```
import numpy as np
import pandas as pd
import matplotlib as mpl
```

```
df
```

	Unnamed: 0	trans_date_trans_time	cc_num	merchant	category	amt	first	last	gender	s
0	0	2020-06-21 12:14:25	2291163933867244	fraud_Kilback LLC	personal_care	2.86	Jeff	Elliott	M	351
1	1	2020-06-21 12:14:33	3573030041201292	fraud_Cormier LLC	personal_care	29.84	Joseph	Williams	M	3
2	2	2020-06-21 12:15:50	3592034782432626	fraud_Stokes-Hauck	gas_transport	41.28	Ashley	Lopez	F	95
3	3	2020-06-21 12:15:51	3591919803438423	fraud_Herman, Weber and Hane	misc_pos	60.05	Brian	Williams	M	6
4	4	2020-06-21 12:17:32	5560532966068667	fraud_Schumm PLC	travel	3.19	Nathan	Massey	M	3
555718	555718	2020-12-31 23:58:47	3560348462420883	fraud_Rolfson-Donnelly	misc_pos	99.06	Kim	Lawson	F	7
555719	555719	2020-12-31 23:59:07	6011737457094949	fraud_Rau-Robel	misc_pos	28.06	Kim	Lawson	F	7
555720	555720	2020-12-31 23:59:42	3587767990292039	fraud_Breitenberg LLC	gas_transport	19.65	Donna	Garcia	F	8
555721	555721	2020-12-31 23:59:44	4728822152275629	fraud_Langosh, Wunsch and Morissette	entertainment	30.74	Samuel	Frey	M	7

```
df.shape
```

```
(555719, 23)
```

```
df.columns
```

```
Index(['Unnamed: 0', 'trans_date_trans_time', 'cc_num', 'merchant',
       'category', 'amt', 'first', 'last', 'gender', 'street',
       'city', 'state', 'zip', 'lat', 'long', 'city_pop', 'job',
       'dob', 'trans_num', 'unix_time', 'merch_lat', 'merch_long',
       'is_fraud'],
      dtype=object)
```

lat_long_merch_lat,merch_long

convert lat_long, merch_lat, merch_long to distance

new column: distance

drop column: lat_long, merch_lat, merch_long

drop: Unnamed: 0

```
ハバサイン公式 : 距離の計算
df['distance'] = 6371 * 2 * np.arcsin(np.sqrt(np.sin((df['merch_lat'] - df['lat'])/2)**2 + np.cos(df['lat']) * np.cos(df['merch_lat']) * np.sin((df['merch_long'] - df['long'])/2)**2))
```

```
df.drop(['Unnamed: 0', 'lat_long', 'merch_lat', 'merch_long'], axis=1, inplace=True)
```

```
df.head(5)
```

	trans_date_trans_time	cc_num	merchant	category	amt	first	last	gender	street	city	state	zip	lat	long	city_pop	job	dob	trans_num	unix_time	is_fraud	distance
0	2020-06-21 12:14:25	2291163933867244	fraud_Kilback LLC	personal_care	2.86	Jeff	Elliott	M	351 S Ridge Top Rd	Ashland	OR	97520	42.1808	-122.7042	19157	Chemical engineer	1977-03-23	0b242abb623afa9f1f131446a3775777	1371816865	0	78.773509
1	2020-06-21 12:14:33	3573030041201292	fraud_Cormier LLC	personal_care	29.84	Joseph	Williams	M	3655 Breanna Stravenue	Garden Grove	CA	92843	33.7757	-117.9372	165475	Geochemist, analytical	1988-07-30	1f9f643cf194ab8a901a4316ca0a6424	1371816873	0	30.063936
2	2020-06-21 12:15:50	3592034782432626	fraud_Stokes-Hauck	gas_transport	41.28	Ashley	Lopez	F	974 Sandy Mall	Monroe	WI	53566	42.6018	-90.3678	39690	Intelligence analyst	1962-01-19	414519b7e8b65116d514fd0f67bb67ec	1371816950	0	108.824047
3	2020-06-21 12:15:51	3591919803438423	fraud_Herman, Weber and Hane	misc_pos	60.05	Brian	Williams	M	6549 Cynthia Court	Madison	OH	43062	40.1475	-83.1772	71555	Dance movement psychotherapist	1955-07-28	2d91a119a1f3ba14d4a2e0f0cb0a04e3	1371816951	0	80.246502
4	2020-06-21 12:17:32	5560532966068667	fraud_Schumm PLC	travel	3.19	Nathan	Massey	M	3578 Walnut Avenue	Fairfax	VA	22030	38.8566	-77.3164	32303	Office: Marketing agency manager	1954-03-16	2d7579a7a422cb8d05ac5003d1d495cd	1371816952	0	26.186161

```
df.shape
```

```
(555719, 22)
```

trans_date_trans_time

convert trans_date_trans_time to datetime64[ns]

convert trans_date_trans_time to time of day: categorized from 00:00:00 to before_dawn, 4:00:00-7:59:59 to early_morning,8:00:00-11:59:59 to morning,12:00:00-16:59:59 to afternoon,17:00:00-19:59:59 to evening, 20:00:00-23:59:59 to night.

