

Build a Data Pipeline for Processing and Storing Sales Data

THE SITUATION

Created a data pipeline to process and store sales data from a CSV file. The pipeline will transform and load the data with Python into a PostgreSQL and MySQL database for further analysis using Power BI.

Tools required: Python (Pandas), PostgreSQL, MySQL, DBdiagram.io and Power BI

THE STEPS

Used DBdiagram.io:

- To design a dimensional model for MavenMarket Database.

Used MySQL and PostgreSQL to:

- Create a database title MavenMarket.
- Define the database schema under the database (MavenMarket):
Product,
region,
store,
return product.

Used Python to:

- Use pandas library to load the sales data from the CSV file into pandas dataframe.
- Use pandas to clean and transform the data as needed. For instance, drop some columns that are not needed in product and store table.
- Use the psycopg2 and mysql library to connect to the PostgreSQL and MySQL database.
- Use pandas, psycopg2, and mysql to load the cleaned and transformed data into the database.
- Runed some test queries on the data in the database to ensure the data pipeline is working correctly.

Used PowerBI to:

- To connect the raw data
- Build a relational data model
- Create new calculated columns and DAX measures
- Design an interactive report to analyze and visualize the data

THE SUMMARY

After performing data wrangling, data modelling and data visualization. Here, are the following insights from the dataset.

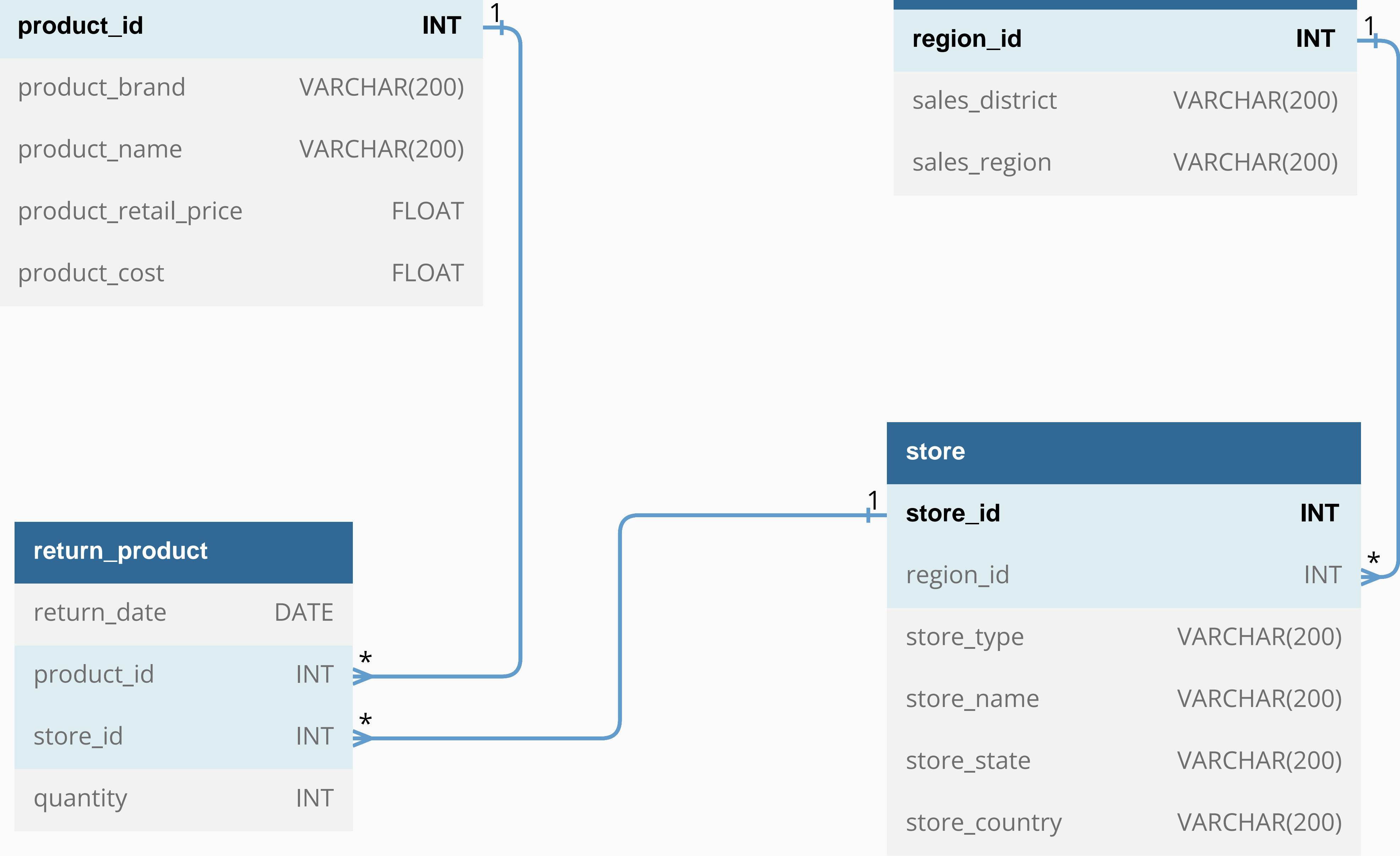
- Portland reached 1,000 sales in December to close out the year
- High Top product returns doubled in Mexico (4 to 8), at a return rate of 1.2%
- Plato products drove the strongest overall profit margin (63.55%) in 1998.

