In [47]:	import os import pandas a import numpy as import seaborn	s np									
In [48]:	<pre>import seaborn as sns import matplotlib.pyplot as plt  Reading the dataset  train = pd.read_csv/da("Documents/dataset ibm/train.csv") test = pd.read_csv("Documents/dataset ibm/test.csv")</pre>										
In [49]:	Exploratory data ana train.head()	alysis									
Out[49]:	id       week         0       1379560       1         1       1466964       1         2       1346989       1         3       1338232       1         4       1448490       1	55 55 55 55 55	meal_id chec 1885 1993 2539 2139 2631	136.83 136.83 134.86 339.50 243.50	152.29 135.83 135.86 437.53 242.50	for_promotion homep  0  0  0  0  0  0	age_featured num_  0  0  0  0  0  0	177 270 189 54 40			
In [50]: Out[50]:	id week  0 1028232 146	center_id	meal_id chec	ckout_price ba	ase_price emailer_ 159.11	for_promotion homep	age_featured 0				
	1       1127204       146         2       1212707       146         3       1082698       146	55 55 55	1993 2539 2631	160.11 157.14 162.02	159.11 159.14 162.02	0 0 0	0 0 0				
In [51]:	train.info() <class #="" 'pandas.c="" (to="" 0="" 1="" 2="" 3="" 32<="" 4="" 4568="" 5="" 6="" 7="" 8="" base_price="" center_id="" checkout_pr="" column="" columns="" data="" dtypes:="" emailer_for="" float640="" homepage_for="" id="" meal_id="" memory="" num_orders="" rangeindex:="" th="" usage:="" week=""><th>otal 9 co rice r_promoti eatured</th><th>es, 0 to 45 Non-Null 456548 456548 456548 456548 456548 456548 456548 456548 456548 456548</th><th>56547  Ll Count [ non-null i non-null i</th><th>float64 float64 int64 int64</th><th>0</th><th>0</th><th></th><th></th><th></th><th></th></class>	otal 9 co rice r_promoti eatured	es, 0 to 45 Non-Null 456548 456548 456548 456548 456548 456548 456548 456548 456548 456548	56547  Ll Count [ non-null i non-null i	float64 float64 int64 int64	0	0				
In [52]: Out[52]:	train['num_orders'].describe()  count										
In [53]: Out[53]:	id week center_id meal_id checkout_price base_price emailer_for_pron homepage_feature	notion	0 0 0 0 0 0								
In [63]:	num_orders dtype: int64 Reading and mergin  meal_info = pd. center_info = p  trainfinal = po trainfinal = po trainfinal.head	read_csv od.read_cs d.merge(t	("Documents sv("Documer rain, meal_	nts/dataset _info, on="m	<pre>ibm/fulfilment neal_id", how="</pre>	_center_info.csv" outer")					
Out[63]: In [64]:	0       1379560       1         1       1018704       2         2       1196273       3         3       1116527       4         4       1343872       5	55 55 55 55 55	1885 1885 1885 1885 1885	136.83 135.83 132.92 135.86 146.50	152.29 152.29 133.92 134.86 147.50	for_promotion homep 0 0 0 0 0	age_featured num_  0  0  0  0  0  0	orders category of 177 Beverages 323 Beverages 96 Beverages 163 Beverages 215 Beverages	Thai 647	n_code center_type  56    TYPE_C  56    TYPE_C  56    TYPE_C  56    TYPE_C  56    TYPE_C  57    TYPE_C  57    TYPE_C	2.0 2.0 2.0 2.0 2.0 2.0
Out[64]:	0       1379560       1         1       1018704       2         2       1196273       3         3       1116527       4	checkout_p  13  13  13		ice emailer_fo .29 .29 .92			<ul><li>Beverages</li><li>Beverages</li><li>Beverages</li><li>Beverages</li><li>Beverages</li></ul>	cuisine city_code re Thai 647 Thai 647 Thai 647 Thai 647 Thai 647 Thai 647	egion_code center_type  56    TYPE_C  56    TYPE_C  56    TYPE_C  56    TYPE_C	2.0 2.0 2.0 2.0	
