

# e\_commerce

November 11, 2018

## 1 CASE STUDY: E\_COMMERCE DELIVERY

Here we are looking at the data obtained from an online delivery store, that delivers customized apparels to customers. Given various details such as the order details, factory details, expected\_delivery\_date, we need to predict whether the product would be delivered on time or not.

### 1.1 IMPORTS

```
In [480]: import pandas as pd
          from datetime import timedelta
          import matplotlib.pyplot as plt
          from tqdm import tqdm_notebook as tqdm
          import numpy as np
          import seaborn as sns
          from statsmodels.stats.outliers_influence import variance_inflation_factor
          from sklearn.model_selection import train_test_split
          from sklearn.linear_model import LogisticRegression
          from sklearn.svm import SVC, LinearSVC
          from sklearn.metrics import accuracy_score
          from sklearn.model_selection import cross_val_score
          from sklearn.metrics import confusion_matrix
          from sklearn.preprocessing import StandardScaler
          from sklearn.decomposition import PCA
          from sklearn.preprocessing import scale
          %matplotlib inline
```

### 1.2 READING THE DATA

We will work with the e\_commerce data provided by the company. The data has 79,999 rows and 19 columns. It has variables the following : - Order ID - Country - Shipping Method - Units per Order - Facility - Product Category - On Sale - Transit days - Datetime Ordered - Datetime Sourced - Datetime Product Ready - Datetime Planned - Deadline Source - Deadline Make - Deadline Deliver - Delivered to Plan

The below code reads the data from excel file using python package pandas

```
In [431]: data=pd.read_excel(r"E:\Sem5\MS4110(Data_Analytics)\E-Commerce_Delivery (1)\E-Commerce
```

Check the head of the loaded data, and check out its info() and describe() methods.

```
In [432]: data.head()
```

```
Out[432]:
```

	order_id	country	shipping_method	units_per_order	facility	\
0	E00000001	UNITED KINGDOM	Ground	1	OXFORD	
1	E00000002	FRANCE	Ground	1	ANTWERP	
2	E00000003	FRANCE	Ground	1	ANTWERP	
3	E00000004	FRANCE	Ground	1	ANTWERP	
4	E00000005	UNITED KINGDOM	Next Day	1	OXFORD	

  

	product_category	on_sale	transit_days	datetime_ordered	\
0	ACCESSORIES	Y	2	2016-07-03 03:07:29	
1	JACKETS & VESTS	N	3	2016-07-03 00:08:43	
2	TOPS	Y	3	2016-07-03 00:36:00	
3	JACKETS & VESTS	Y	5	2016-07-03 00:47:45	
4	JACKETS & VESTS	Y	1	2016-07-03 03:52:13	

  

	datetime_sourced	datetime_product_ready	datetime_planned	\
0	2016-07-03 04:09:49	2016-07-06 00:59:42	2016-07-08	
1	2016-07-03 02:16:24	2016-07-03 07:17:04	2016-07-06	
2	2016-07-03 02:16:18	2016-07-03 06:16:57	2016-07-06	
3	2016-07-03 02:16:18	2016-07-03 06:16:55	2016-07-07	
4	2016-07-03 07:56:33	2016-07-05 08:53:19	2016-07-06	

  

	deadline_source	deadline_make	deadline_deliver	delivered_to_plan
0	1612.0	38	3	PASS
1	531.0	8	3	PASS
2	504.0	8	3	PASS
3	492.0	8	5	PASS
4	1567.0	6	1	PASS

