

PHASE 3:DOCUMENTATION

Credit Card Fraud Detection in loading and preprocessing the dataset

The IMDb dataset typically includes a variety of features that describe different aspects of a movie.

While the exact features can vary depending on the dataset and the source,

here are some common features you might find in an IMDb dataset:

The Credit card fraud detection dataset extracted from kaggle platform accurately consists of 31 columns with 2 lakhs+ entries.

COLUMN NAMES

>Time

>V1-V28

>Amount

>Class

Time	V1	V2	V3	V4	V5	V6	V7	V8	V9	V10	V11	V12	V13	V14	V15	V16	V17	V18	V19	V20	V21	V22	V23	V24	V25	V26	V27	V28	Amount
0	-1.35981	-0.07278	2.536347	1.378155	-0.33832	0.462388	0.239599	0.098998	0.363787	0.090794	-0.5516	-0.6178	-0.99139	-0.31117	1.468177	-0.4704	0.207971	0.025791	0.403993	0.251412	-0.01831	0.277838	-0.11047	0.066928	0.128539	-0.18911	0.133558	-0.02105	0.014724
0	1.31857	0.266151	0.16648	0.448154	0.060018	-0.08236	-0.0788	0.085102	-0.25543	-0.16697	1.612727	1.065235	0.480995	-0.14377	0.635558	0.463917	-0.1148	-0.18336	-0.14578	-0.06908	-0.22578	-0.63867	0.101288	-0.33985	0.16717	0.125895	-0.00898	0.014724	0.014724
1	-1.38815	-1.34016	1.773209	0.37978	-0.5032	1.800499	0.791461	0.247676	-1.51465	0.207643	0.624501	0.066084	0.717293	-0.16595	2.345865	-2.89008	1.109969	-0.12136	-1.26186	0.52498	0.247998	0.771679	0.909412	-0.68928	-0.32764	-0.1391	-0.05535	-0.05975	0.05975
1	-0.96627	-0.18523	1.792993	-0.88529	-0.01031	1.247203	0.237609	0.377436	-1.38702	-0.05495	-0.22649	0.178228	0.507757	-0.28792	-0.83142	-1.05965	-0.68429	1.965775	-1.23262	-0.20804	-0.1083	0.002374	-1.19302	-1.17758	0.062723	0.22193	0.062723	0.061458	0.061458
2	-1.15253	0.877737	1.548718	0.403004	-0.40719	0.095921	0.592941	-0.27053	0.817739	0.753074	-0.82384	0.338196	1.345852	-1.11967	1.75121	-0.45145	-0.23703	-0.03819	0.803487	0.405452	-0.00943	0.798278	-0.13746	0.141267	-0.20601	0.502292	0.212422	0.215153	0.215153
2	-0.42597	0.960523	1.341109	-0.16825	0.420987	-0.02973	0.476201	0.260314	-0.56867	-0.1741	1.341262	0.558994	-0.35809	-0.13713	0.517617	0.401726	-0.05813	0.086653	-0.03319	0.084968	-0.20825	-0.55982	-0.0264	-0.37143	-0.23279	0.105915	0.253844	0.08108	0.08108
4	1.229658	0.141004	0.045171	1.202613	0.191881	0.272708	-0.005316	0.081213	0.46496	-0.09925	-1.41691	-0.15383	-0.75106	0.167372	0.050144	-0.44359	0.002821	-0.61199	-0.04558	-0.21963	-0.16772	-0.27071	-0.1541	-0.78006	0.750137	-0.25724	0.034507	0.005168	0.005168
7	-0.64427	-1.417964	1.07438	-0.4922	0.948934	0.428118	1.120631	-1.30786	0.615375	1.249376	-0.61947	0.291474	1.757964	-1.32387	0.686133	-0.07813	-1.22213	-0.35822	0.324505	-0.15674	1.943465	-1.01545	0.057504	-0.64971	-0.41527	-0.05163	-1.20692	-1.08534	0.08534
7	-0.89429	0.286137	-0.11319	-0.27153	2.669599	1.721818	0.170145	0.851084	-0.39205	-0.41043	-0.70512	-0.11945	-0.28625	0.074355	-0.32878	-0.21008	-0.49977	0.118765	0.570328	0.052738	-0.07343	-0.26809	-0.20423	1.011592	0.373205	-0.08416	0.011747	0.142404	0.142404
9	-0.33825	1.119593	1.040467	-0.22215	0.499361	-0.24676	0.651583	0.095535	-0.73673	-0.36685	1.017614	0.83639	1.00644	-0.44352	0.150219	0.759453	-0.54008	0.476677	0.451773	0.203711	-0.24051	-0.03375	-0.121079	-0.38505	-0.06979	0.094159	0.246219	0.030376	0.030376
10	1.449044	-1.17834	0.91386	-1.37567	-1.97138	-0.29515	-1.42324	0.084856	-1.72041	1.626659	-1.39944	-0.67144	-0.51195	-0.09505	0.20393	0.031967	0.253415	0.854344	-0.22137	-0.38723	-0.0093	0.311894	0.02774	0.500512	0.251367	-0.12948	0.04263	0.016253	0.016253