In [65]: In [66]:	<pre>Dropping columns  cols = trainfir print(cols)</pre>	nal.colum 'checkout 'op_area'	ns.tolist() _price', 'b	oase_price',	. 'emailer_for_		215 Beverages  Dage_featured',		56 TYPE_C		egion_code',
In [67]:	<pre>print(cols)</pre>	'city_cod m_orders' rainfinal	e', 'region ]			p_area', 'categor	y', 'cuisine', '	'checkout_price',	'base_price', 'em	ailer_for_promotio	n', 'homepage
Out[67]:	id week city_code region_code center_type op_area category cuisine checkout_price base_price emailer_for_prom homepage_feature num_orders dtype: object	notion	int64 int64 int64 object float64 object object float64 float64 int64 int64								
In [68]:	from sklearn.pr Label encoding trainfinal.head		ing <b>import</b>	LabelEncode	er						
Out[69]:			region_code  56  56  56  56  56	Center_type  TYPE_C  TYPE_C  TYPE_C  TYPE_C  TYPE_C	op_area category  2.0 Beverages  2.0 Beverages  2.0 Beverages  2.0 Beverages  2.0 Beverages  3.0 Beverages  4.0 Beverages  4.0 Beverages	5 Thai 13 5 Thai 13 6 Thai 13 6 Thai 13	rice base_price en 6.83	mailer_for_promotion  0  0  0  0  0	homepage_featured 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	um_orders  177  323  96  163  215	
In [70]: Out[70]:	trainfinal.shap (456548, 13)	oe									
In [71]:	plt.figure(figs sns.distplot(tr plt.xlabel("num plt.ylabel("Num plt.title("num_	size=(12, rainfinal n_orders" nber of Bu orders Di aconda3\l apt your (msg, Fut	7)) .num_orders ) uyers") istribution ib\site-pac code to use ureWarning	ckages\seabo e either `d: ) ution')	orn\distributio isplot` (a figu	re-level function	eWarning: `distp with similar f]	olot` is a depred lexibility) or `h	cated function and nistplot` (an axes-	will be removed in level function for	a future ver histograms).
	0.0040			nu	m_orders D	DISCRIBUCION					
	0.0000 O		5000		10000	15000	20000	25000			
In [72]:	<pre>trainfinal2 = t correlation = t columns = corre columns</pre>	rainfina elation.n	l2.corr(met largest(8,	hod='pearso 'num_orders	on') s').index	ders					
Out[72]: In [73]:	correlation_map	de', 'reg ject') o = np.co cale=1.0)	ion_code',	'week', 'ba	ase_price'], .umns].values.T .rue, annot=Tru	)	nt='.2f', ytickl	labels=columns.va	lues, xticklabels=0	columns.values)	
	homepage_featured emailer_for_promotion	0.29 1.00 0 0.28 0.39 1 0.18 0.04 4 0.04 0.01 4 0.03 0.00 4 -0.02 -0.01 4 -0.22 0.06 0	0.39 0.04 0.01 1.00 -0.02 -0.01 0.02 1.00 0.13 0.01 0.02 0.04 0.00 0.00 0.00 0.17 0.02 -0.00 unified do	0.03 -0.02 -0.22 0.00 -0.01 0.06 -0.01 -0.00 0.17 0.02 0.00 0.02 0.04 0.00 -0.00 1.00 0.00 -0.00 0.00 1.00 0.03 -0.00 0.03 1.00  poolujba	- 0.8 - 0.6 - 0.4 - 0.2 - 0.0 0.2						
<pre>In [74]: Out[74]:</pre>		umns.drop crainfina 3.values ['num_ordo ad()	(['num_orde 1[features] ers'].value r_for_promotio	ers']) es on op_area o	city_code region_c	ode week base_price	_				
	0 1 2 3 4 Split the dataset into	0 0 0 0 0 train set a		0 2.0 0 2.0 0 2.0 0 2.0 0 2.0	647 647 647 647	56     1     152.29       56     2     152.29       56     3     133.92       56     4     134.86       56     5     147.50	2				
In [75]:	<pre>from sklearn.mc X_train, X_val,</pre>					t_size=0.25)					