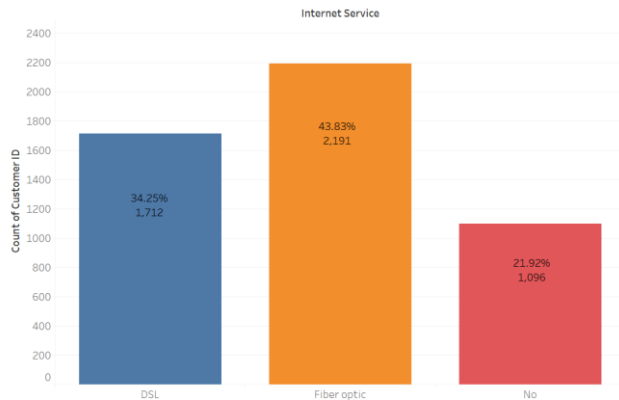


# Digital Services Case Study

## Distribution of Customers based on the type of internet service

Distiribution

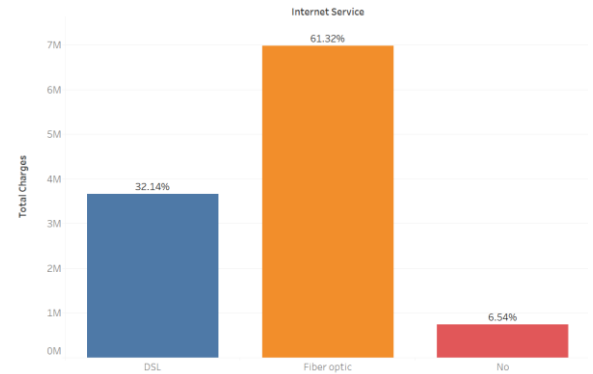


There are about **22%** of customers who do not avail Internet service.

## Net Spend

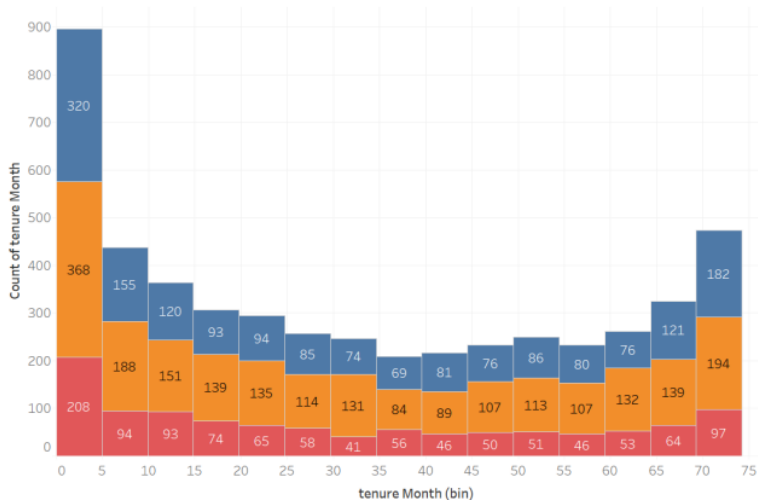
Internet ..	Avg. Total Charges	Avg. Monthly Charges
Fiber optic	3,183	92
DSL	2,136	59
No	679	21

Conitribution to Revenue



They contribute to **6.54%** of the total revenues .

## Impact of the type of internet Services



The tenure of services availed is spread evenly across the three types of internet services, indicating that the duration of contract with the **company does not depend** on the type of service.

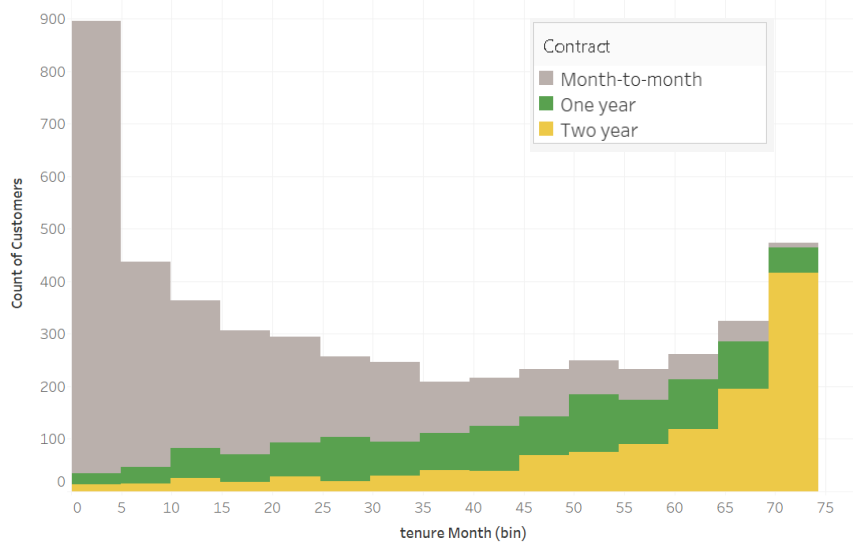
## Churn rate for Internet Service Types

