

Task 3

Sub task – 1:

Think through what key drivers of churn could be for our client

```
proc logistic data=work.inner_join01;
model churn(event='1') = cons_12m cons_gas_12m cons_last_month forecast_base_bill_ele
forecast_base_bill_year forecast_bill_12m forecast_cons forecast_cons_12m
forecast_cons_year forecast_discount_energy forecast_price_energy_p1
forecast_meter_rent_12m forecast_price_energy_p2 forecast_price_pow_p1
margin_gross_pow_ele margin_net_pow_ele nb_prod_act net_margin num_years_antig
pow_max / link=probit rsquare;
run;
```

Parameter	DF	Estimate	Standard Error	Wald Chi-Square	Pr > ChiSq
Intercept	1	-1.2531	0.3203	15.3056	<.0001
cons_12m	1	2.508E-8	2.816E-7	0.0079	0.9290
cons_gas_12m	1	-7.48E-7	3.099E-7	5.8216	0.0158
cons_last_month	1	-2.5E-6	2.679E-6	0.8704	0.3508
forecast_base_bill_e	1	-0.00002	0.000293	0.0062	0.9375
forecast_base_bill_y	0	0	.	.	.
forecast_bill_12m	1	0.000025	0.000053	0.2193	0.6396
forecast_cons	1	3.295E-6	0.000456	0.0001	0.9942
forecast_cons_12m	1	-0.00004	0.000062	0.3590	0.5490
forecast_cons_year	1	7.326E-7	0.000034	0.0005	0.9829
forecast_discount_en	1	-0.00548	0.00544	1.0121	0.3144
forecast_price_energ	1	1.1050	1.7367	0.4048	0.5246
forecast_meter_rent_	1	0.00150	0.000626	5.7805	0.0162
forecast_price_energ	1	0.4619	0.8493	0.2958	0.5866
forecast_price_pow_p	1	0.00857	0.00693	1.5289	0.2163
margin_gross_pow_ele	1	0.00457	0.00156	8.5666	0.0034
margin_net_pow_ele	1	0.000365	0.00120	0.0926	0.7609
nb_prod_act	1	0.00624	0.0458	0.0185	0.8917
net_margin	1	0.000118	0.000078	2.2734	0.1316
num_years_antig	1	-0.1190	0.0233	26.0703	<.0001
pow_max	1	-0.00257	0.00316	0.6630	0.4155

Base on Maximum likelihood estimation, we find that num_years_antig has the significant relation with churn variable.

```
proc freq data=work.only_churn nlevels;
tables churn*num_years_antig;
run;
```

Table of churn by num_years_antig										
churn	num_years_antig									
	2	3	4	5	6	7	8	10	11	Total
1	1	161	196	72	71	2	1	1	1	506
	0.20	31.82	38.74	14.23	14.03	0.40	0.20	0.20	0.20	100.00
	0.20	31.82	38.74	14.23	14.03	0.40	0.20	0.20	0.20	
	100.00	100.00	100.00	100.00	100.00	100.00	100.00	100.00	100.00	
Total	1	161	196	72	71	2	1	1	1	506
	0.20	31.82	38.74	14.23	14.03	0.40	0.20	0.20	0.20	100.00

Observing the table, it shows little positive skewed distributions where we have maximum churn rate when antiquity of the client is 3 to 4 years old.

```
proc logistic data=work.price inner join;
model num_years_antig = price_p1_fix price_p1_var price_p2_fix
                        price_p2_var price_p3_fix price_p3_var;
run;
```