### Dimensions of the data

```
In [433]: print(data.shape)
```

```
(79999, 16)
```

We can see the number of unique categories in the categorical data by using data.describe

```
In [434]: data.describe(include='all')
```

```
Out[434]:
```

	order_id	country	shipping_method	units_per_order	facility	\
count	79999	79999	79999	79999.000000	79999	
unique	79999	5	4	NaN	6	
top	E00031961	GERMANY	Ground	NaN	AUGSBURG	
freq	1	46259	55494	NaN	19080	
first	NaN	NaN	NaN	NaN	NaN	

last	NaN	NaN	NaN	NaN	NaN
mean	NaN	NaN	NaN	1.040013	NaN
std	NaN	NaN	NaN	0.466923	NaN
min	NaN	NaN	NaN	1.000000	NaN
25%	NaN	NaN	NaN	1.000000	NaN
50%	NaN	NaN	NaN	1.000000	NaN
75%	NaN	NaN	NaN	1.000000	NaN
max	NaN	NaN	NaN	50.000000	NaN

	product_category	on_sale	transit_days	datetime_ordered	\
count	79999	79999	79999.000000	79999	
unique	3	2	NaN	78681	
top	JACKETS & VESTS	Y	NaN	2016-08-20 17:03:18	
freq	48523	41168	NaN	8	
first	NaN	NaN	NaN	2016-07-03 00:08:43	
last	NaN	NaN	NaN	2016-10-21 17:55:30	
mean	NaN	NaN	2.338067	NaN	
std	NaN	NaN	0.992345	NaN	
min	NaN	NaN	1.000000	NaN	
25%	NaN	NaN	2.000000	NaN	
50%	NaN	NaN	2.000000	NaN	
75%	NaN	NaN	3.000000	NaN	
max	NaN	NaN	8.000000	NaN	

	datetime_sourced	datetime_product_ready	datetime_planned	\
count	79999	79999	79999	
unique	72611	77461	127	
top	2016-09-16 08:39:53	2016-07-11 11:38:10	2016-09-30 00:00:00	
freq	6	4	1539	
first	2016-07-03 02:16:18	2016-07-03 06:16:55	2016-07-05 00:00:00	
last	2016-11-10 00:33:50	2016-11-10 19:30:18	2016-11-17 00:00:00	
mean	NaN	NaN	NaN	
std	NaN	NaN	NaN	
min	NaN	NaN	NaN	
25%	NaN	NaN	NaN	
50%	NaN	NaN	NaN	
75%	NaN	NaN	NaN	
max	NaN	NaN	NaN	

	deadline_source	deadline_make	deadline_deliver	delivered_to_plan
count	79464.000000	79999.000000	79999.000000	79999
unique	NaN	NaN	NaN	2
top	NaN	NaN	NaN	PASS
freq	NaN	NaN	NaN	70978
first	NaN	NaN	NaN	NaN
last	NaN	NaN	NaN	NaN
mean	1349.708975	41.173127	3.450756	NaN
std	1104.229891	25.686936	1.617486	NaN

min	1.000000	1.000000	1.000000	NaN
25%	484.000000	23.000000	2.000000	NaN
50%	1115.000000	33.000000	3.000000	NaN
75%	1600.250000	55.000000	5.000000	NaN
max	6392.000000	168.000000	13.000000	NaN

In [435]: data.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 79999 entries, 0 to 79998
Data columns (total 16 columns):
order_id          79999 non-null object
country           79999 non-null object
shipping_method   79999 non-null object
units_per_order   79999 non-null int64
facility           79999 non-null object
product_category  79999 non-null object
on_sale           79999 non-null object
transit_days       79999 non-null int64
datetime_ordered  79999 non-null datetime64[ns]
datetime_sourced   79999 non-null datetime64[ns]
datetime_product_ready 79999 non-null datetime64[ns]
datetime_planned   79999 non-null datetime64[ns]
deadline_source    79464 non-null float64
deadline_make      79999 non-null int64
deadline_deliver   79999 non-null int64
delivered_to_plan  79999 non-null object
dtypes: datetime64[ns](4), float64(1), int64(4), object(7)
memory usage: 9.8+ MB
```

### 1.3 CREATING NEW COLUMNS THAT RECORD THE TIME INTERVAL BETWEEN DATES

We create new columns to record the time intervals between the various dates and the time ordered as this is easier to work with and captures the necessary information from this data. \_\_Old variables\_\_ are being replaced by **new variables** here.

**Old variables - Datetime Ordered - Datetime Sourced - Datetime Product Ready - Datetime Planned**

**New Variables - Time Sourced:** Time taken to take the product to work station - **Time Make**: Time taken to make the product at the work station - **Time Ready:** Total time taken to make the product ready from the time it has been ordered

In [436]: *#converting the given dates to time intervals which can be worked with. check the lambda function for conversion to seconds minutes or days*

