

```
In [2]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
```

```
In [4]: # Load data
client_data = pd.read_csv('client_data (1).csv')
price_data = pd.read_csv('price_data (1).csv')
```

```
In [7]: client_data.head()
```

Out[7]:

	id	channel_sales	cons_12m	cons_gas_12m	cons_last_month	date_activ
0	24011ae4ebbe3035111d65fa7c15bc57	foosdfpfkusacimwkcsoibcdxkicaua	0	54946	0	2013-06-15
1	d29c2c54acc38ff3c0614d0a653813dd	MISSING	4660	0	0	2009-08-21
2	764c75f661154dac3a6c254cd082ea7d	foosdfpfkusacimwkcsoibcdxkicaua	544	0	0	2010-04-16
3	bba03439a292a1e166f80264c16191cb	lmkebamcaaclubfxadlmueccxoimlema	1584	0	0	2010-03-30
4	149d57cf92fc41cf94415803a877cb4b	MISSING	4425	0	526	2010-01-13

5 rows × 7 columns

```
In [8]: price_data.head()
```

Out[8]:

	id	price_date	price_off_peak_var	price_peak_var	price_mid_peak_var	price_off_peak_fix	price
0	038af19179925da21a25619c5a24b745	2015-01-01	0.151367	0.0	0.0	44.266931	
1	038af19179925da21a25619c5a24b745	2015-02-01	0.151367	0.0	0.0	44.266931	
2	038af19179925da21a25619c5a24b745	2015-03-01	0.151367	0.0	0.0	44.266931	
3	038af19179925da21a25619c5a24b745	2015-04-01	0.149626	0.0	0.0	44.266931	
4	038af19179925da21a25619c5a24b745	2015-05-01	0.149626	0.0	0.0	44.266931	

```
In [9]: client_data.info()
```

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

```
RangeIndex: 14606 entries, 0 to 14605
```

```
Data columns (total 26 columns):
```

#	Column	Non-Null Count	Dtype
0	id	14606 non-null	object
1	channel_sales	14606 non-null	object
2	cons_12m	14606 non-null	int64
3	cons_gas_12m	14606 non-null	int64
4	cons_last_month	14606 non-null	int64
5	date_activ	14606 non-null	object
6	date_end	14606 non-null	object
7	date_modif_prod	14606 non-null	object
8	date_renewal	14606 non-null	object
9	forecast_cons_12m	14606 non-null	float64
10	forecast_cons_year	14606 non-null	int64
11	forecast_discount_energy	14606 non-null	float64
12	forecast_meter_rent_12m	14606 non-null	float64
13	forecast_price_energy_off_peak	14606 non-null	float64
14	forecast_price_energy_peak	14606 non-null	float64
15	forecast_price_pow_off_peak	14606 non-null	float64
16	has_gas	14606 non-null	object
17	imp_cons	14606 non-null	float64
18	margin_gross_pow_ele	14606 non-null	float64
19	margin_net_pow_ele	14606 non-null	float64
20	nb_prod_act	14606 non-null	int64
21	net_margin	14606 non-null	float64
22	num_years_antig	14606 non-null	int64
23	origin_up	14606 non-null	object
24	pow_max	14606 non-null	float64
25	churn	14606 non-null	int64

```
dtypes: float64(11), int64(7), object(8)
```

```
memory usage: 2.9+ MB
```

```
In [10]: client_data.dtypes
```

```
Out[10]: id                                object
channel_sales                             object
cons_12m                                   int64
cons_gas_12m                               int64
cons_last_month                           int64
date_activ                                object
date_end                                  object
date_modif_prod                           object
date_renewal                              object
forecast_cons_12m                         float64
forecast_cons_year                        int64
forecast_discount_energy                  float64
forecast_meter_rent_12m                  float64
forecast_price_energy_off_peak            float64
forecast_price_energy_peak                float64
forecast_price_pow_off_peak               float64
has_gas                                  object
imp_cons                                 float64
margin_gross_pow_ele                     float64
margin_net_pow_ele                       float64
nb_prod_act                              int64
net_margin                               float64
num_years_antig                          int64
origin_up                                object
pow_max                                  float64
churn                                    int64
dtype: object
```

```
In [12]: print(client_data.describe())
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 14606 entries, 0 to 14605
Data columns (total 26 columns):
 #   Column                                Non-Null Count  Dtype

