

# Power BI – Extract, Transform and Load

## ETL 1 – Examine Data with Power Query Editor

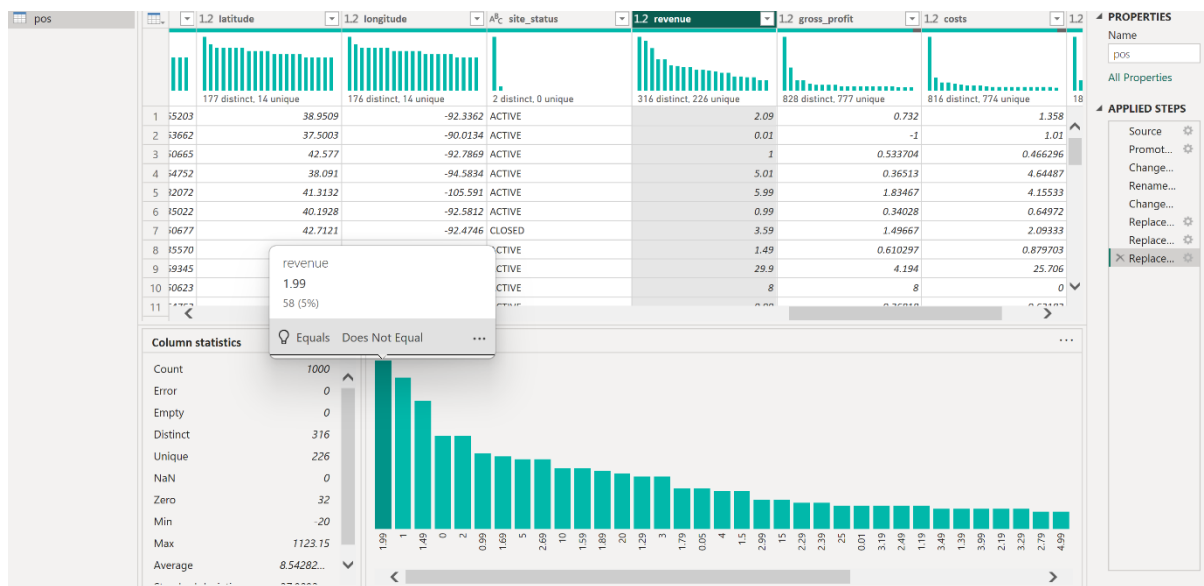
Load the data -> pos for (Point of Sale). The data is from a fictitious convenience store/gas station.

unique_id	transaction_id	date	customer_id	product_id	product_name	category_id	category_name	parent_id
2557308	201703042081111959504	28 May 2017	NA	65	Sort ascending	(271)	136	
1734017	201706079271112398802	31 August 2017	NA	53	Sort descending	onated Drinks	234	
2584067	201902181535523728812	14 May 2019	NA	65	Clear sort	(271)	136	
860898	20171019250314174331	12 January 2018	NA	150	Clear filter	Drinks	234	
80476	20180518184311204058	11 August 2018	NA	53	Clear all filters	onated Drinks	234	
2814039	20181105453334682996	29 January 2019	NA	53	Number filters	Disp Bev-retail (301)	232	
2562108	201810213883112441251	14 January 2019	NA	65	(Select all)	(271)	136	
1895841	2019042229262113381423	16 July 2019	NA	53	14	Bev-refills (314)	233	
2772081	2017062218401123276687	15 September 2017	NA	104	17	Bev-retail (311)	233	
2791806	2017020808722124025554	04 May 2017	NA	104	17	Bev-retail (311)	233	
1226183	20170718471112021691	11 October 2017	NA	8	30	is	281	
2585195	201805115611117489473	04 August 2018	NA	65	35	(271)	136	
2568235	201809248761123508588	18 December 2018	NA	65	205	(271)	136	
2683563	201801244421213611298	19 April 2018	NA	55	214	Bev-retail (311)	233	
2396312	201907303164112609535	23 October 2019	NA	8	215	Disp Bev-retail (301)	232	
2782494	201708093972233995982	02 November 2017	NA	104	216	Bev-retail (311)	233	
2561260	201702138761112597574	09 May 2017	NA	65	217	(271)	136	
1188081	201802084532114073184	04 May 2018	NA	53	218	Bev-refills (314)	233	
2804566	20190115293215061825	10 April 2019	NA	53	219	Disp Bev-retail (301)	232	
2870224	2018061840113214982867	11 September 2018	NA	56	223	Bev-retail (311)	233	
2815907	201711205141133834721	13 February 2018	NA	53	224	Disp Bev-retail (301)	232	
343759	20190608381214654354	01 September 2019	NA	106	225	fast Food	271	
2891060	201809075622124681034	01 December 2018	NA	56	229	Bev-retail (311)	233	
2708676	201707209051113098257	13 October 2017	NA	55	231	Bev-retail (311)	233	
2818485	201712135622124123275	08 March 2018	NA	53	240	Disp Bev-retail (301)	232	
2565254	2018101714472113739375	10 January 2019	NA	65	242	(271)	136	

