype int32
```

```
ds_jobs_transformed['student_id'] = ds_jobs_transformed['student_id'].astype('int32')
ds_jobs_transformed['training_hours'] = ds_jobs_transformed['training_hours'].astype('int32')
```

```
ds_jobs_transformed.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 19158 entries, 0 to 19157
Data columns (total 14 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   student_id                            19158 non-null  int32
1   city                                  19158 non-null  object
2   city_development_index                19158 non-null  float64
3   gender                                14650 non-null  object
4   relevant_experience                    19158 non-null  bool
5   enrolled_university                  18772 non-null  object
6   education_level                       18698 non-null  object
7   major_discipline                      16345 non-null  object
8   experience                            19093 non-null  object
9   company_size                          13220 non-null  object
10  company_type                          13018 non-null  object
11  last_new_job                          18735 non-null  object
12  training_hours                        19158 non-null  int32
13  job_change                            19158 non-null  bool
dtypes: bool(2), float64(1), int32(2), object(9)
memory usage: 1.6+ MB
```

### Columns containing floats must be stored as float16

The columns with floats are:

- city\_development\_index

```
ds_jobs_transformed['city_development_index'] = ds_jobs_transformed['city_development_index'].astype('float16')
```

```
ds_jobs_transformed.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 19158 entries, 0 to 19157
Data columns (total 14 columns):
#   Column                Non-Null Count  Dtype
---  -
0   student_id            19158 non-null  int32
1   city                  19158 non-null  object
2   city_development_index 19158 non-null  float16
3   gender                14650 non-null  object
4   relevant_experience    19158 non-null  bool
5   enrolled_university   18772 non-null  object
6   education_level        18698 non-null  object
7   major_discipline       16345 non-null  object
8   experience             19093 non-null  object
9   company_size           13220 non-null  object
10  company_type           13018 non-null  object
11  last_new_job            18735 non-null  object
12  training_hours          19158 non-null  int32
13  job_change             19158 non-null  bool
dtypes: bool(2), float16(1), int32(2), object(9)
memory usage: 1.5+ MB
```

### Columns containing nominal categorical data must be stored as category dtype

The columns with nominal categorical data are:

- city
- gender
- major\_discipline

```
ds_jobs_transformed[['city', 'gender', 'major_discipline', 'company_type']] = ds_jobs_transformed[['city', 'gender', 'major_discipline', 'company_type']].astype('category')
```

```
ds_jobs_transformed.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 19158 entries, 0 to 19157
Data columns (total 14 columns):
#   Column                Non-Null Count  Dtype
---  -
0   student_id            19158 non-null  int32
1   city                  19158 non-null  category
2   city_development_index 19158 non-null  float16
3   gender                14650 non-null  category
4   relevant_experience    19158 non-null  bool
5   enrolled_university   18772 non-null  object
6   education_level        18698 non-null  object
7   major_discipline       16345 non-null  category
8   experience             19093 non-null  object
9   company_size           13220 non-null  object
10  company_type           13018 non-null  category
11  last_new_job            18735 non-null  object
12  training_hours          19158 non-null  int32
13  job_change             19158 non-null  bool
dtypes: bool(2), category(4), float16(1), int32(2), object(5)
memory usage: 1.0+ MB
```

### Columns with ordinal data must be stored as an ordered category

The columns with ordinal categorical data are:

- last\_new\_job
- experience
- education\_level
- enrolled\_university
- company\_size

```
ds_jobs_transformed['enrolled_university'].unique()
```

```
array(['no_enrollment', 'Full time course', nan, 'Part time course'],  
      dtype=object)
```

```
ds_jobs_transformed[['last_new_job', 'experience', 'education_level', 'enrolled_university', 'company_size']] =  
ds_jobs_transformed[['last_new_job', 'experience', 'education_level', 'enrolled_university', 'company_size']].astype('category')
```

```

# last new job
ds_jobs_transformed['last_new_job'] = ds_jobs_transformed['last_new_job'].cat.set_categories(
    new_categories = ['never', '1', '2', '3', '4',
                     '>4'],
    ordered = True) #Nan

# experience
ds_jobs_transformed['experience'] = ds_jobs_transformed['experience'].cat.set_categories(
    new_categories = ['<1', '1', '2', '3', '4', '5', '6',
                     '7',
                     '8', '9', '10', '11', '12', '13', '14',
                     '15',
                     '16', '17', '18', '19', '20', '>20'],
    ordered = True) #Nan