```

0	id	14606	non-null	object
1	channel_sales	14606	non-null	object
2	cons_12m	14606	non-null	int64
3	cons_gas_12m	14606	non-null	int64
4	cons_last_month	14606	non-null	int64
5	date_activ	14606	non-null	object
6	date_end	14606	non-null	object
7	date_modif_prod	14606	non-null	object
8	date_renewal	14606	non-null	object
9	forecast_cons_12m	14606	non-null	float64
10	forecast_cons_year	14606	non-null	int64
11	forecast_discount_energy	14606	non-null	float64
12	forecast_meter_rent_12m	14606	non-null	float64
13	forecast_price_energy_off_peak	14606	non-null	float64
14	forecast_price_energy_peak	14606	non-null	float64
15	forecast_price_pow_off_peak	14606	non-null	float64
16	has_gas	14606	non-null	object
17	imp_cons	14606	non-null	float64
18	margin_gross_pow_ele	14606	non-null	float64
19	margin_net_pow_ele	14606	non-null	float64
20	nb_prod_act	14606	non-null	int64
21	net_margin	14606	non-null	float64
22	num_years_antig	14606	non-null	int64
23	origin_up	14606	non-null	object
24	pow_max	14606	non-null	float64
25	churn	14606	non-null	int64

dtypes: float64(11), int64(7), object(8)

memory usage: 2.9+ MB

Client Data Preview:

	id	channel_sales	\
0	24011ae4ebbe3035111d65fa7c15bc57	foosdfpfkusacimwkcso	bicdxkica
1	d29c2c54acc38ff3c0614d0a653813dd		MISSING
2	764c75f661154dac3a6c254cd082ea7d	foosdfpfkusacimwkcso	bicdxkica
3	bba03439a292a1e166f80264c16191cb	lmkebamcaclubfxadlmueccxoim	lema
4	149d57cf92fc41cf94415803a877cb4b		MISSING

	cons_12m	cons_gas_12m	cons_last_month	date_activ	date_end	\
0	0	54946	0	2013-06-15	2016-06-15	
1	4660	0	0	2009-08-21	2016-08-30	
2	544	0	0	2010-04-16	2016-04-16	
3	1584	0	0	2010-03-30	2016-03-30	
4	4425	0	526	2010-01-13	2016-03-07	

	date_modif_prod	date_renewal	forecast_cons_12m	...	has_gas	imp_cons	\
0	2015-11-01	2015-06-23	0.00	...	t	0.00	
1	2009-08-21	2015-08-31	189.95	...	f	0.00	
2	2010-04-16	2015-04-17	47.96	...	f	0.00	
3	2010-03-30	2015-03-31	240.04	...	f	0.00	
4	2010-01-13	2015-03-09	445.75	...	f	52.32	

	margin_gross_pow_ele	margin_net_pow_ele	nb_prod_act	net_margin	\
0	25.44	25.44	2	678.99	
1	16.38	16.38	1	18.89	
2	28.60	28.60	1	6.60	
3	30.22	30.22	1	25.46	
4	44.91	44.91	1	47.98	

	num_years_antig	origin_up	pow_max	churn
0	3	lxidpiddsbxsbosboudacockeimpuepw	43.648	1
1	6	kamkkxfxxuwbdslkwifmmcsiusuosws	13.800	0
2	6	kamkkxfxxuwbdslkwifmmcsiusuosws	13.856	0
3	6	kamkkxfxxuwbdslkwifmmcsiusuosws	13.200	0
4	6	kamkkxfxxuwbdslkwifmmcsiusuosws	19.800	0

