

CREDIT CARD FRAUD DETECTION

```
<class 'pandas.core.frame.DataFrame'>
```

```
RangeIndex: 1296675 entries, 0 to 1296674
```

```
Data columns (total 23 columns):
```

#	Column	Non-Null Count	Dtype
0	Unnamed: 0	1296675 non-null	int64
1	trans_date_trans_time	1296675 non-null	object
2	cc_num	1296675 non-null	int64
3	merchant	1296675 non-null	object
4	category	1296675 non-null	object
5	amt	1296675 non-null	float64
6	first	1296675 non-null	object
7	last	1296675 non-null	object
8	gender	1296675 non-null	object
9	street	1296675 non-null	object
10	city	1296675 non-null	object
11	state	1296675 non-null	object
12	zip	1296675 non-null	int64
13	lat	1296675 non-null	float64
14	long	1296675 non-null	float64
15	city_pop	1296675 non-null	int64
16	job	1296675 non-null	object
17	dob	1296675 non-null	object
18	trans_num	1296675 non-null	object
19	unix_time	1296675 non-null	int64
20	merch_lat	1296675 non-null	float64
21	merch_long	1296675 non-null	float64
22	is_fraud	1296675 non-null	int64

```
dtypes: float64(5), int64(6), object(12)
```

```
memory usage: 227.5+ MB
```

```
None
```

```
[5 rows x 23 columns]
Training Logistic Regression...
Logistic Regression Accuracy: 0.9955211176871764
[[553230 344]
 [ 2145 0]]
precision recall f1-score support
0 1.00 1.00 1.00 553574
1 0.99 0.99 0.99 2145

accuracy 1.00 555719
macro avg 0.50 0.50 0.50 555719
weighted avg 0.99 1.00 0.99 555719

---
Training Random Forest...
Random Forest Accuracy: 0.9971154486350116
[[553399 175]
 [ 1428 717]]
precision recall f1-score support
0 1.00 1.00 1.00 553574
1 0.80 0.33 0.47 2145

accuracy 1.00 555719
macro avg 0.90 0.67 0.74 555719
weighted avg 1.00 1.00 1.00 555719

---
```

```
None
```

```
Unnamed: 0 trans_date_trans_time cc_num \
0 0 2019-01-01 00:00:18 2701186189652095
1 1 2019-01-01 00:00:44 630423337322
2 2 2019-01-01 00:00:51 38059492057661
3 3 2019-01-01 00:01:16 3534093764340240
4 4 2019-01-01 00:03:06 375534200663984
```

```
merchant category amt first \
0 fraud_Ripkin, Kub and Mann misc_net 4.97 Jennifer
1 fraud_Heller, Gutmann and Zieme grocery_pos 107.23 Stephanie
2 fraud_Lind-Buckridge entertainment 220.11 Edward
3 fraud_Kutch, Hermiston and Farrell gas_transport 45.00 Jeremy
4 fraud_Keeling-Crist misc_pos 41.96 Tyler
```

```
last gender street ... lat long \
0 Banks F 561 Perry Cove ... 36.0788 -81.1781
1 Gill F 43039 Riley Greens Suite 393 ... 48.8878 -118.2305
2 Sanchez M 594 White Dale Suite 530 ... 42.1800 -112.2620
3 White M 9443 Cynthia Court Apt. 038 ... 46.2306 -112.1138
4 Garcia M 408 Bradley Rest ... 38.4207 -79.4629
```

```
city_pop job dob \
0 3495 Psychologist, counselling 1988-03-09
1 149 Special educational needs teacher 1978-06-21
2 4154 Nature conservation officer 1962-01-19
3 1939 Patent attorney 1967-01-12
4 99 Dance movement psychotherapist 1986-03-28
```

```
trans_num unix_time merch_lat merch_long \
0 0b242abb621afc57057560ndf30655b9 1325376018 36.411293 -82.048315
1 1f76529f8574734946361c461b024d99 1325376044 49.159047 -118.186462
2 a1a22d70485983eac12b5b88dadcf95 1325376051 43.150704 -112.154481
3 6b849c168bdad6f867558c3793159a81 1325376076 47.434331 -112.561071
4 a41d7549acf90789359e8a5346dcb46 1325376186 38.674999 -78.632459
```

```
is_fraud
0 0
1 0
2 0
3 0
4 0
```

SPAM SMS DETECTION

SPAM SMS DETECTION

```
Index(['v1', 'v2', 'Unnamed: 2', 'Unnamed: 3', 'Unnamed: 4'], dtype='object')
```