```
data['time_sourced']=(data.datetime_sourced-data.datetime_ordered).map(lambda x:int(x))
data['time_make']=(data.datetime_product_ready-data.datetime_ordered).map(lambda x:int(x))
data['time_ready']=(data.datetime_product_ready-data.datetime_ordered).map(lambda x:int(x))
```

## 1.4 IMPUTING NULL VALUES INTO THE DATA

Here we check for the null value in the data

```
In [437]: for i in data:
          print(i,data[i].isnull().values.any(),sep='\t\t\t')
```

```
order_id          False
country           False
shipping_method    False
units_per_order    False
facility           False
product_category   False
on_sale           False
transit_days       False
datetime_ordered   False
datetime_sourced   False
datetime_product_ready False
datetime_planned   False
deadline_source     True
deadline_make       False
deadline_deliver    False
delivered_to_plan   False
time_sourced        False
time_make          False
time_ready         False
```

We find out that only the deadline source column has null value. The code below is used to find the number of rows which have null values

```
In [438]: data[data.isnull().values]
```

```
Out[438]:
```

	order_id	country	shipping_method	units_per_order	facility	\
25	E00000026	FRANCE	Ground	1	ANTWERP	
34	E00000035	FRANCE	Ground	1	ANTWERP	
37	E00000038	FRANCE	Ground	1	ANTWERP	
39	E00000040	FRANCE	Ground	1	ANTWERP	
40	E00000041	FRANCE	Ground	1	ANTWERP	
41	E00000042	FRANCE	Ground	1	ANTWERP	
42	E00000043	FRANCE	Ground	1	ANTWERP	
43	E00000044	FRANCE	Ground	1	ANTWERP	
44	E00000045	FRANCE	Ground	1	ANTWERP	
45	E00000046	FRANCE	Ground	1	ANTWERP	
47	E00000048	FRANCE	Ground	1	ANTWERP	
48	E00000049	FRANCE	Ground	1	ANTWERP	
49	E00000050	FRANCE	Ground	1	ANTWERP	
52	E00000053	FRANCE	Ground	1	ANTWERP	

53	E00000054	FRANCE	Ground	1	ANTWERP
54	E00000055	FRANCE	Ground	1	ANTWERP
55	E00000056	FRANCE	Ground	1	ANTWERP
58	E00000059	FRANCE	Ground	1	ANTWERP
60	E00000061	FRANCE	Ground	1	ANTWERP
65	E00000066	FRANCE	Ground	1	ANTWERP
70	E00000071	FRANCE	Ground	1	ANTWERP
77	E00000078	FRANCE	Ground	1	ANTWERP
78	E00000079	FRANCE	Ground	1	ANTWERP
83	E00000084	FRANCE	Ground	1	ANTWERP
97	E00000098	FRANCE	Ground	1	ANTWERP
104	E00000105	FRANCE	Ground	1	ANTWERP
111	E00000112	GERMANY	Ground	1	EINDHOVEN
115	E00000116	FRANCE	Ground	1	ANTWERP
116	E00000117	FRANCE	Ground	1	ANTWERP
118	E00000119	UNITED KINGDOM	Next Day	1	OXFORD
...	...	...	...	...	...
75449	E00075531	UNITED KINGDOM	Ground	1	MANCHESTER