[5 rows x 26 columns]

Price Data Info:

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 193002 entries, 0 to 193001
Data columns (total 8 columns):
```

#	Column	Non-Null Count	Dtype
0	id	193002 non-null	object
1	price_date	193002 non-null	object
2	price_off_peak_var	193002 non-null	float64
3	price_peak_var	193002 non-null	float64
4	price_mid_peak_var	193002 non-null	float64
5	price_off_peak_fix	193002 non-null	float64
6	price_peak_fix	193002 non-null	float64
7	price_mid_peak_fix	193002 non-null	float64

dtypes: float64(6), object(2)

memory usage: 11.8+ MB

Price Data Preview:

	id	price_date	price_off_peak_var	\
0	038af19179925da21a25619c5a24b745	2015-01-01	0.151367	
1	038af19179925da21a25619c5a24b745	2015-02-01	0.151367	
2	038af19179925da21a25619c5a24b745	2015-03-01	0.151367	
3	038af19179925da21a25619c5a24b745	2015-04-01	0.149626	
4	038af19179925da21a25619c5a24b745	2015-05-01	0.149626	

	price_peak_var	price_mid_peak_var	price_off_peak_fix	price_peak_fix	\
0	0.0	0.0	44.266931	0.0	
1	0.0	0.0	44.266931	0.0	
2	0.0	0.0	44.266931	0.0	
3	0.0	0.0	44.266931	0.0	
4	0.0	0.0	44.266931	0.0	

	price_mid_peak_fix
0	0.0
1	0.0
2	0.0
3	0.0
4	0.0

Client Data Description:

	id	channel_sales	\
count	14606	14606	
unique	14606	8	
top	563dde550fd624d7352f3de77c0cdfcd	foosdfpfkusacimwkcsosbicdxkicaau	
freq	1	6754	
mean	NaN	NaN	
std	NaN	NaN	
min	NaN	NaN	
25%	NaN	NaN	
50%	NaN	NaN	
75%	NaN	NaN	
max	NaN	NaN	

	cons_12m	cons_gas_12m	cons_last_month	date_activ	date_end	\
count	1.460600e+04	1.460600e+04	14606.000000	14606	14606	
unique	NaN	NaN	NaN	1796	368	
top	NaN	NaN	NaN	2009-08-01	2016-02-01	
freq	NaN	NaN	NaN	95	145	
mean	1.592203e+05	2.809238e+04	16090.269752	NaN	NaN	
std	5.734653e+05	1.629731e+05	64364.196422	NaN	NaN	
min	0.000000e+00	0.000000e+00	0.000000	NaN	NaN	
25%	5.674750e+03	0.000000e+00	0.000000	NaN	NaN	
50%	1.411550e+04	0.000000e+00	792.500000	NaN	NaN	
75%	4.076375e+04	0.000000e+00	3383.000000	NaN	NaN	
max	6.207104e+06	4.154590e+06	771203.000000	NaN	NaN	

	date_modif_prod	date_renewal	forecast_cons_12m	...	has_gas	\
count	14606	14606	14606.000000	...	14606	
unique	2129	386	NaN	...	2	
top	2015-11-01	2015-06-23	NaN	...	f	
freq	721	587	NaN	...	11955	
mean	NaN	NaN	1868.614880	...	NaN	
std	NaN	NaN	2387.571531	...	NaN	
min	NaN	NaN	0.000000	...	NaN	
25%	NaN	NaN	494.995000	...	NaN	
50%	NaN	NaN	1112.875000	...	NaN	
75%	NaN	NaN	2401.790000	...	NaN	
max	NaN	NaN	82902.830000	...	NaN	

	imp_cons	margin_gross_pow_ele	margin_net_pow_ele	nb_prod_act	\
count	14606.000000	14606.000000	14606.000000	14606.000000	