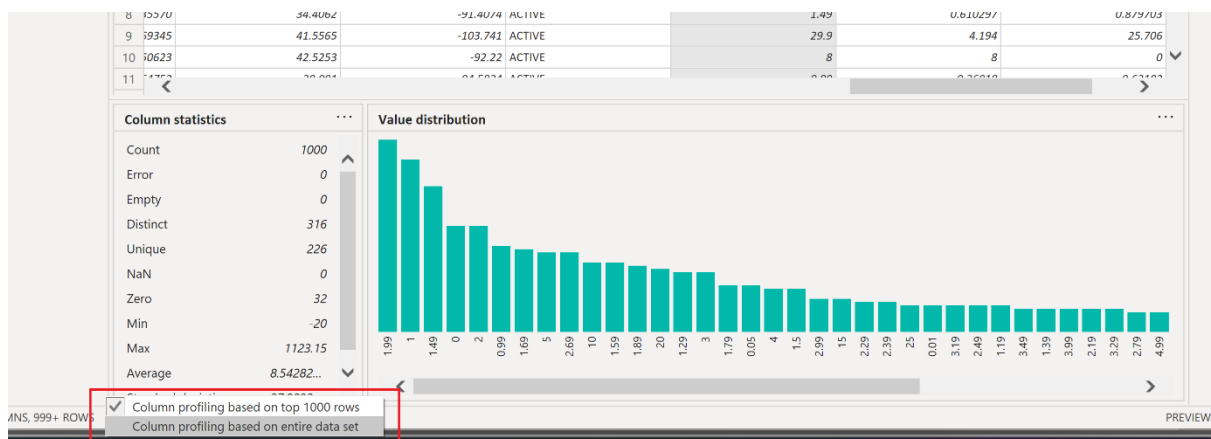
View a column distribution by clicking on a particular column, clicking on 'View' and selecting 'Column distribution'. Transformation steps are shown on right-hand side under 'Applied Steps'.

1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18
15203	38.9509	-92.3362	ACTIVE	1.2 revenue	2.09	0.732	1.358										
33662	37.5003	-90.0134	ACTIVE	0.01	1	0.533704	0.466296										
10665	42.577	-92.7869	ACTIVE	5.01	0.36513	0.36513	4.64487										
44752	38.091	-94.5834	ACTIVE	5.01	1.83467	1.83467	4.15533										
12072	41.3132	-105.591	ACTIVE	5.01	0.99	0.34028	0.64972										
15022	40.1928	-92.5812	ACTIVE	3.59	1.49667	1.49667	2.09333										
10677	42.7121	-92.4746	CLOSED	1.49	0.610297	0.610297	0.879703										
15570	34.4062	-91.4074	ACTIVE	29.9	4.194	4.194	25.706										
19345	41.5565	-103.741	ACTIVE	8	8	8	0										
10623	42.5253	-92.22	ACTIVE	0.99	0.36818	0.36818	0.62182										
14752	38.091	-94.5834	ACTIVE	1.99	0.84625	0.84625	1.14375										
10122	39.5957	-104.963	ACTIVE	84.01	8.1327	8.1327	75.8773										
15594	36.5326	-98.1534	ACTIVE	2	0.14	0.14	1.86										
11002	42.6745	-95.3054	ACTIVE	1	1	1	0										
10120	39.6124	-105.006	ACTIVE	6	0.42	0.42	5.58										
11301	43.1564	-95.1451	ACTIVE	13.1	1.90182	1.90182	11.19818										
10605	42.8351	-93.0034	ACTIVE	3.29	2.162	2.162	1.128										
15594	36.5326	-98.1534	ACTIVE														

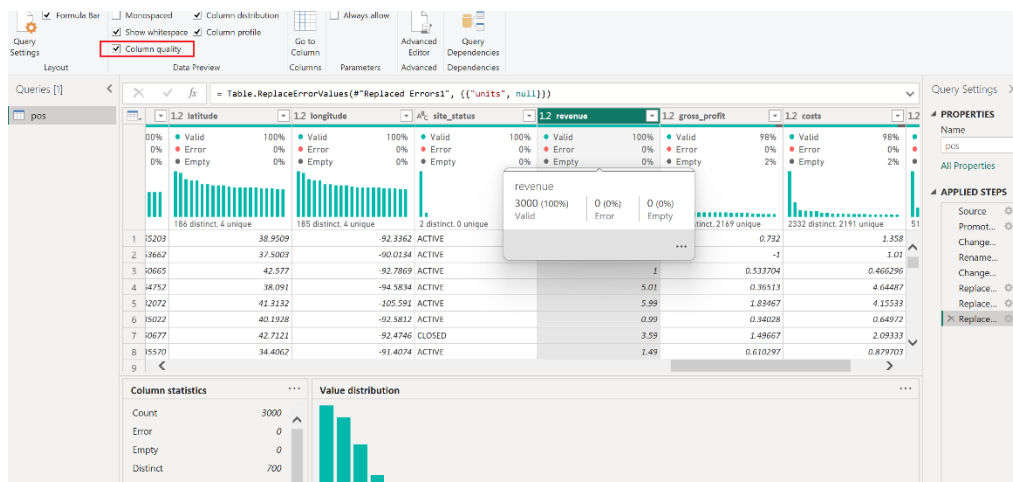
Clicking on 'View', then checking 'Column profile' gives the following view