75464	E00075546	UNITED KINGDOM	Ground	1	MANCHESTER
75465	E00075547	UNITED KINGDOM	Ground	1	MANCHESTER
75471	E00075553	UNITED KINGDOM	Ground	1	OXFORD
75476	E00075558	UNITED KINGDOM	Ground	1	OXFORD
75479	E00075561	UNITED KINGDOM	Ground	1	OXFORD
75484	E00075566	UNITED KINGDOM	Ground	1	OXFORD
75488	E00075570	UNITED KINGDOM	Ground	1	MANCHESTER
75498	E00075580	UNITED KINGDOM	Ground	1	MANCHESTER
75500	E00075582	UNITED KINGDOM	Ground	1	OXFORD
75502	E00075584	UNITED KINGDOM	Ground	1	OXFORD
75503	E00075585	UNITED KINGDOM	Ground	1	MANCHESTER
75507	E00075589	UNITED KINGDOM	Ground	1	OXFORD
75511	E00075593	UNITED KINGDOM	Ground	2	OXFORD
75513	E00075595	UNITED KINGDOM	Ground	1	MANCHESTER
75526	E00075608	UNITED KINGDOM	Ground	1	OXFORD
75529	E00075611	UNITED KINGDOM	Ground	1	OXFORD
75544	E00075626	UNITED KINGDOM	Ground	1	OXFORD
75545	E00075627	UNITED KINGDOM	Ground	1	OXFORD
75550	E00075632	UNITED KINGDOM	Ground	1	OXFORD
75552	E00075634	UNITED KINGDOM	Ground	1	OXFORD
75554	E00075636	UNITED KINGDOM	Ground	1	MANCHESTER
75556	E00075638	UNITED KINGDOM	Ground	2	MANCHESTER
75557	E00075639	UNITED KINGDOM	Ground	1	OXFORD
75572	E00075654	UNITED KINGDOM	Ground	1	OXFORD
75583	E00075665	UNITED KINGDOM	Ground	1	OXFORD
75606	E00075688	UNITED KINGDOM	Ground	2	MANCHESTER
75620	E00075702	UNITED KINGDOM	Ground	1	OXFORD
75678	E00075760	UNITED KINGDOM	Ground	1	MANCHESTER
79998	E00080081	UNITED KINGDOM	Ground	1	OXFORD

	product_category	on_sale	transit_days	datetime_ordered	\
25	TOPS	Y	4	2016-07-03 04:52:12	
34	JACKETS & VESTS	N	3	2016-07-03 05:42:00	
37	JACKETS & VESTS	Y	3	2016-07-03 06:01:54	
39	TOPS	N	3	2016-07-03 06:38:27	
40	TOPS	Y	3	2016-07-03 06:39:40	
41	TOPS	N	4	2016-07-03 07:16:03	
42	JACKETS & VESTS	Y	3	2016-07-03 07:33:24	
43	JACKETS & VESTS	Y	3	2016-07-03 07:49:19	
44	JACKETS & VESTS	N	4	2016-07-03 07:55:02	
45	TOPS	Y	4	2016-07-03 07:59:58	
47	ACCESSORIES	Y	2	2016-07-03 08:16:49	
48	JACKETS & VESTS	N	3	2016-07-03 08:22:15	
49	JACKETS & VESTS	N	3	2016-07-03 08:25:17	
52	JACKETS & VESTS	Y	4	2016-07-03 08:51:53	
53	JACKETS & VESTS	Y	3	2016-07-03 08:52:33	
54	TOPS	N	4	2016-07-03 08:55:24	
55	JACKETS & VESTS	Y	3	2016-07-03 08:56:31	
58	TOPS	Y	3	2016-07-03 09:41:02	
60	TOPS	Y	3	2016-07-03 09:42:17	
65	JACKETS & VESTS	Y	4	2016-07-03 10:18:41	
70	TOPS	Y	3	2016-07-03 10:36:41	
77	TOPS	Y	3	2016-07-03 11:27:04	
78	JACKETS & VESTS	Y	3	2016-07-03 11:29:29	
83	ACCESSORIES	Y	3	2016-07-03 11:36:10	