unique	NaN	NaN	NaN	NaN	
top	NaN	NaN	NaN	NaN	
freq	NaN	NaN	NaN	NaN	
mean	152.786896	24.565121	24.562517	1.292346	
std	341.369366	20.231172	20.230280	0.709774	
min	0.000000	0.000000	0.000000	1.000000	
25%	0.000000	14.280000	14.280000	1.000000	
50%	37.395000	21.640000	21.640000	1.000000	
75%	193.980000	29.880000	29.880000	1.000000	
max	15042.790000	374.640000	374.640000	32.000000	

	net_margin	num_years_antig	origin_up	\
count	14606.000000	14606.000000	14606	
unique	NaN	NaN	6	
top	NaN	NaN	lxidpiddsbxsbosboudacockeimpuepw	
freq	NaN	NaN	7097	
mean	189.264522	4.997809	NaN	

std	311.798130	1.611749	NaN
min	0.000000	1.000000	NaN
25%	50.712500	4.000000	NaN
50%	112.530000	5.000000	NaN
75%	243.097500	6.000000	NaN
max	24570.650000	13.000000	NaN

	pow_max	churn
count	14606.000000	14606.000000
unique	NaN	NaN
top	NaN	NaN
freq	NaN	NaN
mean	18.135136	0.097152
std	13.534743	0.296175
min	3.300000	0.000000
25%	12.500000	0.000000
50%	13.856000	0.000000
75%	19.172500	0.000000
max	320.000000	1.000000

[11 rows x 26 columns]

Price Data Description:

	id	price_date	price_off_peak_var	\
count	193002	193002	193002.000000	
unique	16096	12	NaN	
top	c18b6305122e4976739b8420d5b54ec5	2015-12-01	NaN	
freq	12	16094	NaN	
mean	NaN	NaN	0.141027	
std	NaN	NaN	0.025032	
min	NaN	NaN	0.000000	
25%	NaN	NaN	0.125976	
50%	NaN	NaN	0.146033	
75%	NaN	NaN	0.151635	
max	NaN	NaN	0.280700	

	price_peak_var	price_mid_peak_var	price_off_peak_fix	\
count	193002.000000	193002.000000	193002.000000	
unique	NaN	NaN	NaN	
top	NaN	NaN	NaN	
freq	NaN	NaN	NaN	
mean	0.054630	0.030496	43.334477	
std	0.049924	0.036298	5.410297	
min	0.000000	0.000000	0.000000	
25%	0.000000	0.000000	40.728885	
50%	0.085483	0.000000	44.266930	
75%	0.101673	0.072558	44.444710	
max	0.229788	0.114102	59.444710	

	price_peak_fix	price_mid_peak_fix
count	193002.000000	193002.000000
unique	NaN	NaN
top	NaN	NaN
freq	NaN	NaN
mean	10.622875	6.409984
std	12.841895	7.773592
min	0.000000	0.000000
25%	0.000000	0.000000
50%	0.000000	0.000000
75%	24.339581	16.226389
max	36.490692	17.458221

Unique values in client_data:

id	14606
channel_sales	8
cons_12m	11065
cons_gas_12m	2112
cons_last_month	4751
date_activ	1796
date_end	368
date_modif_prod	2129
date_renewal	386
forecast_cons_12m	13993
forecast_cons_year	4218
forecast_discount_energy	12
forecast_meter_rent_12m	3528
forecast_price_energy_off_peak	516
forecast_price_energy_peak	329
forecast_price_pow_off_peak	41
has_gas	2
imp_cons	7752
margin_gross_pow_ele	2391
margin_net_pow_ele	2391