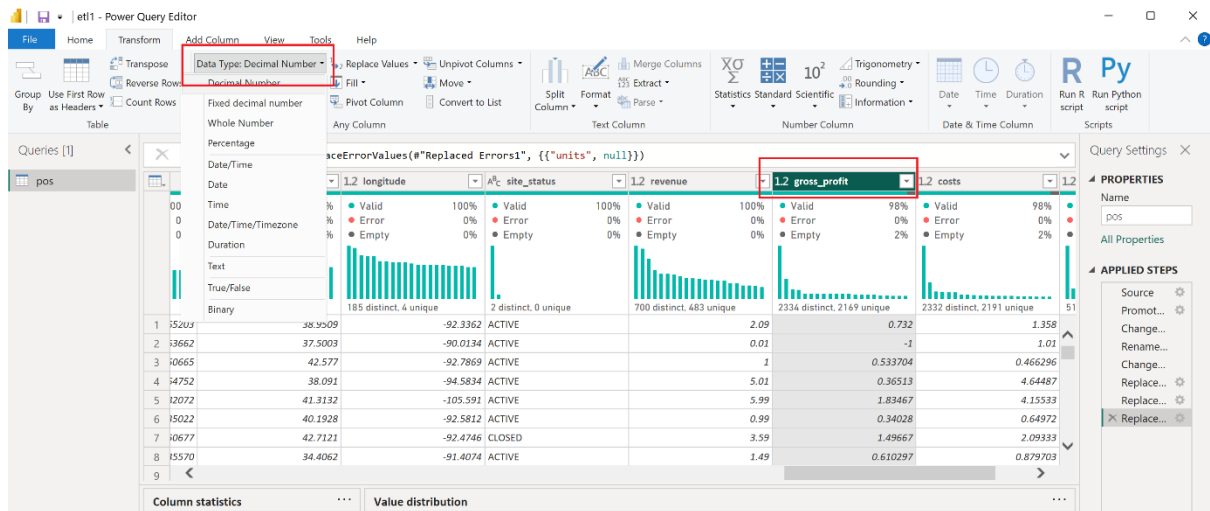
We can base the column profiling on the entire dataset



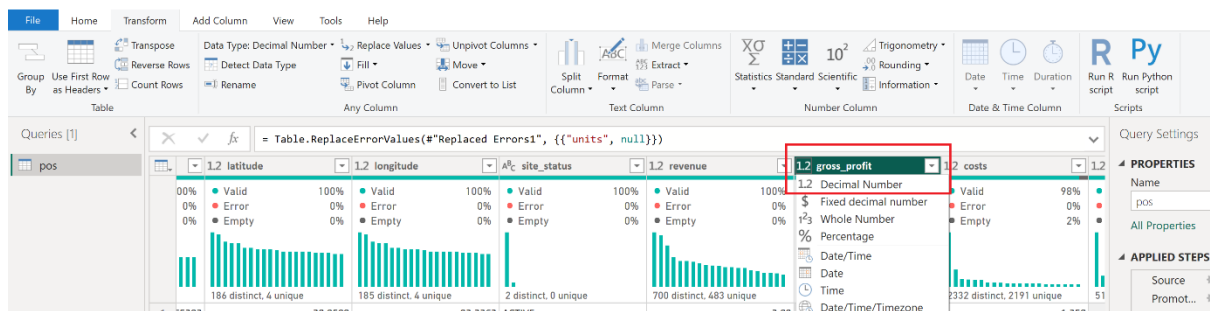
Now tick the 'Column quality' checkbox. This gives the percentage of Valid, Error and Empty entries



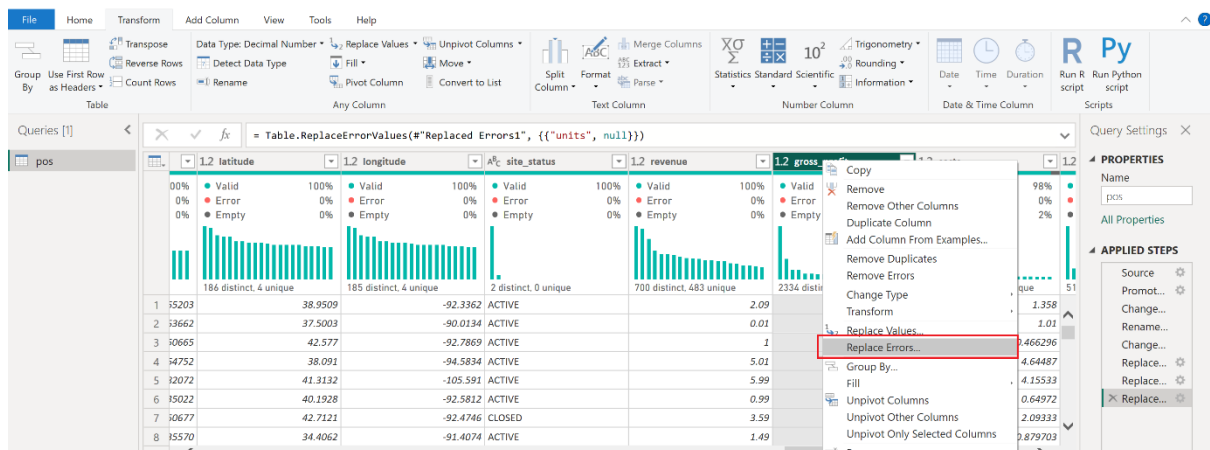
Change number columns to correct type using Data Type drop-down.