97	TOPS	Y	2	2016-07-03 12:28:30	
104	JACKETS & VESTS	N	5	2016-07-03 13:03:22	
111	ACCESSORIES	N	2	2016-07-03 15:19:55	
115	JACKETS & VESTS	Y	3	2016-07-03 13:30:37	
116	TOPS	Y	3	2016-07-03 13:34:35	
118	JACKETS & VESTS	N	1	2016-07-03 16:38:00	
...	...	...	...	...	
75449	TOPS	Y	1	2016-10-14 17:59:01	
75464	TOPS	Y	1	2016-10-14 18:08:15	
75465	ACCESSORIES	N	3	2016-10-14 18:10:40	
75471	JACKETS & VESTS	Y	2	2016-10-14 18:18:18	
75476	JACKETS & VESTS	Y	1	2016-10-14 18:26:35	
75479	JACKETS & VESTS	N	2	2016-10-14 18:27:03	
75484	JACKETS & VESTS	Y	1	2016-10-14 18:37:41	
75488	TOPS	Y	2	2016-10-14 18:44:12	
75498	TOPS	N	2	2016-10-14 18:57:49	
75500	JACKETS & VESTS	N	2	2016-10-14 18:58:52	
75502	JACKETS & VESTS	Y	1	2016-10-14 19:02:06	
75503	TOPS	N	1	2016-10-14 19:03:39	
75507	JACKETS & VESTS	Y	1	2016-10-14 19:14:17	
75511	JACKETS & VESTS	Y	1	2016-10-14 19:17:36	
75513	TOPS	Y	3	2016-10-14 19:19:50	
75526	JACKETS & VESTS	Y	1	2016-10-14 19:34:48	

75529	JACKETS & VESTS	Y	3	2016-10-14	19:36:43
75544	JACKETS & VESTS	N	1	2016-10-14	19:50:58
75545	JACKETS & VESTS	Y	1	2016-10-14	19:53:14
75550	JACKETS & VESTS	Y	1	2016-10-14	20:00:25
75552	JACKETS & VESTS	Y	2	2016-10-14	20:01:00
75554	TOPS	Y	1	2016-10-14	20:02:16
75556	ACCESSORIES	N	1	2016-10-14	20:03:24
75557	JACKETS & VESTS	Y	2	2016-10-14	20:03:43
75572	JACKETS & VESTS	N	2	2016-10-14	20:24:07
75583	JACKETS & VESTS	Y	1	2016-10-14	20:40:16
75606	ACCESSORIES	N	1	2016-10-14	21:04:55
75620	JACKETS & VESTS	N	1	2016-10-14	21:28:55
75678	TOPS	N	2	2016-10-14	22:46:46
79998	JACKETS & VESTS	N	3	2016-10-21	12:15:57

	datetime_sourced	datetime_product_ready	datetime_planned	\
25	2016-07-03 07:17:09	2016-07-03 09:17:13	2016-07-06	
34	2016-07-03 07:16:36	2016-07-03 10:16:49	2016-07-05	
37	2016-07-03 08:16:37	2016-07-03 10:17:00	2016-07-05	
39	2016-07-03 08:17:11	2016-07-04 08:17:50	2016-07-05	
40	2016-07-03 08:16:57	2016-07-04 03:16:16	2016-07-06	
41	2016-07-03 09:16:34	2016-07-04 04:16:23	2016-07-06	
42	2016-07-03 09:17:23	2016-07-04 04:16:47	2016-07-05	
43	2016-07-03 09:17:20	2016-07-04 07:18:13	2016-07-06	
44	2016-07-03 10:16:45	2016-07-04 05:17:01	2016-07-06	
45	2016-07-03 10:16:52	2016-07-04 07:18:00	2016-07-07	
47	2016-07-03 10:17:01	2016-07-04 04:18:33	2016-07-06	
48	2016-07-03 10:16:57	2016-07-04 04:17:25	2016-07-06	
49	2016-07-03 10:16:57	2016-07-04 07:19:29	2016-07-06	
52	2016-07-03 10:16:24	2016-07-04 04:17:49	2016-07-07	
53	2016-07-03 11:16:29	2016-07-04 07:17:06	2016-07-06	