product	
product_id	INT
product_brand	VARCHAR(200)
product_name	VARCHAR(200)
product_retail_price	FLOAT
product_cost	FLOAT

region	
region_id	INT
sales_district	VARCHAR(200)
sales_region	VARCHAR(200)

return_product	
return_date	DATE
product_id	INT
store_id	INT
quantity	INT

store	
store_id	INT
region_id	INT
store_type	VARCHAR(200)
store_name	VARCHAR(200)
store_state	VARCHAR(200)
store_country	VARCHAR(200)



SQL SCRIPTS FOR MAVENMARKET

```
CREATE TABLE product (  
    product_id INT PRIMARY KEY,  
    product_brand VARCHAR(200),  
    product_name VARCHAR(200),  
    product_retail_price FLOAT,  
    product_cost FLOAT  
);
```

```
CREATE TABLE region (  
    region_id INT PRIMARY KEY,  
    sales_district VARCHAR(200),  
    sales_region VARCHAR(200)  
);
```

```
CREATE TABLE return_product (  
    return_date DATE,  
    product_id INT,  
    store_id INT,  
    quantity INT,  
    FOREIGN KEY (product_id)  
        REFERENCES product (product_id),  
    FOREIGN KEY (store_id)  
        REFERENCES store (store_id)  
);
```

```
CREATE TABLE store (  
    store_id INT PRIMARY KEY,  
    region_id INT,  
    store_type VARCHAR(200),  
    store_name VARCHAR(200),  
    store_state VARCHAR(200),  
    store_country VARCHAR(200),  
    FOREIGN KEY (region_id)  
        REFERENCES region (region_id );
```

Read Dataset

```
In [1]: import pandas as pd
```

```
In [2]: # Read Products data
MavenMarket_Products = pd.read_csv(r'C:\Users\Lenovo\Desktop\Personal Documents\DataSet\Maven+Market+CSV+Files\MavenMarket_Products.csv', delimiter = ",")
MavenMarket_Products.head(5)
```

Out[2]:

	product_id	product_brand	product_name	product_sku	product_retail_price	product_cost	product_weight	recyclable	low_fat
0	1	Washington	Washington Berry Juice	90748583674	2.85	0.94	8.39	NaN	NaN
1	2	Washington	Washington Mango Drink	96516502499	0.74	0.26	7.42	NaN	1.0
2	3	Washington	Washington Strawberry Drink	58427771925	0.83	0.40	13.10	1.0	1.0
3	4	Washington	Washington Cream Soda	64412155747	3.64	1.64	10.60	1.0	NaN
4	5	Washington	Washington Diet Soda	85561191439	2.19	0.77	6.66	1.0	NaN

```
In [3]: # Read Regions data
Regions = pd.read_csv(r'C:\Users\Lenovo\Desktop\Personal Documents\DataSet\Maven+Market+CSV+Files\MavenMarket_Regions.csv', delimiter = ",")
Regions.head(5)
```

Out[3]:

	region_id	sales_district	sales_region
0	1	San Francisco	Central West
1	2	Mexico City	Mexico Central
2	3	Los Angeles	South West
3	4	Guadalajara	Mexico West
4	5	Vancouver	Canada West

```
In [4]: # Read MavenMarket_Returns_1997-1998 data
Returns = pd.read_csv(r'C:\Users\Lenovo\Desktop\Personal Documents\DataSet\Maven+Market+CSV+Files\MavenMarket_Returns_1997-1998.csv', delimiter = ",", parse_dates = ["return_date"])
Returns.head(5)
```