	Internet Service		
Churn	DSL	Fiber optic	No
No	81.25%	58.19%	93.07%
	1,391	1,275	1,020
Yes	18.75%	41.81%	6.93%
	321	916	76

Indicates the Churn rate for **Fiber optic** to be the highest.

And the Churn rate for People who **do not** avail internet services to be relatively negligible.

Distribution of Customers based on the type of Contract



As observed in the graph.

Customers who tend to stay with the company for a longer time are usually on a one year or two year contract.

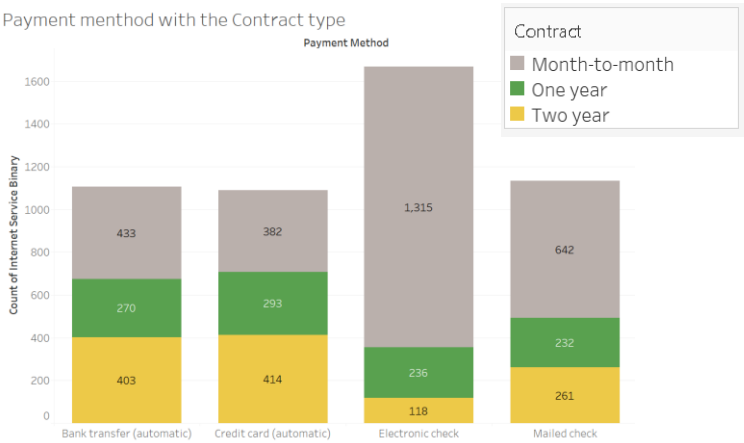
Churn Rate by Contract Type

Churn	Month-to-month	Contract One year	Two year
No	57.72% 1,600	89.04% 918	97.66% 1,168
Yes	42.28% 1,172	10.96% 113	2.34% 28

Indicates the Churn rate for **Contract on Month basis to be higher**

Churn rate for customers who avail a long term contract is lesser.

Distribution of Customers based on the type of Payment Method

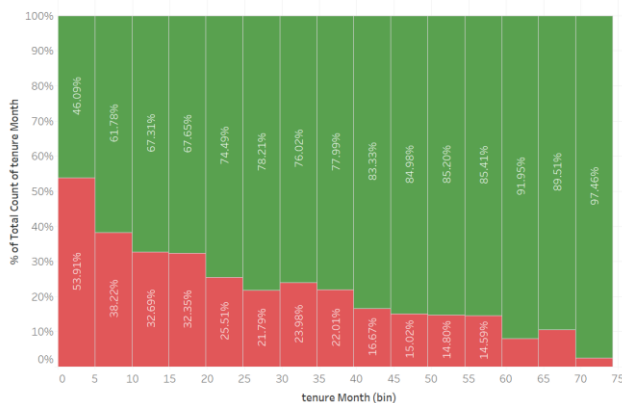


Shows the distribution of type of payments.

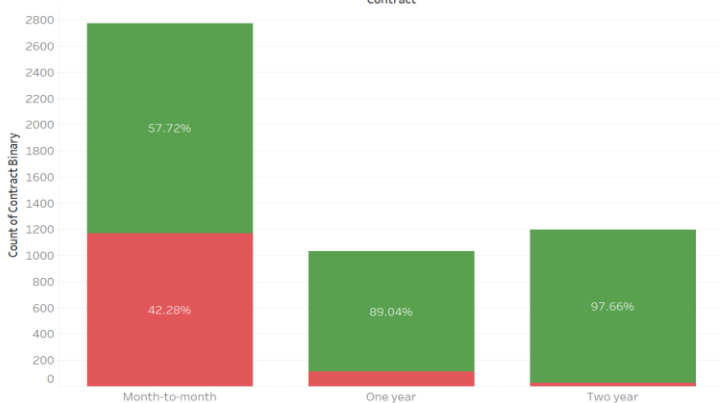
Here most of the electronic payments are done by monthly customers and this also impacts the churn rate.