10	0.384978	0.616109	-0.8743	-0.09402	2.924584	3.317027	0.470455	0.538247	-0.55889	0.309755	-0.25912	-0.32614	-0.09005	0.362832	0.528904	-0.12949	-0.80998	0.359985	0.707664	0.125992	0.049924	0.238422	0.00913	0.99671	-0.76731	-0.49221	0.042472	-0.05434	0.05434
10	1.249999	-1.22164	0.38393	-1.2349	-1.48542	-0.75323	-0.6894	-0.22749	-0.29401	1.323729	0.227866	-0.34268	1.205417	-0.31763	0.725675	-0.81361	0.877936	-0.84779	-0.68319	-0.10276	-0.23181	-0.48329	0.084668	0.392811	0.161135	-0.35499	0.026416	0.042422	0.042422
11	1.069374	0.287722	0.828813	2.71252	-0.1784	0.337544	-0.09672	0.115982	-0.22108	0.46023	-0.77366	0.323387	-0.26657	-0.17849	-0.65556	-0.19993	0.124005	-0.9805	-0.98292	-0.1532	-0.03688	0.074412	-0.07141	0.104744	0.548265	0.104094	0.021491	0.021293	0.021293
12	-2.79185	-0.32777	1.04175	1.76747	-0.13659	0.607596	-0.42291	-1.90711	0.757313	1.151087	0.844555	0.762944	0.379448	-0.73498	0.489796	-0.30906	-0.15587	0.778365	2.231868	-1.56312	1.151463	0.222162	1.025586	0.038117	-0.23275	-0.23556	-0.16476	-0.03015	0.03015
12	-0.75242	0.245485	2.057323	-1.46864	-1.15939	-0.07785	-0.60558	0.003603	0.43637	0.747731	-0.73398	-0.77041	1.047627	-1.0666	1.109593	1.660114	-0.27927	-0.41599	0.432535	0.263451	0.499625	1.35365	-0.25657	-0.06508	0.039512	-0.08709	-0.181	0.123594	0.123594
13	1.103215	-0.0403	1.267332	1.289091	-0.736	0.288069	-0.58606	0.18938	0.782333	-0.26798	-0.45031	0.936708	0.70838	-0.46865	0.354574	-0.24663	-0.00921	-0.59591	-0.57568	-0.11391	-0.02461	0.196002	0.013802	0.103758	0.364298	-0.38226	0.059809	0.037051	0.037051
13	-0.43691	0.918966	0.924591	-0.72722	0.915679	-0.12787	0.707642	0.087962	-0.66527	-0.73798	0.324098	0.277192	0.252624	-0.2919	-0.18452	1.143174	-0.92871	0.68047	0.025436	-0.04702	-0.1948	-0.67264	-0.15686	-0.88839	-0.34241	-0.04903	0.079692	0.131024	0.131024
14	-5.40126	-5.45015	1.186305	1.736239	0.309106	-1.76341	-1.55974	0.100842	1.23309	0.345173	0.91723	0.701117	-0.26657	-0.47913	-0.52661	0.472704	-0.72548	0.075081	-0.40687	-2.19685	-0.5036	0.98446	2.458389	0.042119	-0.48163	-0.62127	0.392053	0.949594	0.949594
15	1.492936	1.029315	0.847975	-1.48801	-1.55543	-0.70296	-1.08066	-0.05313	-1.97868	1.838076	1.077542	-0.63205	-0.41696	0.052011	-0.04298	0.16643	0.304241	0.554832	0.05432	-0.34793	-0.17765	-0.17507	0.040502	0.295814	0.332931	-0.22048	0.022298	0.067062	0.067062
16	0.684885	1.36182	1.059221	0.841819	-1.31211	1.309109	-0.87859	0.44529	-0.4462	0.568511	1.019153	1.298129	0.42040	-0.37265	-0.80796	0.248456	0.515663	0.625847	-1.30041	-0.13831	-0.29558	-0.57196	-0.05088	-0.30421	0.077001	-0.42223	0.086553	0.063499	0.063499
17	0.962496	0.328461	-0.17148	1.205204	1.129566	1.698038	0.107712	0.521502	-1.13931	0.724396	1.69031	0.406774	-0.95642	0.983739	0.710911	-0.60223	0.402484	-1.77318	-0.02761	-0.26932	0.143997	0.402492	0.04851	-1.37387	0.396814	0.199964	0.016371	-0.01461	0.01461
18	1.166616	0.50212	-0.0673	2.261569	0.428804	0.089474	0.341147	0.138082	0.089474	0.922175	0.744786	-0.53138	-1.25535	1.12687	0.003075	0.424425	-0.45448	-0.98887	-0.8166	-0.30717	0.018702	-0.06197	-0.10385	-0.17042	0.6032	0.108556	-0.04052	-0.01342	0.01342
