Ce	train_df = pd.read_csv('Credit card/fraudTrain.csv.zip') test_df = pd.read_csv('Credit card/fraudTest.csv.zip') FileNotFoundError											
Fil en rma												
	<pre>nt, encoding, encoding_errors, dialect, on_bad_lines, delim_whitespace, low_memory, memory_map, float_precision, storage_options, dtype_backend) 1013 kwds_defaults = _refine_defaults_read(1014</pre>											
->	1023 1024 kwds.update(kwds_defaults) -> 1026 return _read(filepath_or_buffer, kwds) File ~\AppData\Local\Programs\Python\Python312\Lib\site-packages\pandas\io\parsers\readers.py:620, in _read(filepath_or_buffer, kwds) 617 _validate_names(kwds.get("names", None)) 619 # Create the parser> 620 parser = _TextFileReader(filepath_or_buffer, **kwds) 622 if chunksize or iterator: 623											
-> Fil	1617 se 1619 self.h 1620 self le ~\AppData	lf.options[" andles: IOHa engine = <mark>sel</mark> \ Local\Progr	has_index_na ndles None fmake_engi ams\Python\F	nmes"] = kwds["has_ e = None ne(f, self.engine)	_index_names"]				_(self, f, engine, engine(self, f, engi			
->	<pre>if "b" not in mode: mode += "b" > 1880 self.handles = get_handle(</pre>											
	1884 CC 1885 me 1886 is 1887 er 1888 st	mpression=se mory_map=sel _text=is_tex rors=self.op	<pre>lf.options.ge f.options.ge t, tions.get("e</pre>		None), alse), 'strict"),							
	1891 f = se		andle ams\Python\F		-packages\pandas	s\io\common.py:873	. in get_handle (pa	th_or_buf, mode, e	ncoding, compression	ı, memory_map, is_t	text, errors, stora	nge_optio
:	869 # 870 #	Binary mode	r the filena does not sup ding and "b'	ume is to be opened oport 'encoding' ar ' not in ioargs.mod	nd 'newline'.	e.						
	874 875 876 877 878 879	encodi	.mode, ng=ioargs.er =errors,	ncoding,								
Fi:	881 882		pen(handle,	ioargs.mode)	'fraudTrain.csv	, '						
t		read_csv('Cr		fraudTrain.csv.zip raudTest.csv.zip')	')							
[4]: 	0	rans_date_trans 2019-01-01 00:		cc_num merc fraud_Ri 6189652095 Kub				treet lat 561 Perry 36.0788 -81	1781 3/195	job dob hologist, 1988- 0b24	tr -2abb623afc578575680d	rans_num
1		2019-01-01 00:			dann eller,		43 F	Cove 8039 Riley eens 48.8878 -118	CO	Special 1978-	529f8574734946361c46 ²	
1	I	2019-01-01 00.	00.44 630		ieme	s 107.23 Stephanie	S	Suite 393 594		06-21 teacher	52916574734946361046	10024099
2	2 2	2019-01-01 00:	00:51 38859	fraud_l 9492057661 Buckı	entertainmen	t 220.11 Edward	Sanchez M [S	Vhite Dale 42.1808 -112 Suite 530	2620 4154 cons	Nature servation officer 1962- a1a2: 01-19	2d70485983eac12b5b88	dad1cf95
3	3	2019-01-01 00:	01:16 3534093	fraud_Ki 3764340240 Herm and Fa	iston gas_transpor	t 45.00 Jeremy	Cyr White M C	9443 nthia Court 46.2306 -112 Apt. 038	2.1138 1939 Patent	1967- attorney 01-12 6b84	9c168bdad6f867558c379	93159a81
4		2019-01-01 00:	03:06 375534	fraud_Kee 1208663984	eling- misc_pos Crist	s 41.96 Tyler	Garcia M Bra	408 Idley 38.4207 -79 Rest		Dance 1986- ovement 03-28 therapist	7549acf90789359a9aa53	346dcb46
	rows × 23 colurain_df.info											
<c.< th=""><th>lass 'pandas ngeIndex: 12 ta columns (</th><th>.core.frame. 96675 entrie total 23 col</th><th>s, 0 to 1296</th><th></th><th></th><th></th><th></th><th></th><th></th><th></th><th></th><th></th></c.<>	lass 'pandas ngeIndex: 12 ta columns (.core.frame. 96675 entrie total 23 col	s, 0 to 1296									
0 1 2 3	Unnamed: trans_dat cc_num		1296675 no 1296675 no 1296675 no									
4 5 6 7 8	category amt first last gender		1296675 no 1296675 no 1296675 no 1296675 no	on-null object on-null float64 on-null object on-null object on-null object								
1: 1:	street 0 city 1 state 2 zip		1296675 no 1296675 no 1296675 no 1296675 no	on-null object on-null object on-null object on-null int64								
1; 1; 1;	3 lat 4 long 5 city_pop		1296675 no 1296675 no 1296675 no 1296675 no	on-null float64 on-null float64 on-null int64 on-null object on-null object								
18 19 20 21	8 trans_num 9 unix_time 0 merch_lat 1 merch_lor		1296675 no 1296675 no 1296675 no 1296675 no	on-null object on-null int64 on-null float64 on-null float64								
dty mer 6]: t	mory usage: est_df.info)	6), object(1	on-null int64 .2)								
<c.< td=""><td>lass 'pandas ngeIndex: 55 ta columns (Column</td><td>.core.frame. 5719 entries total 23 col</td><td>, 0 to 55571</td><td>Count Dtype</td><td></td><td></td><td></td><td></td><td></td><td></td><td></td><td></td></c.<>	lass 'pandas ngeIndex: 55 ta columns (Column	.core.frame. 5719 entries total 23 col	, 0 to 55571	Count Dtype								
0 1 2 3	Unnamed: trans_dat cc_num merchant		555719 nor 555719 nor 555719 nor 555719 nor	n-null int64 n-null object n-null int64 n-null object								