Out[4]:

	return_date	product_id	store_id	quantity
0	1997-01-01	250	6	1
1	1997-01-01	628	6	1
2	1997-01-01	869	6	1
3	1997-01-02	469	11	1
4	1997-01-02	532	23	2

```
In [5]: # Read MavenMarket_Stores_1997-1998 data
MavenMarket_Stores = pd.read_csv(r'C:\Users\Lenovo\Desktop\Personal Documents\DataSet\Maven+Market+CSV+Files\MavenMarket_Stores.csv', delimiter = ",")
MavenMarket_Stores.head(5)
```

Out[5]:

	store_id	region_id	store_type	store_name	store_street_address	store_city	store_state	store_country	store_phone	first_opened_date	last_remodel_date	total_sqft	grocery_sqft
0	1	28	Supermarket	Store 1	2853 Bailey Rd	Acapulco	Guerrero	Mexico	262-555-5124	1/9/1982	12/5/1990	23593	17475
1	2	78	Small Grocery	Store 2	5203 Catanzaro Way	Bellingham	WA	USA	605-555-8203	4/2/1970	6/4/1973	28206	22271
2	3	76	Supermarket	Store 3	1501 Ramsey Circle	Bremerton	WA	USA	509-555-1596	6/14/1959	11/19/1967	39696	24390
3	4	27	Gourmet Supermarket	Store 4	433 St George Dr	Camacho	Zacatecas	Mexico	304-555-1474	9/27/1994	12/1/1995	23759	16844
4	5	4	Small Grocery	Store 5	1250 Coggins Drive	Guadalajara	Jalisco	Mexico	801-555-4324	9/18/1978	6/29/1991	24597	15012

Data Wrangling

```
In [6]: #drop some columns
Product = MavenMarket_Products.drop(["product_sku", "product_weight", "recyclable", "low_fat"], axis = 1)
Product.head(5)
```

Out[6]:

	product_id	product_brand	product_name	product_retail_price	product_cost
0	1	Washington	Washington Berry Juice	2.85	0.94
1	2	Washington	Washington Mango Drink	0.74	0.26
2	3	Washington	Washington Strawberry Drink	0.83	0.40
3	4	Washington	Washington Cream Soda	3.64	1.64
4	5	Washington	Washington Diet Soda	2.19	0.77

```
In [7]: Store = MavenMarket_Stores.drop(["store_phone", "first_opened_date", "last_remodel_date", "total_sqft", "grocery_sqft", "store_street_address", "store_city"], axis = 1)
Store.head(5)
```

Out[7]:

	store_id	region_id	store_type	store_name	store_state	store_country
0	1	28	Supermarket	Store 1	Guerrero	Mexico
1	2	78	Small Grocery	Store 2	WA	USA
2	3	76	Supermarket	Store 3	WA	USA
3	4	27	Gourmet Supermarket	Store 4	Zacatecas	Mexico
4	5	4	Small Grocery	Store 5	Jalisco	Mexico

Load Data in MySQL

```
In [8]: pip install sqlalchemy

Requirement already satisfied: sqlalchemy in c:\users\lenovo\anaconda3\lib\site-packages (1.4.32)
Requirement already satisfied: greenlet!=0.4.17 in c:\users\lenovo\anaconda3\lib\site-packages (from sqlalchemy) (1.1.1)
Note: you may need to restart the kernel to use updated packages.
```

```
In [9]: import sqlalchemy
```

```
In [10]: pip install mysql

Requirement already satisfied: mysql in c:\users\lenovo\anaconda3\lib\site-packages (0.0.3)
Requirement already satisfied: mysqlclient in c:\users\lenovo\anaconda3\lib\site-packages (from mysql) (2.1.1)
Note: you may need to restart the kernel to use updated packages.
```

```
In [11]: engine1 = sqlalchemy.create_engine("mysql://sammy:password@localhost/mavenmarket")
```

```
In [20]: Product.to_sql(name = 'product', con = engine1, index = False, if_exists = "replace")
```

Out[20]: 1560

```
In [19]: Regions.to_sql(name = 'region', con = engine1, index = False, if_exists = 'replace')
```

Out[19]: 109

```
In [18]: Store.to_sql(name = 'store', con = engine1, index = False, if_exists = 'replace')
```

Out[18]: 24

```
In [17]: Returns.to_sql(name = 'return_product', con = engine1, index = False, if_exists = 'replace')
```

Out[17]: 7087

Load Data in PostgreSQL

```
In [21]: engine = sqlalchemy.create_engine("postgresql+psycopg2://Uche:diamond@localhost/MavenMarket")
```

```
In [26]: Product.to_sql(name = 'product', con = engine, index = False, if_exists = 'replace')
```