54	2016-07-03 11:16:22	2016-07-04 07:16:16	2016-07-08	
55	2016-07-03 11:16:37	2016-07-04 04:17:52	2016-07-06	
58	2016-07-04 03:16:31	2016-07-04 07:19:21	2016-07-06	
60	2016-07-03 11:16:26	2016-07-04 03:17:04	2016-07-06	
65	2016-07-03 12:16:29	2016-07-04 06:16:18	2016-07-09	
70	2016-07-03 13:16:30	2016-07-04 06:16:27	2016-07-06	
77	2016-07-03 13:16:44	2016-07-04 04:17:35	2016-07-07	
78	2016-07-03 13:16:38	2016-07-04 04:17:25	2016-07-07	
83	2016-07-03 13:16:50	2016-07-04 04:16:16	2016-07-07	
97	2016-07-03 14:16:38	2016-07-04 08:18:23	2016-07-06	
104	2016-07-03 15:16:40	2016-07-04 05:16:35	2016-07-13	
111	2016-07-03 15:56:23	2016-07-06 11:58:22	2016-07-07	
115	2016-07-03 15:16:17	2016-07-04 04:18:16	2016-07-07	
116	2016-07-03 16:16:21	2016-07-04 04:17:31	2016-07-07	
118	2016-07-03 17:41:49	2016-07-05 06:19:13	2016-07-06	
...	...	...	...	
75449	2016-10-14 19:09:18	2016-10-18 19:09:56	2016-10-19	



75464	2016-10-14	19:09:22	2016-10-18	19:29:07	2016-10-19
75465	2016-10-14	19:29:24	2016-10-18	15:49:13	2016-10-21
75471	2016-10-14	19:28:20	2016-10-18	19:18:48	2016-10-20
75476	2016-10-14	19:28:36	2016-10-18	10:49:21	2016-10-19
75479	2016-10-14	19:28:32	2016-10-17	10:05:15	2016-10-20
75484	2016-10-14	19:28:38	2016-10-17	10:20:22	2016-10-19
75488	2016-10-14	19:50:33	2016-10-18	19:23:35	2016-10-20
75498	2016-10-14	20:03:00	2016-10-18	19:27:22	2016-10-20
75500	2016-10-14	19:51:31	2016-10-18	14:12:39	2016-10-20
75502	2016-10-14	19:55:39	2016-10-18	14:12:43	2016-10-19
75503	2016-10-14	20:03:22	2016-10-18	19:23:18	2016-10-19
75507	2016-10-14	20:31:47	2016-10-17	10:22:37	2016-10-19
75511	2016-10-14	20:33:45	2016-10-18	18:25:25	2016-10-19
75513	2016-10-14	20:32:29	2016-10-18	19:31:01	2016-10-21
75526	2016-10-14	20:33:56	2016-10-19	18:05:18	2016-10-19
75529	2016-10-14	20:21:51	2016-10-18	14:10:35	2016-10-21
75544	2016-10-14	21:10:36	2016-10-17	10:21:10	2016-10-21
75545	2016-10-14	21:10:32	2016-10-18	10:52:38	2016-10-19
75550	2016-10-14	21:10:31	2016-10-18	19:28:21	2016-10-19
75552	2016-10-14	20:56:40	2016-10-18	14:30:53	2016-10-20
75554	2016-10-14	21:10:21	2016-10-18	19:20:27	2016-10-19
75556	2016-10-14	20:59:27	2016-10-18	19:20:58	2016-10-19
75557	2016-10-14	21:10:37	2016-10-18	18:25:24	2016-10-20
75572	2016-10-14	21:35:47	2016-10-17	10:22:46	2016-10-20
75583	2016-10-14	21:57:33	2016-10-18	14:36:15	2016-10-19
75606	2016-10-14	21:59:30	2016-10-18	19:20:21	2016-10-19