Naive Bayes Accuracy: 0.9722

Naive Bayes Classification Report:

	precision	recall	f1-score	support
0	0.97	1.00	0.98	965
1	1.00	0.79	0.88	150
accuracy			0.97	1115
macro avg	0.98	0.90	0.93	1115
weighted avg	0.97	0.97	0.97	1115

Logistic Regression Accuracy: 0.9578

Logistic Regression Classification Report:

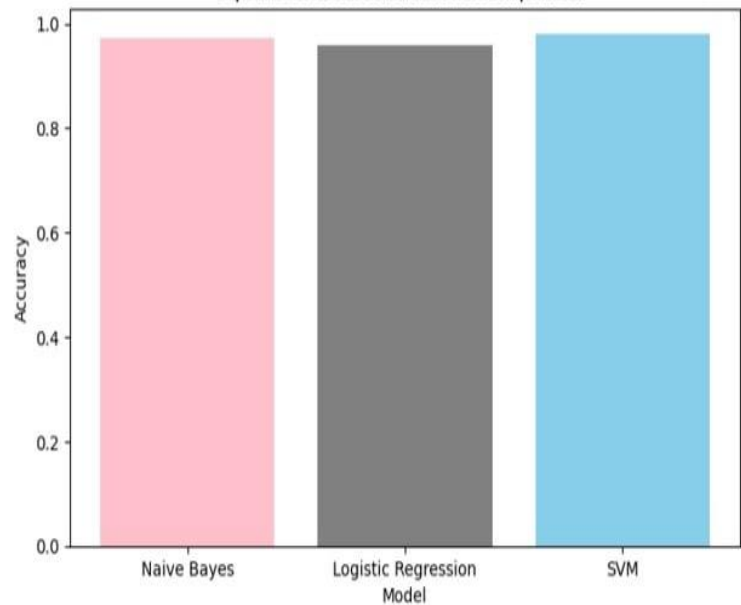
	precision	recall	f1-score	support
0	0.96	1.00	0.98	965
1	0.97	0.71	0.82	150
accuracy			0.96	1115
macro avg	0.96	0.85	0.90	1115
weighted avg	0.96	0.96	0.95	1115

SVM Accuracy: 0.9794

SVM Classification Report:

	precision	recall	f1-score	support
0	0.98	1.00	0.99	965
1	0.97	0.87	0.92	150
accuracy			0.98	1115
macro avg	0.98	0.93	0.95	1115
weighted avg	0.98	0.98	0.98	1115

Spam SMS Detection Model Comparison



CUSTOMER CHURN PREDICTION

CUSTOMER CHURN PREDICTION

dataset information:

```
<class 'pandas.core.frame.DataFrame'>  
RangeIndex: 10000 entries, 0 to 9999  
Data columns (total 14 columns):  
#   Column      Non-Null Count  Dtype  
---  ---  
0   RowNumber    10000 non-null  int64  
1   CustomerId   10000 non-null  int64  
2   Surname      10000 non-null  object  
3   CreditScore   10000 non-null  int64  
4   Geography    10000 non-null  object  
5   Gender       10000 non-null  object  
6   Age          10000 non-null  int64  
7   Tenure       10000 non-null  int64  
8   Balance      10000 non-null  float64  
9   NumOfProducts 10000 non-null  int64  
10  HasCrCard    10000 non-null  int64  
11  IsActiveMember 10000 non-null  int64  
12  EstimatedSalary 10000 non-null float64  
13  Exited       10000 non-null  int64  
dtypes: float64(2), int64(9), object(3)
```

memory usage: 1.1+ MB

None

	RowNumber	CustomerId	Surname	CreditScore	Geography	Gender	Age
0	1	15634602	Hargrave	619	France	Female	42
1	2	15647311	Hill	688	Spain	Female	41
2	3	15619384	Onio	582	France	Female	42
3	4	15701354	Boni	699	France	Female	39
4	5	15737888	Mitchell	850	Spain	Female	43

	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember
0	2	0.00	1	1	1
1	1	83807.86	1	0	1
2	8	159660.88	3	1	0
3	1	0.00	2	0	0
4	2	125510.82	1	1	1

	EstimatedSalary	Exited
0	101340.88	1
1	112542.50	0
2	113911.57	1
3	93826.63	0
4	79084.10	0

(10000, 14)

Logistic Regression Accuracy: 0.8155

Random Forest Accuracy: 0.864

Gradient Boosting Accuracy: 0.866

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CUSTOMER CHURN PREDICTION