4 5 6 7 8	first last gender		555719 nor 555719 nor 555719 nor 555719 nor	n-null object n-null float64 n-null object n-null object n-null object								
9 10 11 12 13	street 0 city 1 state 2 zip 3 lat		555719 nor 555719 nor 555719 nor 555719 nor	n-null object n-null object n-null object n-null int64 n-null float64								
14 15 10 1	4 long 5 city_pop 6 job 7 dob		555719 nor 555719 nor 555719 nor 555719 nor	n-null float64 n-null int64 n-null object n-null object								
19 20 21 22	8 trans_num 9 unix_time 0 merch_lat 1 merch_lor 2 is_fraud	g	555719 nor 555719 nor 555719 nor 555719 nor	n-null object n-null int64 n-null float64 n-null float64 n-null int64								
dty mer 7]: t	ypes: float@mory usage:		6), object(1									
t C m	Jnnamed: 0 trans_date_t cc_num merchant category	rans_time	0 0 0 0									
a f 1 9 s	amt first Last gender street		0 0 0 0									
c s z 1	street city state zip lat Long		0 0 0 0									
c j d t	city_pop job dob trans_num		0 0 0 0 0									
m m i	unix_time nerch_lat nerch_long is_fraud dtype: int64		0 0 0 0									
8]: t	est_df.isnu Jnnamed: 0 trans_date_t		0 0									
c m c a f	cc_num merchant category amt first		9 0 0 0									
1 g s c s	last gender street city state		0 0 0 0									
z 1 1 0	zip Lat Long city_pop		0 0 0 0									
d t u m	job dob trans_num unix_time merch_lat		0 0 0									
m i d	nerch_long is_fraud dtype: int64 drop_columns			','merchant','tran	s_num','unix ti	me','first','last'	,'street','zip'l					
t t:	rain_df.dropeest_df.dropeerint(train_c	(columns=drop columns=drop df.shape)	p_columns,ir	nplace <mark>=True</mark>)	-n_til	, 2000	,					
(12 (5!		itudinal_dis		os(round(train_df[
t	rain_df['lonest_df['lat:	gitudinal_di .tudinal_dist	stance'] = a	abs(round(train_df abs(round(train_df s(round(test_df['mos(round(test_df['i	['merch_long']- erch_lat']-test	train_df['long'],3 _df['lat'],3))))					
t	rain_df.drop est_df.drop	(columns=dro columns=drop	op_columns,in o_columns,inp	olace= True)	long','job','do	b','merch_lat','me	rch_long','state']					
t	rain_df.gendeest_df.gende	ler=train_df.	gender.apply	into numerical /(lambda x: 1 if x: lambda x: 1 if x==								
t t	rain_df = po est_df = pd est_df = tes	l.get_dummies get_dummies(t_df.reindex	(train_df, otest_df, co.	columns=['category lumns=['category'] ain_df.columns, fi	, prefix='catego							
.5]: t .5]:			s_fraud latitud 0	inal_distance longitudir 0.068	nal_distance catego	ory_entertainment cate	gory_food_dining cate False	egory_gas_transport cat False	egory_grocery_net categ False	ory_grocery_pos cateo	gory_health_fitness cate	egory_homo
1	107.23	0 149 1 4154	0	0.271 0.970	0.024 0.108	False True	False False	False False	False False	True False	False False	False False
3	3 45.00 4 41.96	 1 1939 1 99 	0	0.804 0.254	0.447 0.830	False False	False False	True False	False False	False False	False False	False
6]:		er city_pop is							gory_grocery_net catego			
1		1 333497 0 302 0 34496	0 0 0	0.020 0.870 0.177	0.265 0.476 0.660	False False False	False False False	False False False	False False	False False	False False True	False False
3	3 60.05	1 54767 1 1126	0	0.243	0.064	False	False False	False False	False False	False False	False	False False
у. Х.	_train = tra _test = test	in_df.drop(' in_df['is_fr :_df.drop('is :_df['is_frau	aud'] _fraud', ax:	•								
8]: f s X	r om sklearn caler = Star _train = sca	preprocessir dardScaler() ler.fit_trar	ng import Sta nsform(X_tra	andardScaler in)								
9]: f f c .	rom sklearn from sklearn lassifier= [er.transform tree import metrics import decisionTree(DecisionTree ort confusion Classifier(re	,	_score							
c. y. cı p	classifier.fi _pred1= clas cm = confusio crint(cm)	t(X_train, y sifier.predi n_matrix(y_t	v_train) .ct(X_test) .est, y_pred:									
[[! acc	552309 126 844 136 curacy:0.996	5] 1]] 204916513561	7									
f c. c. y.	rom sklearn lassifier = lassifier.fi _pred2=class	linear_model LogisticRegr t(X_train, y ifier.predic	import Logaression(rando v_train) ct(X_test)	n_matrix, accuracy isticRegression om_state = 0)	_score							
p p	em = confusion erint(cm) erint("accura 553251 32	on_matrix(y_t .cy:"+str(acc 3]	est, y_pred2	2) (y_test,y_pred2)))								
[aco 1]: f	2145 curacy:0.995 rom sklearn rom sklearn	0]] 558906569687 ensemble imp metrics impo	oort RandomFo ort confusion	prestClassifier n_matrix, accuracy								
	f=RandomFore	stClassifier .n,y_train)	(n_estimato	rs=10,criterion='e		state=0)						
r r y. cı	_pred3 = rf	on_matrix(y_t	.est, y_preus									