This can also be changed using the type symbol on the column header



Replace errors with null



## ETL 2 – Dates and Calculated Columns

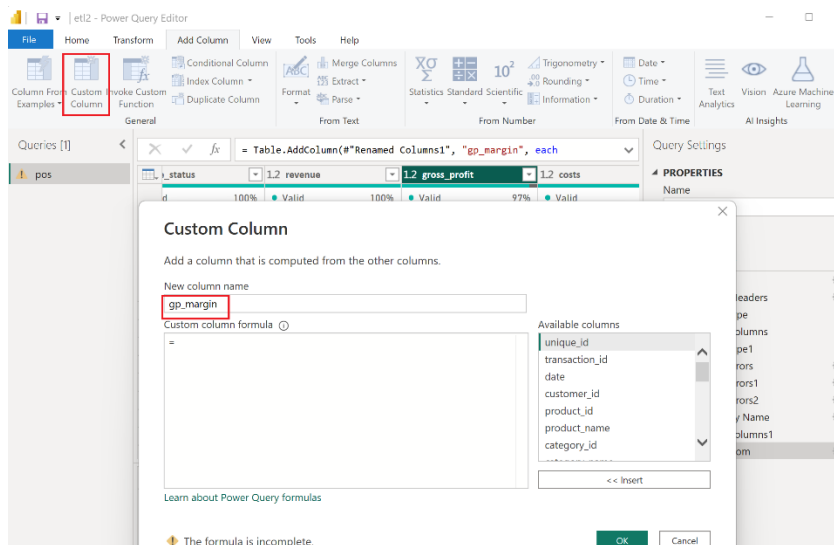
Add new column from selection

The screenshot shows the Power Query Editor interface. The 'Add Column' tab is active, and the 'From Selection' option is highlighted in the 'Column From Selection' dropdown. The formula bar displays the M code: `= Table.AddColumn(#"Renamed Columns1", "gp_margin", each`. The data preview shows columns: `unique_id`, `transaction_id`, `date`, and `customer`. Each column has a data type icon (A, B, C) and a status bar indicating validity (Valid, Error, Empty) and distinct/unique counts.

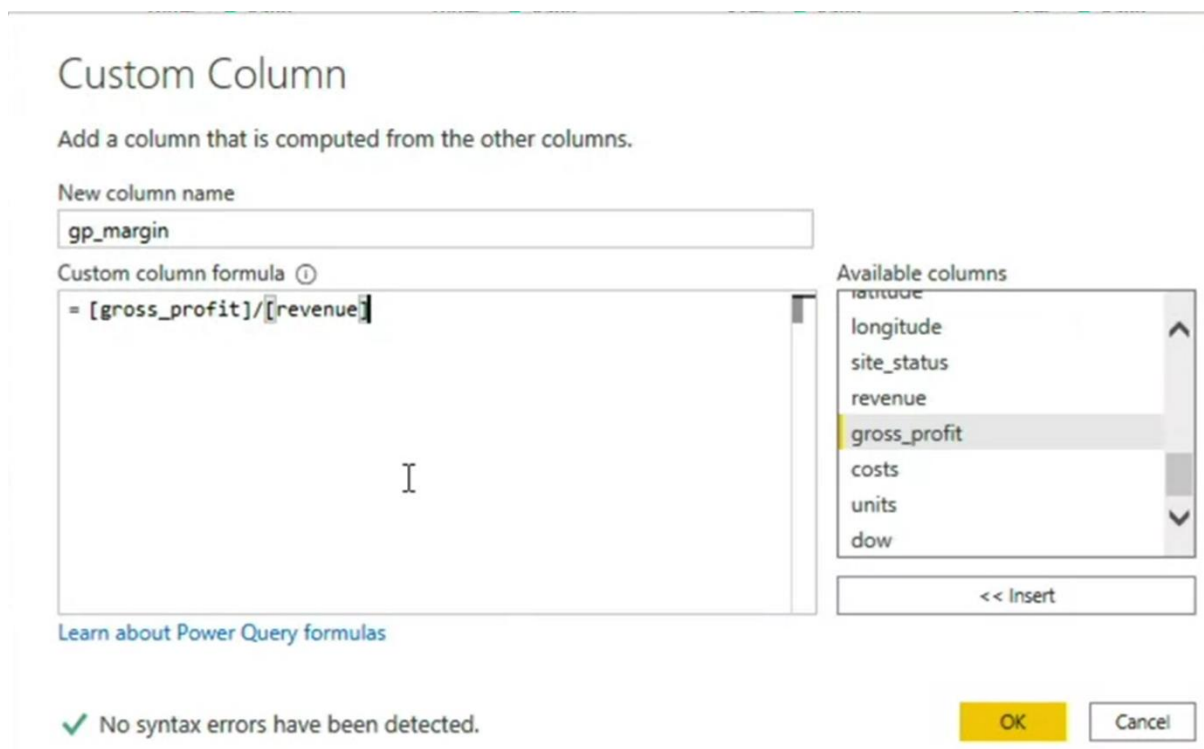
Add new column for Day of Week

The screenshot shows the Power Query Editor interface. The 'Add Column' tab is active, and the 'From Selection' option is highlighted in the 'Column From Selection' dropdown. The formula bar displays the M code: `= Table.AddColumn(#"Renamed Columns1", "gp_margin", each`. The data preview shows columns: `1.2 units`, `dow`, and `gp_margin`. The `dow` column is highlighted with a red box, showing a data type icon (A, B, C) and a status bar indicating validity (Valid, Error, Empty) and distinct/unique counts.