```

nb_prod_act          10
net_margin           11965
num_years_antig      13
origin_up            6
pow_max             698
churn                2
dtype: int64

```

```

Unique values in price_data:
id          16096
price_date   12
price_off_peak_var  1853
price_peak_var   1189
price_mid_peak_var   711
price_off_peak_fix   66
price_peak_fix     31
price_mid_peak_fix   28
dtype: int64

```

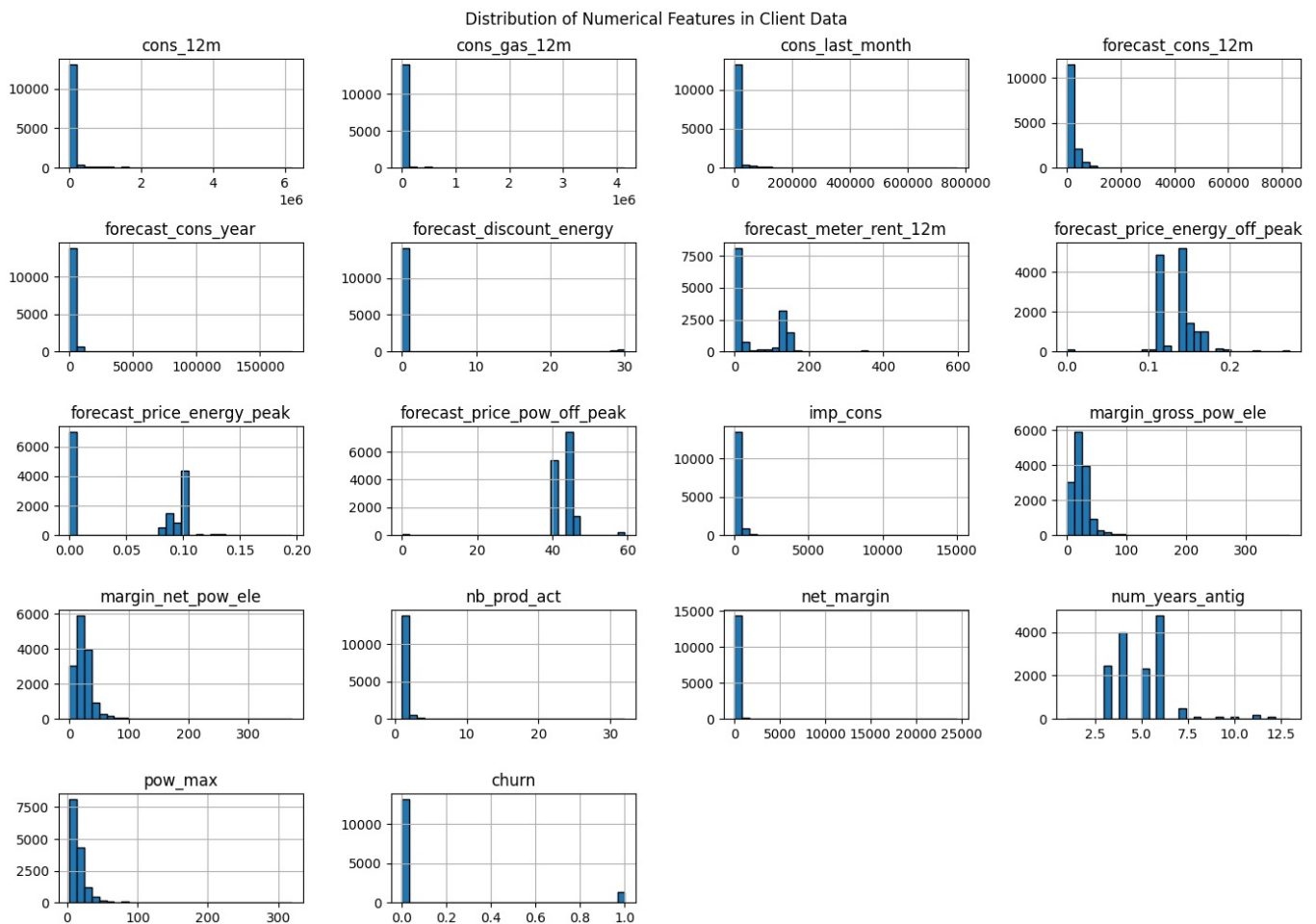
```