75620	2016-10-14	22:21:58	2016-10-18	14:57:24	2016-10-19
75678	2016-10-15	00:04:43	2016-10-17	16:08:02	2016-10-19
79998	2016-10-21	13:19:49	2016-10-24	17:20:54	2016-10-31

	deadline_source	deadline_make	deadline_deliver	delivered_to_plan	\
25	NaN	3	4	PASS	
34	NaN	3	3	PASS	
37	NaN	2	3	PASS	
39	NaN	2	3	PASS	
40	NaN	2	3	PASS	
41	NaN	25	4	PASS	
42	NaN	25	3	FAIL	
43	NaN	25	3	PASS	
44	NaN	24	4	FAIL	
45	NaN	24	4	PASS	
47	NaN	24	2	PASS	
48	NaN	24	3	PASS	
49	NaN	24	3	PASS	
52	NaN	24	4	PASS	
53	NaN	23	3	PASS	
54	NaN	23	4	PASS	
55	NaN	23	3	PASS	

58	NaN	7	3	PASS
60	NaN	23	3	PASS
65	NaN	22	4	PASS
70	NaN	21	3	PASS
77	NaN	21	3	PASS
78	NaN	21	3	PASS
83	NaN	21	3	PASS
97	NaN	20	2	PASS
104	NaN	19	5	PASS
111	NaN	83	2	FAIL
115	NaN	19	3	PASS
116	NaN	18	3	PASS
118	NaN	20	1	PASS
...	...	...	...	...
75449	NaN	143	2	FAIL
75464	NaN	143	2	PASS
75465	NaN	143	3	PASS
75471	NaN	143	3	PASS
75476	NaN	143	1	PASS
75479	NaN	143	2	PASS
75484	NaN	143	1	PASS
75488	NaN	142	3	PASS
75498	NaN	142	3	FAIL
75500	NaN	142	2	FAIL
75502	NaN	142	1	PASS
75503	NaN	142	2	PASS
75507	NaN	141	1	PASS
75511	NaN	141	1	PASS
75513	NaN	141	6	FAIL
75526	NaN	141	1	FAIL
75529	NaN	142	3	PASS
75544	NaN	141	1	FAIL
75545	NaN	141	1	PASS
75550	NaN	141	2	PASS
75552	NaN	141	2	PASS
75554	NaN	141	2	PASS
75556	NaN	141	2	PASS
75557	NaN	141	2	PASS
75572	NaN	140	2	PASS
75583	NaN	140	1	PASS
75606	NaN	140	2	FAIL
75620	NaN	140	1	PASS
75678	NaN	134	3	PASS
79998	NaN	77	3	PASS

	time_sourced	time_make	time_ready
25	144	4	0
34	94	4	0

37	134	4	0
39	98	1	1
40	97	20	0
41	120	21	0
42	103	20	0
43	88	23	0
44	141	21	0
45	136	23	0
47	120	20	0
48	114	19	0
49	111	22	0
52	84	19	0
53	143	22	0
54	140	22	0
55	140	19	0
58	1055	21	0
60	94	17	0
65	117	19	0
70	159	19	0
77	109	16	0
78	107	16	0
83	100	16	0
97	108	19	0
104	133	16	0
111	36	20	2
115	105	14	0
116	161	14	0
118	63	13	1
...	...	...	...
75449	70	1	4
75464	61	1	4
75465	78	21	3
75471	70	1	4
75476	62	16	3
75479	61	15	2
75484	50	15	2
75488	66	0	4
75498	65	0	4
75500	52	19	3
75502	53	19	3
75503	59	0	4
75507	77	15	2
75511	76	23	3
75513	72	0	4
75526	59	22	4
75529	45	18	3
75544	79	14	2
75545	77	14	3

75550	70	23	3
75552	55	18	3
75554	68	23	3
75556	56	23	3
75557	66	22	3
75572	71	13	2
75583	77	17	3
75606	54	22	3
75620	53	17	3
75678	77	17	2
79998	63	5	3

[535 rows x 19 columns]

### 535 rows have null columns in the data matrix

Here we create a new data matrix by the name of **data\_corr**. We drop the old variables and introduce the old variables.