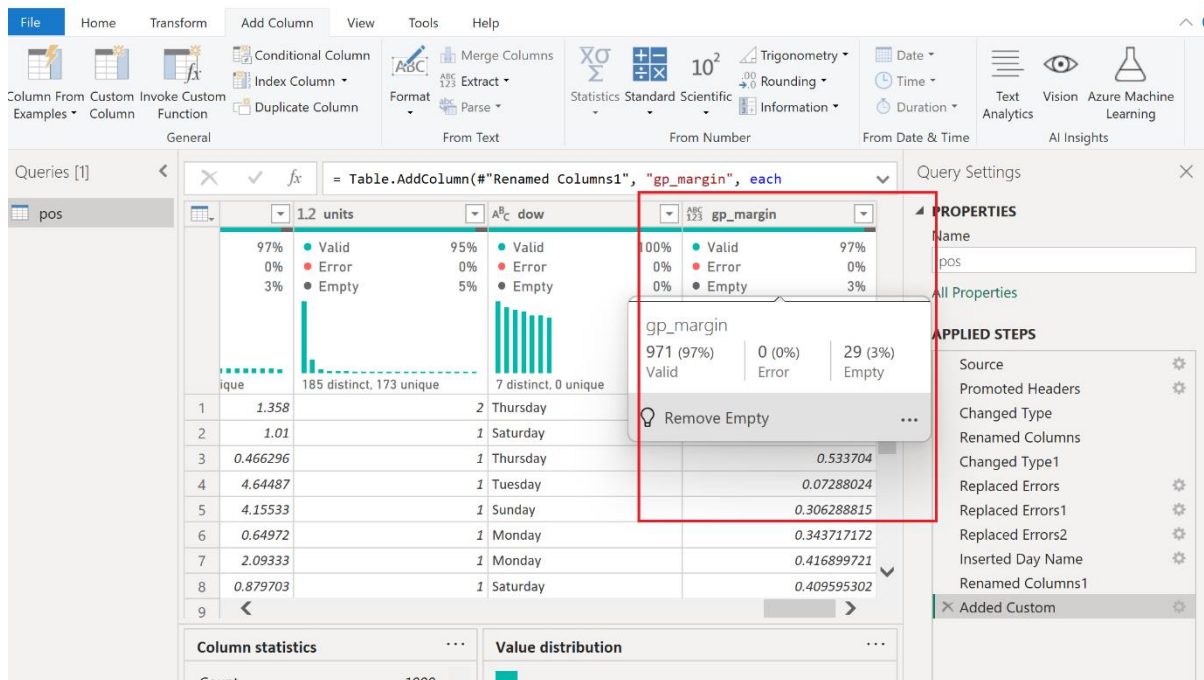
## Add new Custom Column for gross profit margin



## Use custom formula to calculate value for column



## New column added



## ETL 3 – Checking for and Eliminating Outliers

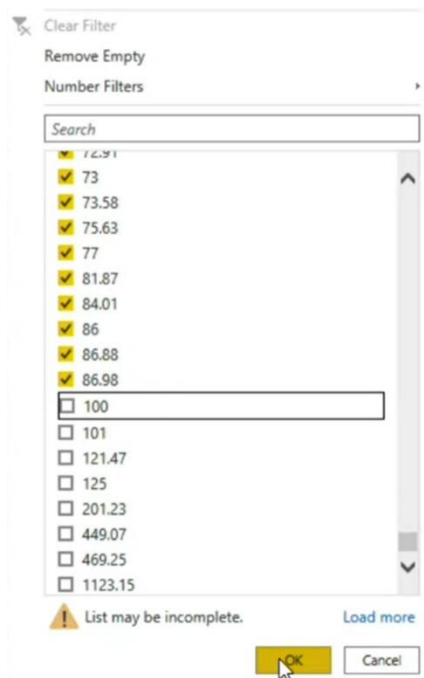
The highlighted rows are 3 standard deviations more than the mean (>87), therefore outliers.

id	site_status	revenue
-94.8933	ACTIVE	1123.15
-95.1451	ACTIVE	469.25
-102.58	ACTIVE	449.07
-93.1276	ACTIVE	201.23
-92.4672	ACTIVE	125
-93.9646	ACTIVE	121.47
-96.124	ACTIVE	101
-98.3491	ACTIVE	100
-104.818	ACTIVE	86.98
-104.998	ACTIVE	86.88
-96.124	ACTIVE	86
-98.1534	ACTIVE	84.01