Out[26]: 560

```
In [25]: Regions.to_sql(name = 'region', con = engine, index = False, if_exists = 'replace')
```

Out[25]: 109

```
In [24]: Store.to_sql(name = 'store', con = engine, index = False, if_exists = 'replace')
```

Out[24]: 24

```
In [23]: Returns.to_sql(name = 'return_product', con = engine, index = False, if_exists = 'replace')
```

Out[23]: 87

In []:

pgAdmin 4

File Object Tools Help

Browser

Servers (1)

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product

region

return_product

Dashboard Properties SQL Statistics Dependencies Dependents MavenMarket/postgres@PostgreSQL 15*

MavenMarket/postgres@PostgreSQL 15

No limit

Query Query History

```
1 select * from region
```

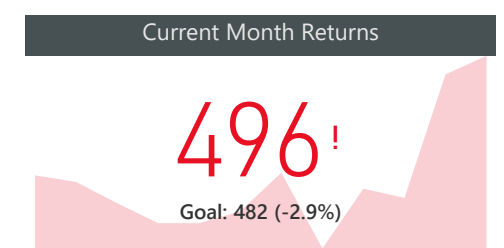
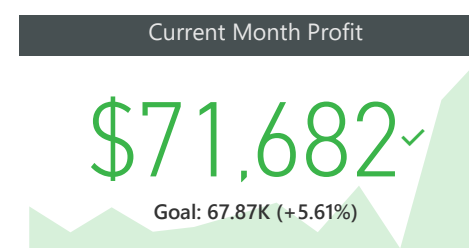
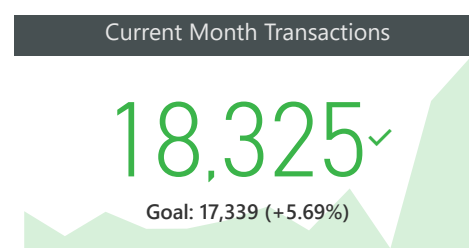
Data Output Messages Notifications

	region_id bigint	sales_district text	sales_region text
1	1	San Francisco	Central West
2	2	Mexico City	Mexico Central
3	3	Los Angeles	South West
4	4	Guadalajara	Mexico West
5	5	Vancouver	Canada West
6	6	Victoria	Canada West
7	7	San Diego	South West
8	8	San Diego	South West
9	9	San Diego	South West
10	10	San Diego	South West

Total rows: 109 of 109 Query complete 00:00:00.075

Successfully run. Total query runtime: 75 msec. 109 rows affected.





Product Brand	Total Transactions	Total Profit	Profit Margin	Return Rate
Hermanos	5,342	\$21,753	58.64%	0.95%
Ebony	5,238	\$20,354	59.81%	0.96%
Tell Tale	5,112	\$19,982	58.05%	0.99%
Tri-State	5,099	\$19,980	58.91%	1.10%
High Top	4,940	\$19,810	60.42%	1.01%
Nationeel	4,408	\$18,617	60.44%	1.18%
Best Choice	4,218	\$18,355	60.64%	0.81%
Horatio	4,195	\$17,737	58.42%	1.26%
Fort West	4,108	\$15,834	59.80%	0.97%
Fast	4,097	\$16,469	61.03%	1.07%
Sunset	3,953	\$14,018	60.45%	1.03%
Carrington	3,891	\$14,883	59.52%	0.78%
Red Wing	3,870	\$15,870	59.36%	1.06%
Big Time	3,816	\$15,560	60.20%	1.05%
Cormorant	3,744	\$15,749	61.60%	0.87%
Imagine	3,634	\$15,102	61.40%	1.06%
Super	3,618	\$13,868	60.59%	0.96%
Denny	3,584	\$16,015	58.02%	0.99%
High Quality	3,577	\$16,139	59.98%	1.13%
Golden	3,550	\$13,256	58.72%	0.88%
BBB Best	3,514	\$12,991	62.12%	0.80%
PigTail	3,467	\$11,617	60.68%	1.04%
Plato	3,352	\$12,748	63.55%	1.06%
Landslide	3,270	\$10,647	58.65%	0.98%
CDR	3,078	\$12,062	58.98%	1.11%
Better	2,823	\$9,179	61.15%	1.07%
Carlson	2,564	\$10,534	61.20%	0.97%
Pleasant	2,564	\$10,187	60.18%	0.92%
Just Right	2,558	\$9,283	59.54%	0.83%
Bravo	2,484	\$11,027	59.15%	0.82%
Total	113,668	\$449,627	59.94%	1.00%