```
In [439]: data_corr=data;
data_corr=data_corr.drop(['order_id','delivered_to_plan','datetime_ordered','datetime_received'])
data_corr.head()
```

```
Out[439]:
```

	country	shipping_method	units_per_order	facility	product_category	\
0	UNITED KINGDOM	Ground	1	OXFORD	ACCESSORIES	
1	FRANCE	Ground	1	ANTWERP	JACKETS & VESTS	
2	FRANCE	Ground	1	ANTWERP	TOPS	
3	FRANCE	Ground	1	ANTWERP	JACKETS & VESTS	
4	UNITED KINGDOM	Next Day	1	OXFORD	JACKETS & VESTS	

	on_sale	transit_days	deadline_source	deadline_make	deadline_deliver	\
0	Y	2	1612.0	38	3	
1	N	3	531.0	8	3	
2	Y	3	504.0	8	3	
3	Y	5	492.0	8	5	
4	Y	1	1567.0	6	1	

	time_sourced	time_make	time_ready
0	62	21	2
1	127	7	0
2	100	5	0
3	88	5	0
4	244	5	2

```
In [440]: print(data_corr.shape)
```

```
(79999, 13)
```

The new matrix: **data\_corr** has 13 columns now.

1.4.1 Creating a dictionary so that we can map the categorical values. Categorical variables are:

- **Country:** UNITED KINGDOM : 0, FRANCE : 1, GERMANY : 2, SWEDEN: 3, BELGIUM : 4
- **Shipping Method :** Ground : 0, : Next Day : 1, 3-Day: 2, 2-Day\*: 3
- **Facility:** OXFORD: 0, ANTWERP: 1, MANCHESTER: 2, AUGSBURG: 3, HANOVER: 4, EINDHOVEN: 5
- **Product Category:** ACCESSORIES: 0, JACKETS & VESTS: 1, TOPS: 2
- **On Sale:** Y: 0, N: 1

```
In [441]: col=['country','shipping_method','facility','product_category','on_sale']
          tmp=[];
          for i,j in enumerate(col):
              a=data_corr[j].unique()
              tmp.append(dict(zip(a,list(range(0,len(a))))))
          print(tmp[i])
          data_corr[j]=data_corr[j].map(tmp[i])
```

```
{'UNITED KINGDOM': 0, 'FRANCE': 1, 'GERMANY': 2, 'SWEDEN': 3, 'BELGIUM': 4}
{'Ground': 0, 'Next Day': 1, '3-Day': 2, '2-Day': 3}
{'OXFORD': 0, 'ANTWERP': 1, 'MANCHESTER': 2, 'AUGSBURG': 3, 'HANOVER': 4, 'EINDHOVEN': 5}
{'ACCESSORIES': 0, 'JACKETS & VESTS': 1, 'TOPS': 2}
{'Y': 0, 'N': 1}
```

```
In [442]: data_corr.head()
```

```
Out[442]:
```

	country	shipping_method	units_per_order	facility	product_category	\
0	0	0	1	0	0	
1	1	0	1	1	1	
2	1	0	1	1	2	
3	1	0	1	1	1	
4	0	1	1	0	1	

  

	on_sale	transit_days	deadline_source	deadline_make	deadline_deliver	\
0	0	2	1612.0	38	3	
1	1	3	531.0	8	3	
2	0	3	504.0	8	3	
3	0	5	492.0	8	5	
4	0	1	1567.0	6	1	

  

	time_sourced	time_make	time_ready
0	62	21	2
1	127	7	0
2	100	5	0
3	88	5	0
4	244	5	2

Pushing the column with NaN values:deadline\_source to the front for easy matrix manipulation which will be required later