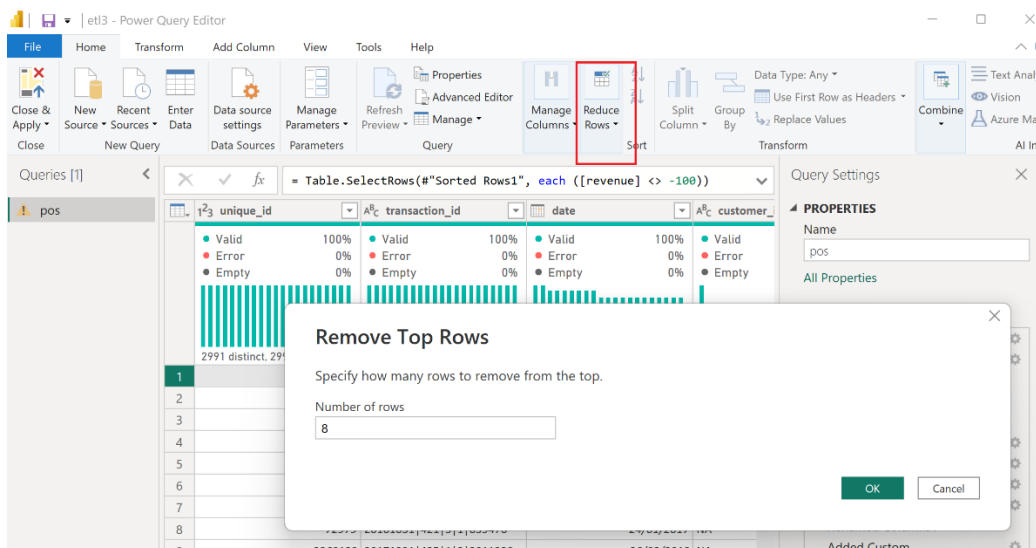
These outliers are related to fuel

DIESELBIO-T	Fuel Premiim	Fuel
DIESELBIO	Fuel Premiim	Fuel
DIESEL	Fuel Premiim	Fuel
DIESELBIO	Fuel Premiim	Fuel
DIESELBIO	Fuel Premiim	Fuel
MARLB SPCL SLCT RED 100 BX	Cig-off Inv-premium (441)	Off Invoice Cigs (104)
DIESELBIO	Fuel Premiim	Fuel
DIESELBIO	Fuel Premiim	Fuel

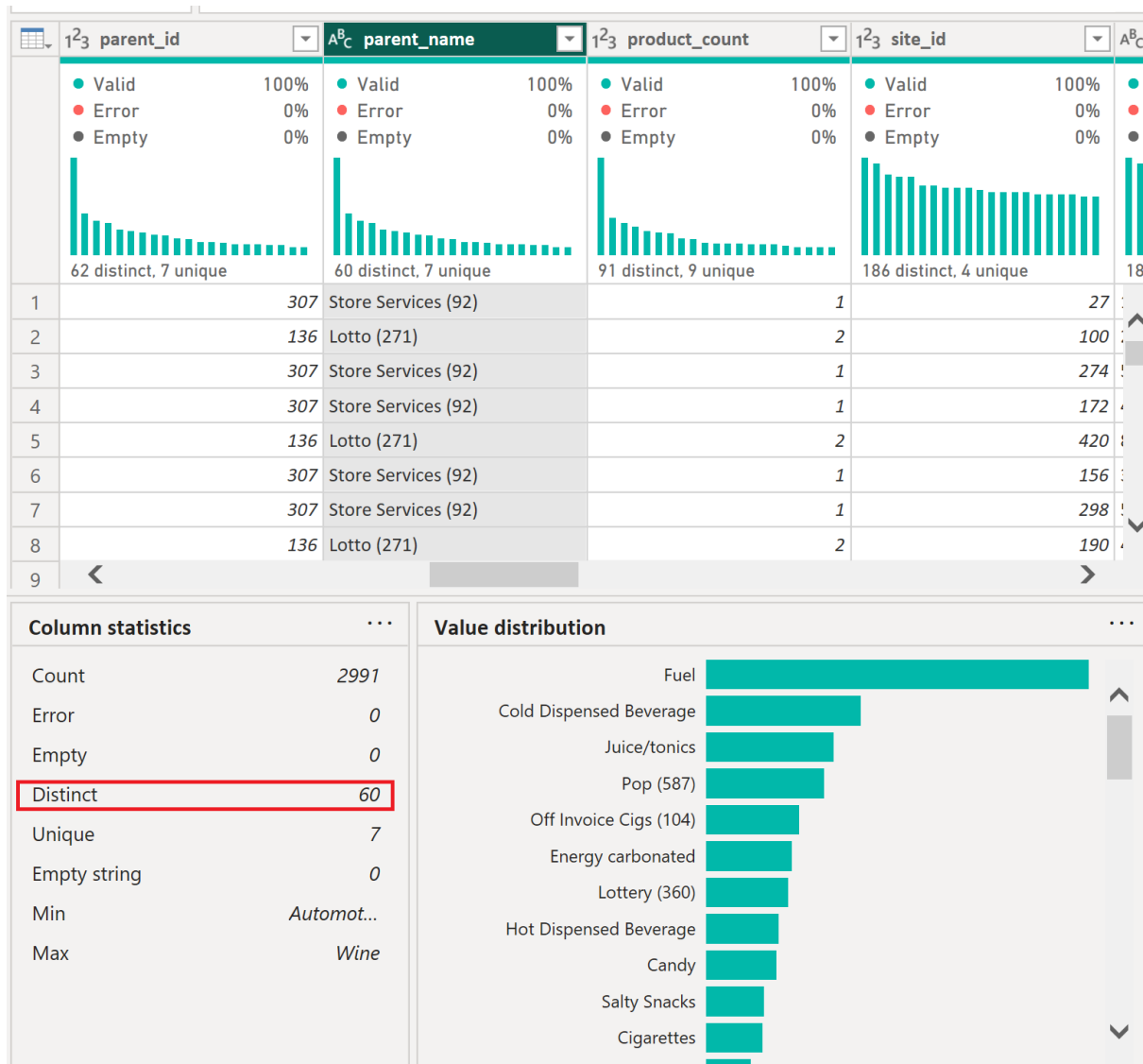
We can remove these outliers by manually deselecting them from the column



Another way to do this is to go to 'Reduce Rows' in the Home tool and select 'Remove Top Rows' and specify the number of rows to remove. Note this will only work if you've sorted the data correctly.