convert trans_date_trans_time to day of week: categorized from 0:Mon to before_dawn,1:Tues to before_dawn,2:Wed to before_dawn,3:Thu to before_dawn,4:Fri to before_dawn,5:Sat to before_dawn,6:Sun to before_dawn,7:Mon to early_morning,8:Tues to early_morning,9:Wed to early_morning,10:Thu to early_morning,11:Fri to early_morning,12:Sat to early_morning,13:Mon to morning,14:Tues to morning,15:Wed to morning,16:Thu to morning,17:Fri to morning,18:Sat to morning,19:Mon to afternoon,20:Tues to afternoon,21:Wed to afternoon,22:Thu to afternoon,23:Fri to afternoon,24:Sat to afternoon,25:Mon to evening,26:Tues to evening,27:Wed to evening,28:Thu to evening,29:Fri to evening,30:Sat to evening,31:Mon to night,32:Tues to night,33:Wed to night,34:Thu to night,35:Fri to night,36:Sat to night

new columns: month, day,time,weekend

drop columns: trans_date_trans_time

```
df['trans_date_trans_time'] = pd.to_datetime(df['trans_date_trans_time'])
```

```
df['month'] = df['trans_date_trans_time'].dt.month
df['day'] = df['trans_date_trans_time'].dt.weekday
df['time'] = df['trans_date_trans_time'].dt.time
df['time'] = df['time'].astype('str')
df['time'] = df['time'].str.replace(':','.')
df['time'] = df['time'].str.rsplit('.',1,expand=True)[0]
df['time'] = df['time'].str.replace('.','')
df['time'] = df['time'].astype('int')
```

```
def time(x):
    if 000000 <= x <= 040000:
        return 'before_dawn'
    elif 040000 <= x <= 075959:
        return 'early_morning'
    elif 080000 <= x <= 115959:
        return 'morning'
    elif 120000 <= x <= 165959:
        return 'afternoon'
    elif 170000 <= x <= 195959:
        return 'evening'
    elif 200000 <= x <= 235959:
        return 'night'
df['time'] = df['time'].apply(time)
```

```
df['weekend'] = df['day'].apply(weekend)
```

```
df.drop(['trans_date_trans_time'], axis=1, inplace=True)
```

```
df.head(5)
```

	cc_num	merchant	category	amt	first	last	gender	street	city	state	zip	lat	long	city_pop	job	dob	trans_num	unix_time	is_fraud	distance	month	day	time	weekend
0	2291163933867244	fraud_Kilback LLC	personal_care	2.86	Jeff	Elliott	M	351 S Ridge Top Rd	Ashland	OR	97520	42.1808	-122.7042	19157	Chemical engineer	1977-03-23	0b242abb623afa9f1f131446a3775777	1371816865	0	78.773509	6	0	before_dawn	0
1	3573030041201292	fraud_Cormier LLC	personal_care	29.84	Joseph	Williams	M	3655 Breanna Stravenue	Garden Grove	CA	92843	33.7757	-117.9372	165475	Geochemist, analytical	1988-07-30	1f9f643cf194ab8a901a4316ca0a6424	1371816873	0	30.063936	6	0	before_dawn	0
2	3592034782432626	fraud_Stokes-Hauck	gas_transport	41.28	Ashley	Lopez	F	974 Sandy Mall	Monroe	WI	53566	42.6018	-90.3678	39690	Intelligence analyst	1962-01-19	414519b7e8b65116d514fd0f67bb67ec	1371816950	0	108.824047	6	0	before_dawn	0
3	3591919803438423	fraud_Herman, Weber and Hane	misc_pos	60.05	Brian	Williams	M	6549 Cynthia Court	Madison	OH	43062	40.1475	-83.1772	71555	Dance movement psychotherapist	1955-07-28	2d91a119a1f3ba14d4a2e0f0cb0a04e3	1371816951	0	80.246502	6	0	before_dawn	0
4	5560532966068667	fraud_Schumm PLC	travel	3.19	Nathan	Massey	M	3578 Walnut Avenue	Fairfax	VA	22030	38.8566	-77.3164	32303	Office: Marketing agency manager	1954-03-16	2d7579a7a422cb8d05ac5003d1d495cd	1371816952	0	26.186161	6	0	before_dawn	0

```
df.shape
```

```
(555719, 23)
```

Gender

encode gender to dif values 0 = female and 1 = male

```
df.gender.unique()
```

```
array(['M', 'F'], dtype=object)
```

```
df['gender'] = df['gender'].apply(gender_fun)
```

```
df.gender.unique()
```

```
array([1, 0], dtype=int64)
```

City and City_pop

encode by city size based on the fqcy quantiles of the city population for each city

city size: small, medium, medium_large, large

new column: city_size

drop column: city_street

```
df.groupby('city').agg({'city_pop':['min','max']})
```

city	city_pop	
	min	max
Acme	1819	1819
Adams	4256	4256
Adamstown	5426	5426
Adelphia	1342	1342
Adena	1335	1335
Adjuntas	5616	5616
Agua Dulce	9519	9519
Airmont	32039	32039
Akron	217074	217074
Alabama	2154	2154
Alabama Hills	2000	2000
Alabaster	33357	33357
Alachua	9514	9514
Alamo	15629	15629
Alamogordo	35582	35582
Alapaha	1478	1478
Albany	97030	97030
Albion	5298	5298
Alcoa	8513	8513
Alcolu	1000	1000
Alder	1052	1052
Alderson	1451	1451
Alexander City	15008	15008
Alexandria	46274	46274
Alfred	1131	1131
Algonquin	30192	30192
Alice	19543	19543
Allen	1005	1005
Allendale	5372	5372
Allenhurst	2337	2337
Allenwood	1000	1000
Alma	9052	9052
Alma	1116	1116
Alma	3000	3000
Almena	1000	1000
Alpharetta	65677	65677
Alpine	1437	1437
Altadena	42854	42854
Altamont	1565	1565
Altamonte Springs	41557	41557
Altoona	46474	46474
Altus	19813	19813
Alva	5416	5416
Amarillo	196189	196189
Amherst	9042	9042
Amherst	2333	2333
Amherst	1000	1000
Amherst	1000	1000
Amherst	1000	1000