The Key Drivers for Churn Rate here are

Tenure VsChurn

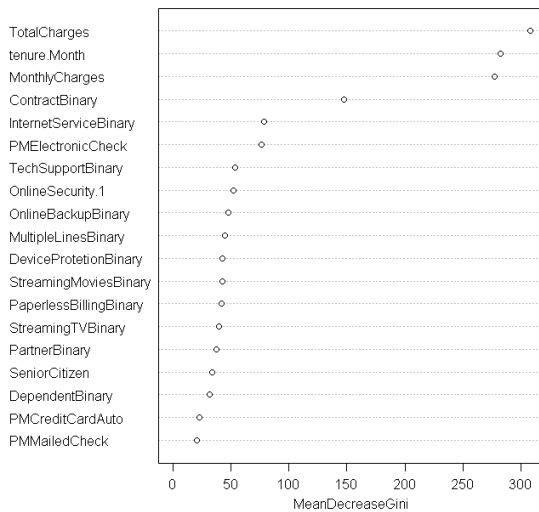


Contract Vs Churn



## Further Analysis using Statistical techniques indicate a similar story

Using Random Forest and checking the importance of variables and its impact on the churn rate, below ranks the effect of variables on the churn.



```
call:
  randomForest(formula = Churn ~ . - GenderBinary - PhoneServiceBinary, data = DSRf, importance = 1
E, proximity = TRUE)
  Type of random forest: classification
    Number of trees: 500
  No. of variables tried at each split: 4

  OOB estimate of error rate: 20.16%
Confusion matrix:
      No Yes class.error
No  3283 403  0.1093326
Yes   605 708  0.4607768
```

Churn rate prediction using Random forest is 80%.

## Using Logistic Regression.: Importance of variables

```
call:
glm(formula = Training$ChurnBinary ~ . - PMCreditCardAuto - PMMAiledCheck -
  DeviceProtetionBinary - OnlineBackupBinary - PhoneServiceBinary -
  DependentBinary - PartnerBinary - SeniorCitizen - GenderBinary,
  family = binomial, data = Training)
```

```
Deviance Residuals:
    Min       1Q   Median       3Q      Max
-1.8039  -0.6770  -0.2928   0.7454   3.3616
```

```
Coefficients:
              Estimate Std. Error z value Pr(>|z|)
(Intercept)  -1.024e+00  1.884e-01  -5.437 5.42e-08 ***
tenure.Month  -5.684e-02  8.300e-03  -6.848 7.48e-12 ***
MultipleLinesBinary  3.250e-01  1.185e-01   2.744 0.006075 **
InternetServiceBinary  1.635e+00  1.902e-01   8.598 < 2e-16 ***
OnlineSecurity.1  -3.227e-01  1.069e-01  -3.019 0.002540 **
TechSupportBinary  -4.825e-01  1.130e-01  -4.270 1.95e-05 ***
StreamingTVBinary  4.406e-01  1.256e-01   3.508 0.000452 ***
StreamingMoviesBinary  5.779e-01  1.243e-01   4.649 3.34e-06 ***
ContractBinary    -6.313e-01  1.093e-01  -5.775 7.67e-09 ***
PaperlessBillingBinary  4.457e-01  1.051e-01   4.242 2.22e-05 ***
PMElectronicCheck  3.966e-01  9.742e-02   4.071 4.68e-05 ***
MonthlyCharges    -3.700e-02  7.841e-03  -4.718 2.38e-06 ***
TotalCharges       3.498e-04  9.484e-05   3.688 0.000226 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

(Dispersion parameter for binomial family taken to be 1)

```
Null deviance: 4047.3 on 3507 degrees of freedom
Residual deviance: 2927.7 on 3495 degrees of freedom
AIC: 2953.7
```

Number of Fisher Scoring iterations: 6

Churn rate prediction using Logistic regression is 81.1%

```
> table(ChurnValuesPredict,Testing$ChurnBinary)
```

```
ChurnValuesPredict    0    1
                   0 996 174
                   1 107 214
```