The column parent\_name has high cardinality, meaning that there are 60 distinct types.



We can create a new column to group some of these items

The image shows the 'Groups' dialog box in Tableau. The 'Name' field is set to 'super\_parent\_name' and the 'Field' is 'parent\_name'. The 'Group type' is set to 'List'.

**Ungrouped values:**

- Eggs/cheese
- Energy carbonated
- Energy non-carb
- Food-
- Food-Grocery
- Frozen Dispensed Beverage
- Frozen Foods
- Fuel
- Gift Cards - SKU (938)
- Gms/novelty

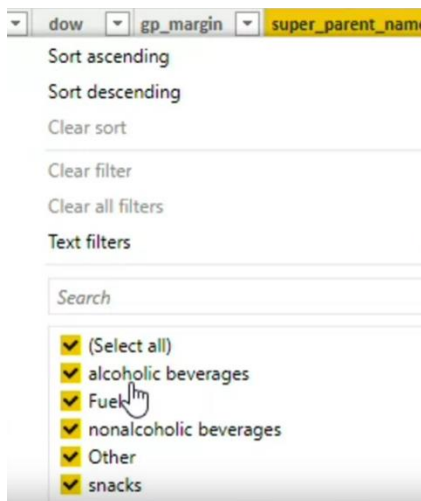
**Groups and members:**

- alcoholic beverages
  - Beer
  - Liquor
  - Wine

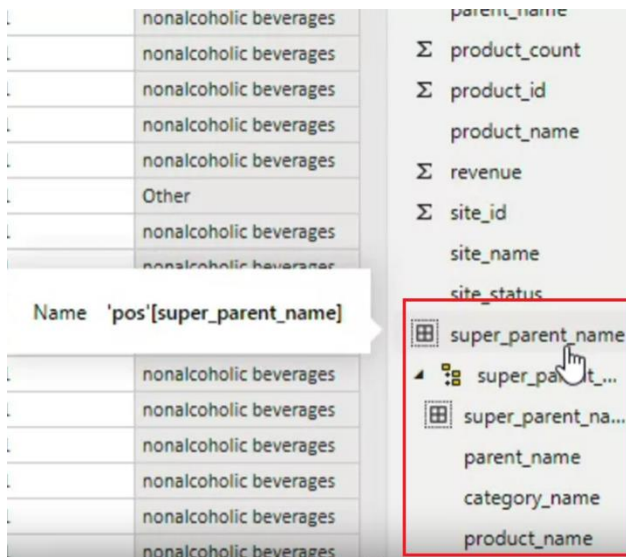
Buttons: Group, Ungroup, Include Other group (checkbox), OK, Cancel.



Now we have just 5 different items



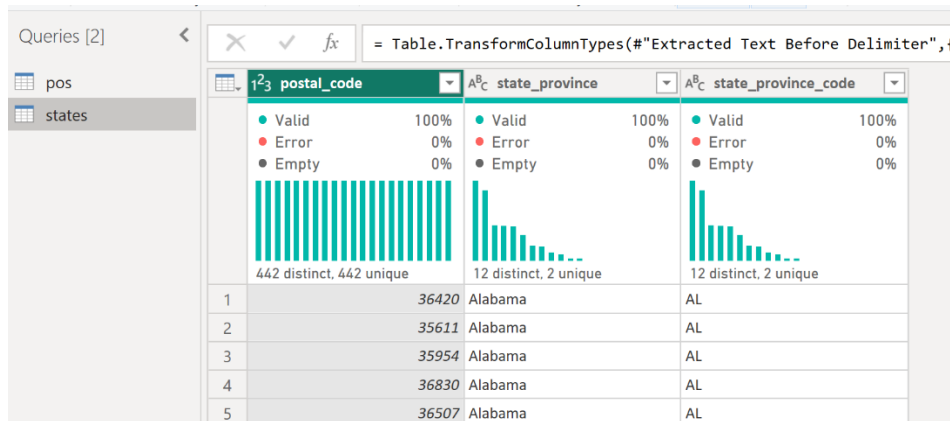
We can create a new hierarchy for our new column



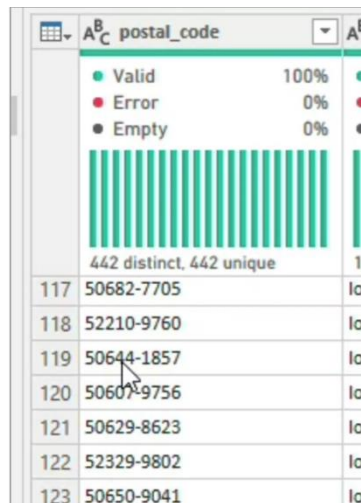
## ETL 4 : Data Models and Joins

Load up dataset `states.csv` that contains a list of states and join with the `pos.csv` dataset.