18	0.247491	0.277666	1.185471	-0.0928	-1.31439	-0.15012	-0.94636	-1.61794	-1.544071	-0.82988	-0.5832	0.524993	-0.45338	0.081393	1.555204	-1.39869	0.783131	0.436621	2.177807	-0.23098	1.650118	0.200454	-0.18355	0.423073	0.820591	-0.22763	0.336634	0.250475	0.250475
22	-1.746053	-0.0449	-0.40527	-1.01306	2.941968	2.950553	-0.06306	0.855546	0.049967	0.573743	-0.08126	-0.21575	0.044161	0.033988	1.190718	0.578343	-0.97567	0.044063	0.488603	-0.21672	-0.57953	-0.79923	0.8703	0.983421	0.321201	0.14965	0.707519	0.0146	0.0146
22	-2.73429	0.12148	1.122021	0.410008	0.295196	-0.05954	0.545985	-0.10463	0.475664	0.149461	-0.85657	-0.18652	-0.05523	-0.27796	-0.11667	0.33332	0.010771	-0.40847	0.505751	-0.38669	-0.40364	-0.2274	0.742455	0.388355	0.240212	0.74404	0.359969	0.24212	0.24212
23	1.173285	0.355488	0.283905	1.133563	-0.17258	-0.91605	0.369025	-0.32726	-0.24665	-0.04614	-0.14342	0.97935	0.142385	0.101418	0.761478	0.01458	-0.51164	-0.32506	-0.39093	0.027878	0.067003	0.227612	-0.15509	0.435045	0.724825	-0.33708	0.014636	0.010041	0.010041
23	1.222977	-0.17004	0.043555	0.576038	-0.83676	-0.83108	-0.2649	-0.32298	-1.07142	0.868539	-0.64151	-0.11132	0.361485	0.171945	0.782167	-1.35587	-0.21694	1.27176	-1.24062	-0.52295	-0.28438	-0.25336	-0.37071	0.347151	0.559639	-0.28016	0.042335	0.028822	0.028822
-0.41429	0.905437	1.772453	1.474731	0.007443	-0.20031	0.740228	-0.02925	-0.59339	-0.34619	-0.01214	0.786796	0.635954	-0.08632	0.076804	-1.40592	0.775592	0.94289	0.543969	0.079708	0.077237	0.457331	-0.10385	0.642522	-0.11839	-0.27746	0.180887	0.152665	0.152665	
23	1.059387	-0.17532	1.26613	1.18611	-0.786	0.578435	-0.76708	0.401046	0.6995	-0.06474	1.048292	1.050618	-0.342	-0.03991	-0.21868	0.004476	-0.19355	0.042388	-0.27783	-0.17802	0.013676	0.231734	0.014462	0.002951	0.294638	-0.39507	0.081481	0.02442	0.02442
24	1.217429	0.061043	0.805236	0.761564	-0.35977	-0.09408	0.060494	-0.13386	0.41881	-0.20736	-0.09218	0.327106	0.348076	-0.13524	-0.21839	0.19155	-0.11658	-0.81379	0.348416	-0.06613	-0.24568	-0.5309	-0.04427	0.791848	0.509196	0.38858	-0.0227	0.011836	0.011836
25	1.134009	0.085046	0.493702	1.33376	-0.30013	-0.01075	-0.11876	0.188617	0.205487	0.082862	1.131556	0.426699	-1.48278	0.52															

DATA PREPROCESSING

Logistic regression is a fundamental and highly interpretable algorithm that offers several advantages when applied to credit card fraud detection. Its simplicity makes it an excellent starting point for developing fraud detection models. This algorithm is particularly well-suited for binary classification problems, which are inherent in fraud detection, where the goal is to distinguish between genuine and fraudulent transactions

```
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler

# Load the dataset (replace 'data.csv' with your dataset file)
data = pd.read_csv('data.csv')

# Split the data into features (X) and the target variable (y)
X = data.drop('Class', axis=1)
y = data['Class']

# Split the data into training and testing sets
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)

# Standardize the features (mean = 0, variance = 1)
scaler = StandardScaler()
X_train = scaler.fit_transform(X_train)
X_test = scaler.transform(X_test)

# Your data is now preprocessed and ready for training a fraud detection model.
```

The logistic regression algorithm is selected due to its enormous benefits notably its straightforward and intuitive model interpretation makes it valuable in fields requiring transparency, like healthcare and finance.