In [14]: # Convert date columns if needed
if 'purchase_date' in price_data.columns:
    price_data['purchase_date'] = pd.to_datetime(price_data['purchase_date'])

# Plot numerical distributions from client_data
num_cols = client_data.select_dtypes(include=['int64', 'float64']).columns
client_data[num_cols].hist(bins=30, figsize=(14, 10), edgecolor='black')
plt.suptitle('Distribution of Numerical Features in Client Data')
plt.tight_layout()
plt.show()

# Plot churn distribution if exists
if 'churned' in client_data.columns:
    sns.countplot(x='churned', data=client_data)
    plt.title("Churn Distribution")
    plt.show()

```



```

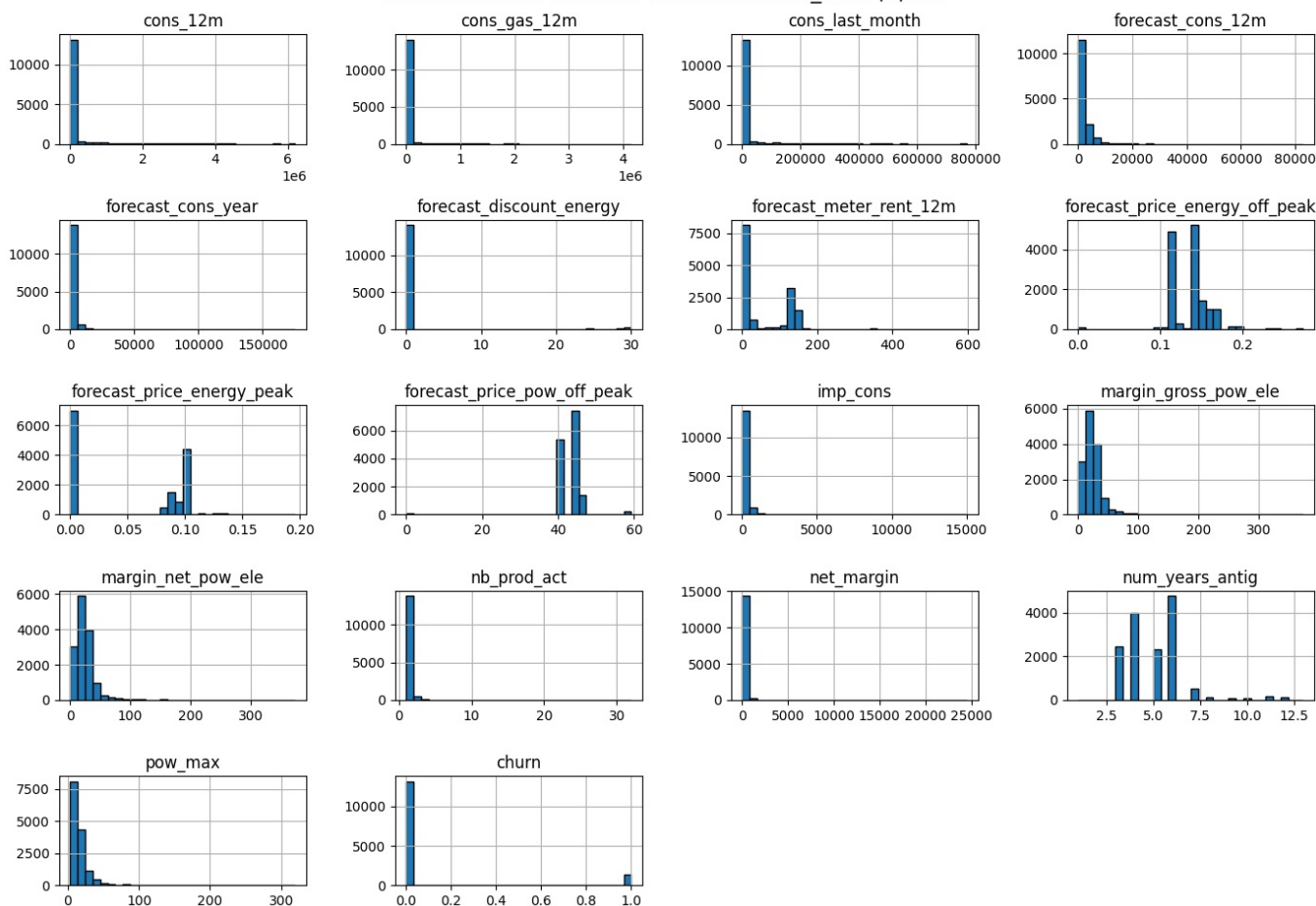
In [22]: def plot_distributions(df, title):
    num_cols = df.select_dtypes(include=['int64', 'float64']).columns
    df[num_cols].hist(bins=30, figsize=(14, 10), edgecolor='black')
    plt.suptitle(f'Distributions of Numeric Columns: {title}', fontsize=16)
    plt.tight_layout()
    plt.show()