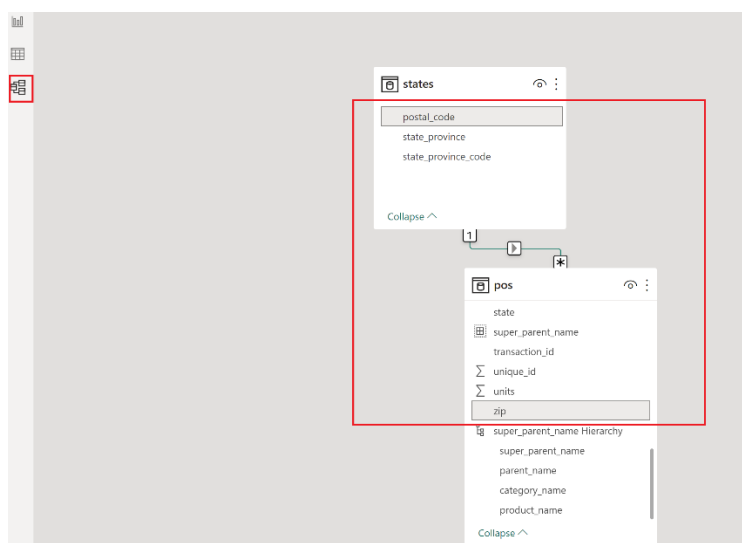
Note that the `states` dataset does not have a zip code column. It is instead called `postal_code`



However, some of the values contain a 9-digit `postal_code`



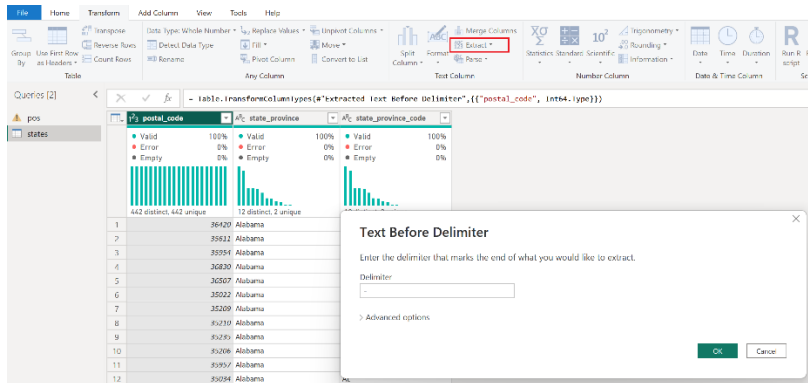
Go to the Model View and create a connection between the two datasets



We can see that there is a One-To-Many relationship between the states and pos table.

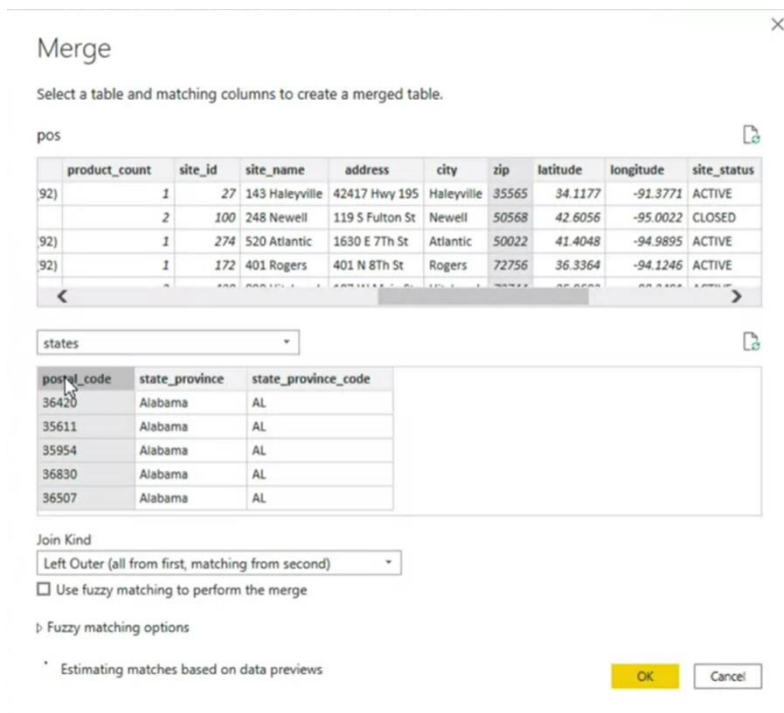
Each transaction in pos matches up with only one observation in the states table. But one entry in the states table could match up with many entries in the pos table.

To remove the last 4 digits of the post codes with 9 digits. Highlight the column, click on 'Transform', then 'Extract', 'Text Before Delimiter' and specify – (hyphen)

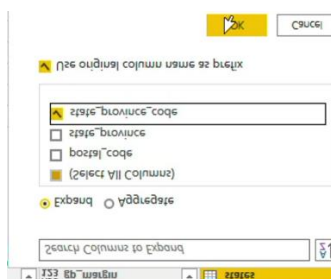


Use Merge Queries to link the zip from the pos dataset to the postal\_code of the states dataset.

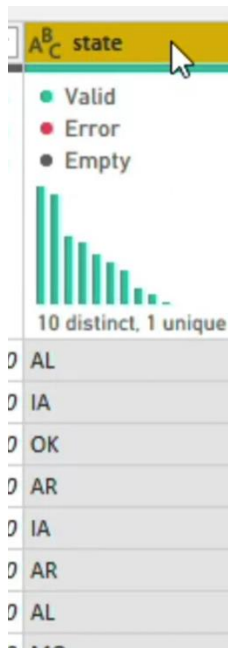
Note that the columns must be the same type.



We only want to keep the state\_province\_code column



The column name is a bit long, so rename it to state



Use star schema, one main dataset, and any other dataset is joined to that key dataset to avoid querying lots of hierarchical relationships.