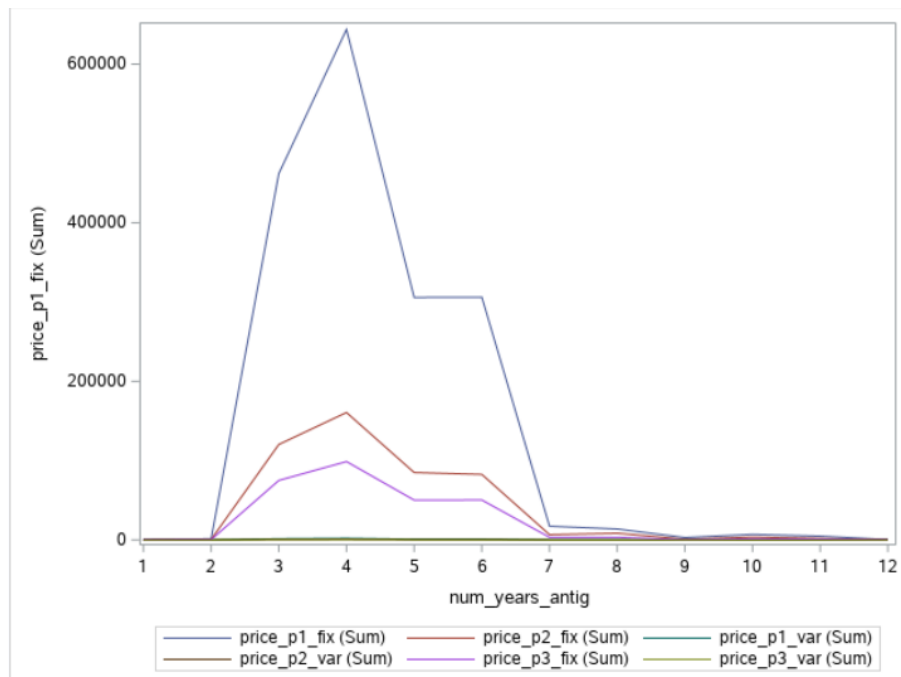
Analysis of Maximum Likelihood Estimates						
Parameter		DF	Estimate	Standard Error	Wald Chi-Square	Pr > ChiSq
Intercept	1	1	-13.6442	0.3622	1418.9355	<.0001
Intercept	2	1	-11.9381	0.1703	4915.9391	<.0001
Intercept	3	1	-6.0884	0.0786	6005.2600	<.0001
Intercept	4	1	-4.5044	0.0779	3342.3974	<.0001
Intercept	5	1	-3.6260	0.0775	2190.4638	<.0001
Intercept	6	1	-1.3594	0.0723	353.9363	<.0001
Intercept	7	1	-0.8264	0.0700	139.3874	<.0001
Intercept	8	1	-0.1907	0.0682	7.8237	0.0052
Intercept	9	1	1.3482	0.0809	277.7379	<.0001
Intercept	10	1	2.1305	0.1043	417.5453	<.0001
Intercept	11	1	4.4339	0.2927	229.5158	<.0001
price_p1_fix		1	0.0771	0.00341	511.0723	<.0001
price_p1_var		1	10.9316	0.8404	169.2051	<.0001
price_p2_fix		1	-0.1144	0.00660	300.5977	<.0001
price_p2_var		1	-1.9031	0.4409	18.6333	<.0001
price_p3_fix		1	0.0650	0.00862	56.7506	<.0001
price_p3_var		1	33.3734	3.2206	107.3838	<.0001

Now this table shows the significant relations between the prices of power and energy of different period with regards to antiquity of the client (in number of years).

```

proc sgplot data = work.price_inner_join;
vline num_years_antig/response = price_p1_fix;
vline num_years_antig/response = price_p2_fix;
vline num_years_antig/response = price_p1_var;
vline num_years_antig/response = price_p2_var;
vline num_years_antig/response = price_p3_fix;
vline num_years_antig/response = price_p3_var;
run;

```



Above graph shows prices of the power and energy is highest in the 3 to 5 num_years_antig where maximum number of churn rate occur.

Sub-task 2

The following input variables or features must include when building a model are:

1. Price_p1_fix
2. Price_p2_fix
3. Price_p3_fix
4. Num_year_antig
5. margin_gross_pow_ele
6. forecast_meter_rent_12m
7. forecast_meter_rent_12m

these variable show significant probability when comparing with churn rate. First table.