# education_level
ds_jobs_transformed['education_level'] = ds_jobs_transformed['education_level'].cat.set_categories(
    new_categories = ['Primary School', 'High School',
                     'Graduate', 'Masters', 'Phd'],
    ordered = True) #Nan

# enrolled_university
ds_jobs_transformed['company_size'] = ds_jobs_transformed['company_size'].cat.set_categories(
    new_categories = ['<10', '10-49',
                     '50-99', '100-499', '1000-4999',
                     '5000-9999', '10000+'],
    ordered = True) #Nan

# company_size
ds_jobs_transformed['enrolled_university'] = ds_jobs_transformed['enrolled_university'].cat.set_categories(
    new_categories = ['no_enrollment', 'Part time
course',
                     'Full time course'],
    ordered = True) #Nan

ds_jobs_transformed.info()

```

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 19158 entries, 0 to 19157
Data columns (total 14 columns):
#   Column                Non-Null Count  Dtype
---  -
0   student_id            19158 non-null  int32
1   city                  19158 non-null  category
2   city_development_index 19158 non-null  float16
3   gender                14650 non-null  category
4   relevant_experience    19158 non-null  bool
5   enrolled_university   18772 non-null  category
6   education_level       18698 non-null  category
7   major_discipline      16345 non-null  category
8   experience            19093 non-null  category
9   company_size          12343 non-null  category
10  company_type          13018 non-null  category
11  last_new_job          18735 non-null  category
12  training_hours        19158 non-null  int32
13  job_change            19158 non-null  bool
dtypes: bool(2), category(9), float16(1), int32(2)
memory usage: 400.2 KB

```

**Df should be filtered to only students with 10 or more years of experience at companies with at least 1000 employees.**

```

# filtering the dataset
filter_exp = ['10', '11', '12', '13', '14', '15', '16', '17', '18', '19', '20', '>20']
filter_comp = ['1000-4999', '5000-9999', '10000+']

ds_jobs_transformed = ds_jobs_transformed[(ds_jobs_transformed['experience'].isin(filter_exp)) &
(ds_jobs_transformed['company_size'].isin(filter_comp))]

```

ds\_jobs\_transformed

...	↑↓	s...	...	↑↓	...	↑↓	city_development_ind...	...	↑↓	...	↑↓	relevant_experie...	...	↑↓	enrolled_universi...	...	↑↓	education...	...
9		699			city_103		0.919921875					True			no_enrollment			Graduate	
12		25619			city_61		0.9130859375		Male			True			no_enrollment			Graduate	
31		22293			city_103		0.919921875		Male			True			Part time course			Graduate	
34		26494			city_16		0.91015625		Male			True			no_enrollment			Graduate	
40		2547			city_114		0.92578125		Female			True			Full time course			Masters	
47		25987			city_103		0.919921875		Other			True			no_enrollment			Graduate	
104		1180			city_16		0.91015625		Male			True			no_enrollment			Graduate	
108		25349			city_16		0.91015625		Male			True			no_enrollment			Graduate	
115		20576			city_97		0.9248046875		Male			True			no_enrollment			Graduate	
130		3921			city_36		0.8930664062					False			no_enrollment			Phd	
144		24796			city_103		0.919921875		Male			True			no_enrollment			Graduate	
146		22718			city_157		0.7690429688					True			no_enrollment			Graduate	
154		12154			city_16		0.91015625					True			no_enrollment			Masters	
157		28817			city_89		0.9248046875		Male			True			Full time course			Graduate	
159		7280			city_103		0.919921875		Male			True			no_enrollment			Graduate	
160		16903			city_103		0.919921875		Male			True			no_enrollment			Graduate	

Rows: 2,201

[Expand](#)

ds\_jobs\_transformed.shape

(2201, 14)

# Final information of the data

ds\_jobs\_transformed.info()

```

<class 'pandas.core.frame.DataFrame'>
Int64Index: 2201 entries, 9 to 19143
Data columns (total 14 columns):
#   Column                Non-Null Count  Dtype
---  -
0   student_id            2201 non-null  int32
1   city                  2201 non-null  category
2   city_development_index 2201 non-null  float16
3   gender                1821 non-null  category
4   relevant_experience    2201 non-null  bool
5   enrolled_university   2185 non-null  category
6   education_level       2184 non-null  category
7   major_discipline      2097 non-null  category
8   experience            2201 non-null  category
9   company_size          2201 non-null  category
10  company_type          2144 non-null  category
11  last_new_job          2184 non-null  category
12  training_hours        2201 non-null  int32
13  job_change            2201 non-null  bool
dtypes: bool(2), category(9), float16(1), int32(2)
memory usage: 134.1 KB

```