plot_distributions(client_data, "client_data (1).csv")
plot_distributions(price_data, "price_data (1).csv")

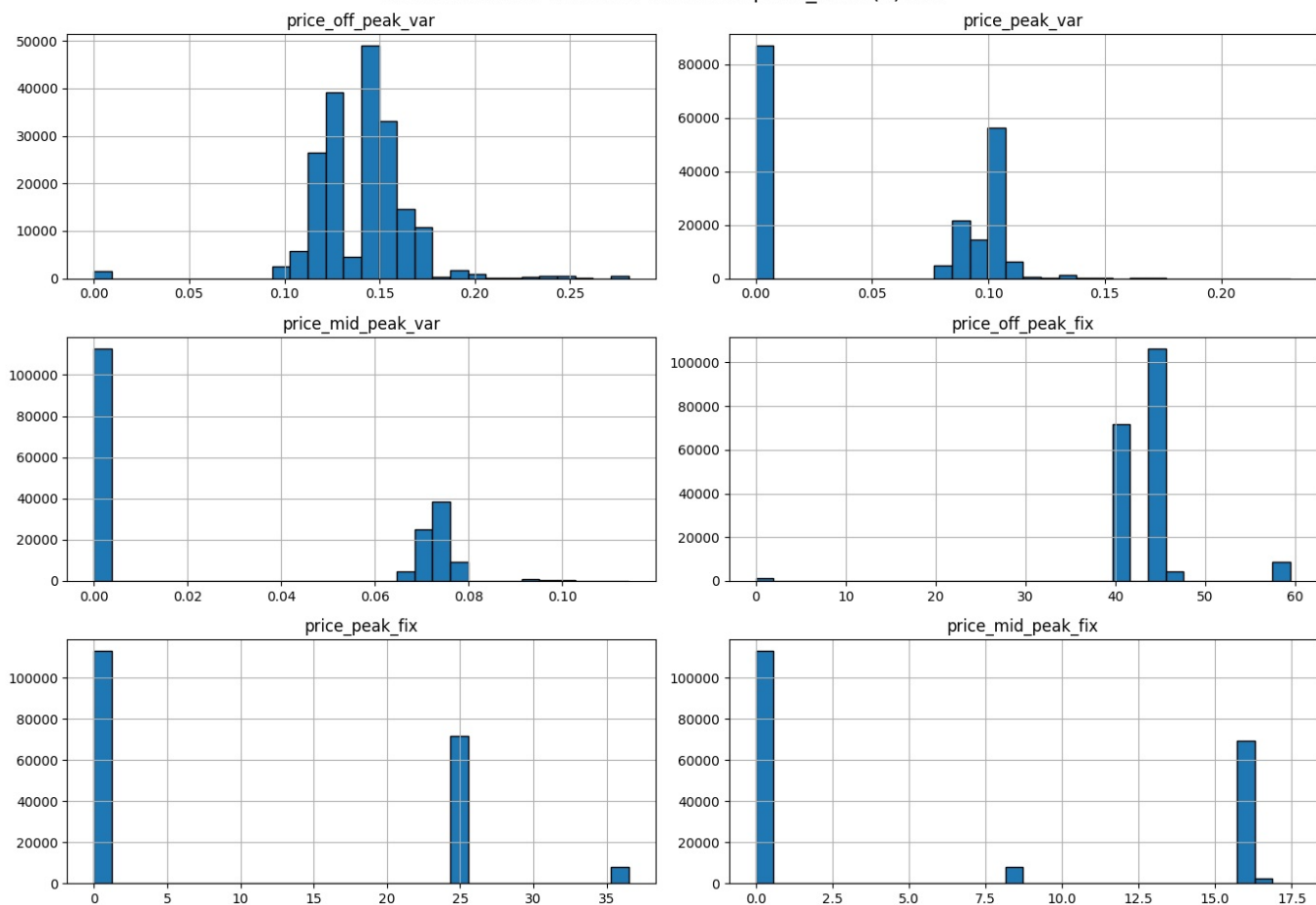
```

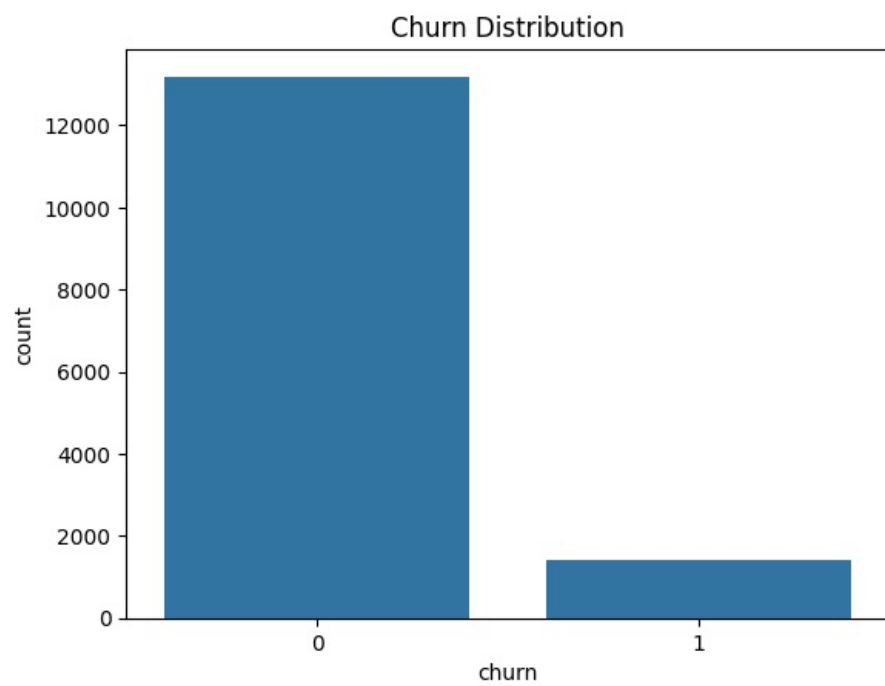
```
# 8. Churn visualization (if it's inside client data)
if 'churn' in client_data.columns or 'churned' in client_data.columns:
    churn_col = 'churn' if 'churn' in client_data.columns else 'churned'
    sns.countplot(x=churn_col, data=client_data)
    plt.title("Churn Distribution")
    plt.show()
```

Distributions of Numeric Columns: client_data (1).csv



Distributions of Numeric Columns: price_data (1).csv





In []: