

Definition of Terms

Term	Definition
SOW	Statement of Work; a contracted agreement of work to be performed by Contractor for Client.
MVP	Minimum Viable Product; a working baseline deliverable intended to provide necessary functionality to be performant of the requirements outlined within the SOW, also carrying the possibility to inspire future work or enhancement of the current work in later SOWs.
Topline Metrics	The top KPIs being monitored by Client in Looker dashboards. These KPIs consist of Net Sales, Total Orders, AOV, New Customer acquisition and CPA performance.

Scope of MVP

The overall scope of this MVP was to provide visibility into the company's topline metrics over different snapshots of time and at different levels of detail to drive actionable decisions. Moreover, this MVP aimed at providing a New Product Launch template dashboard to allow visibility into new product performance upon launch until maturity. Finally, Contractor agreed to provide a rendering of the Finance Cohort Exports within Looker. The dashboards that housed most of these deliverables are within the Company Report and the New Product Launch dashboards. The Company Report dashboard is a multi-tabbed dashboard designed to give visibility in increasing detail into the topline metrics.

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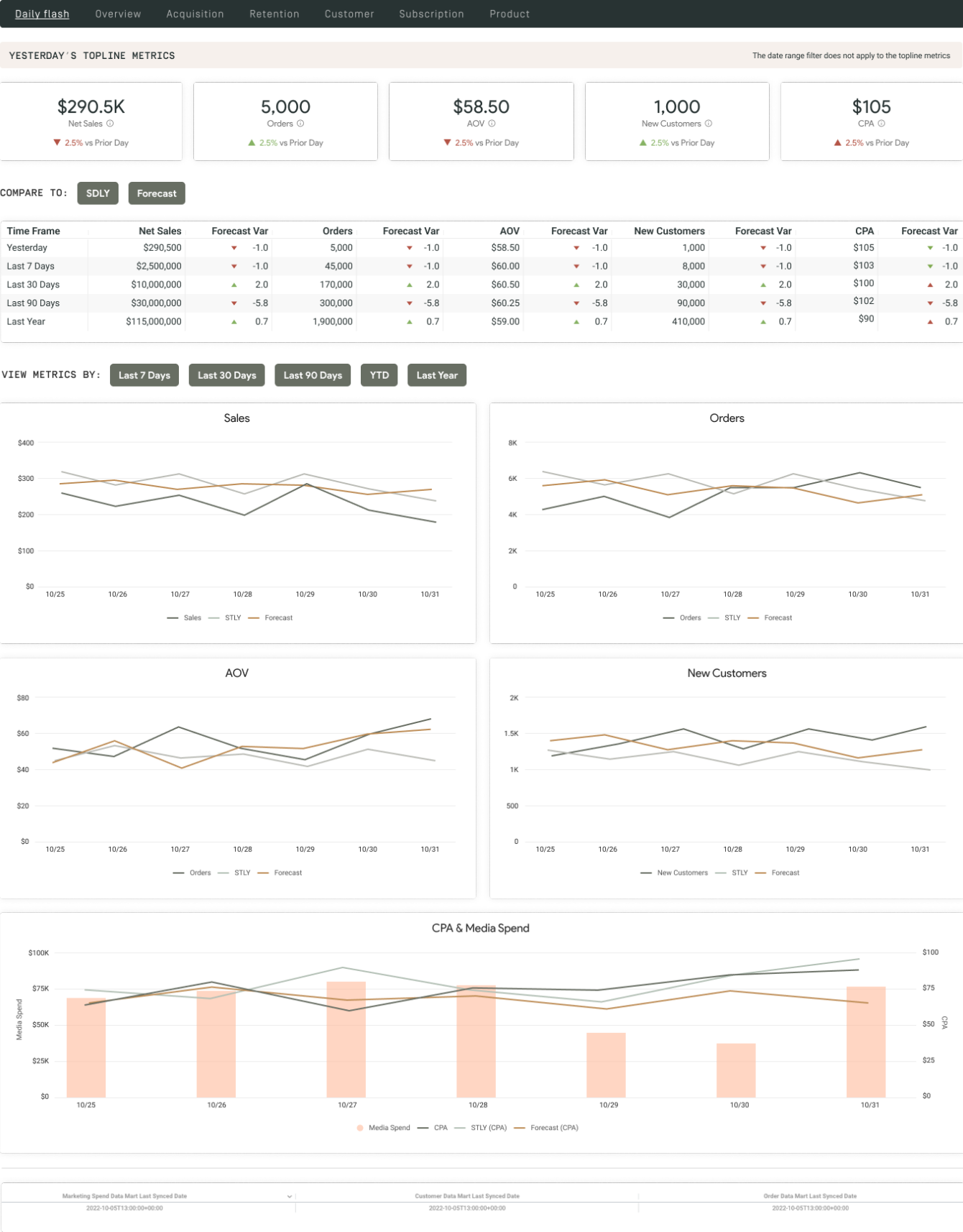
Plan for this SOW

During this SOW, Contractor agreed to deliver the two beginning tabs of the Company Report dashboard (Daily Flash & Overview), the New Product Launch dashboard and the Finance Cohort Export. Below are the wireframes for the dashboards, as well as the previous embodiment of the Finance Cohort Export:

High Fidelity Designs

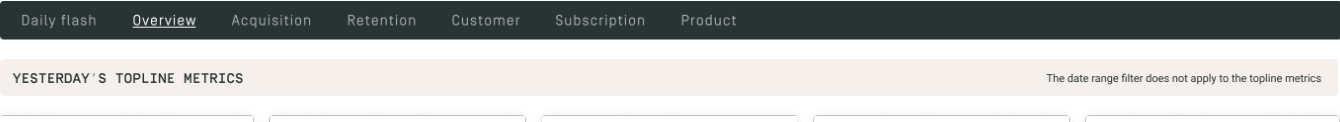
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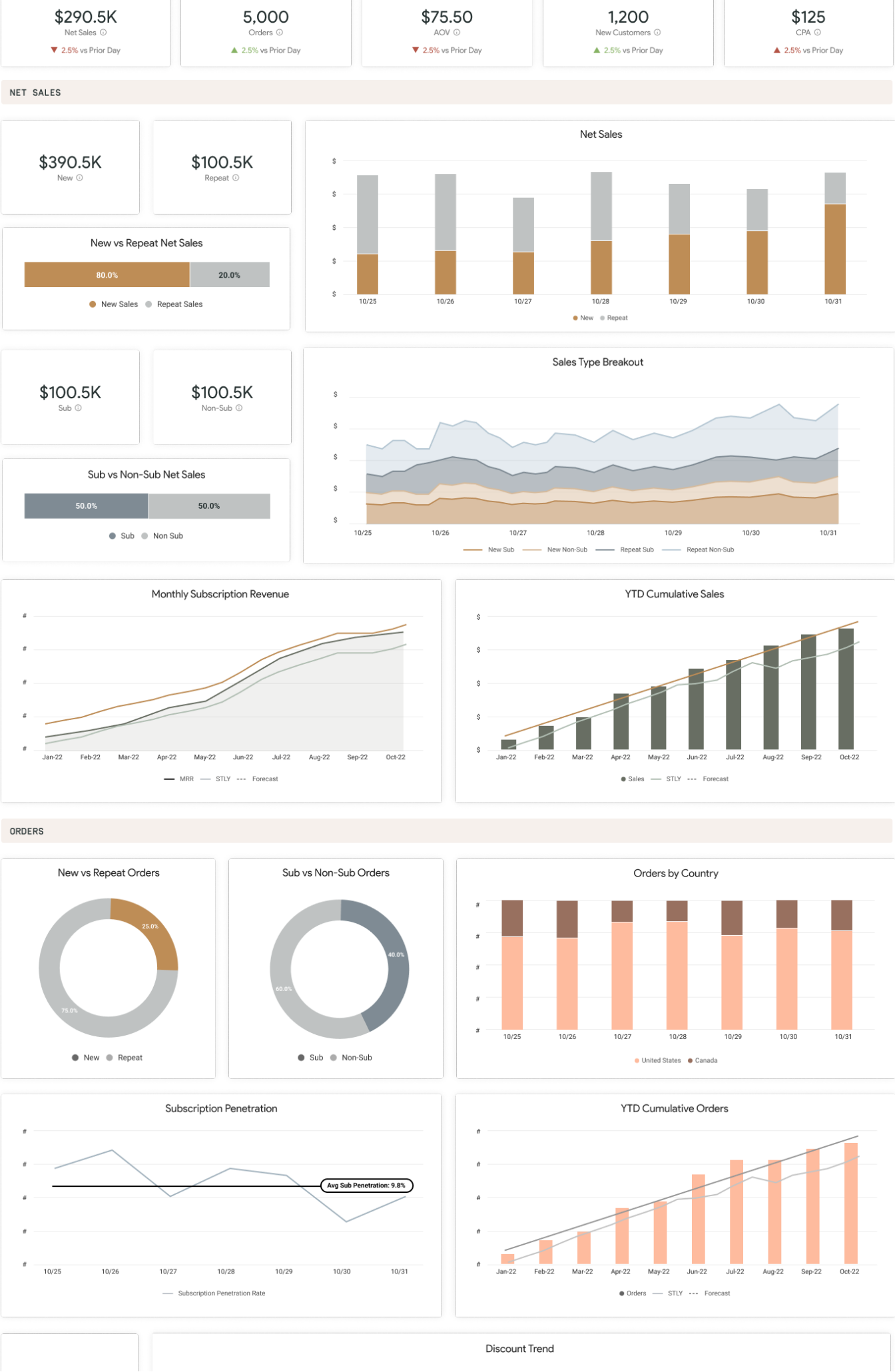
Daily Flash Tab High Fidelity Design

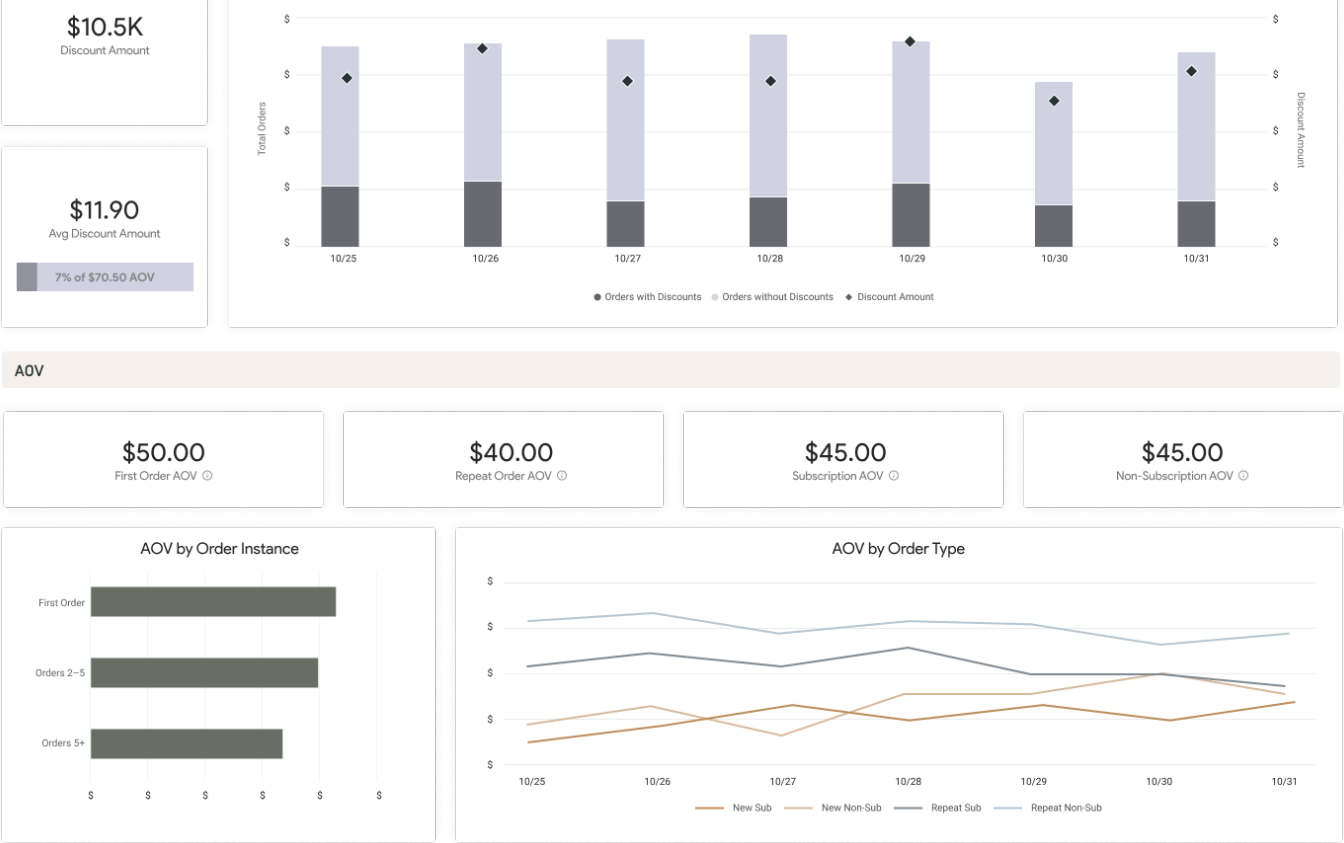


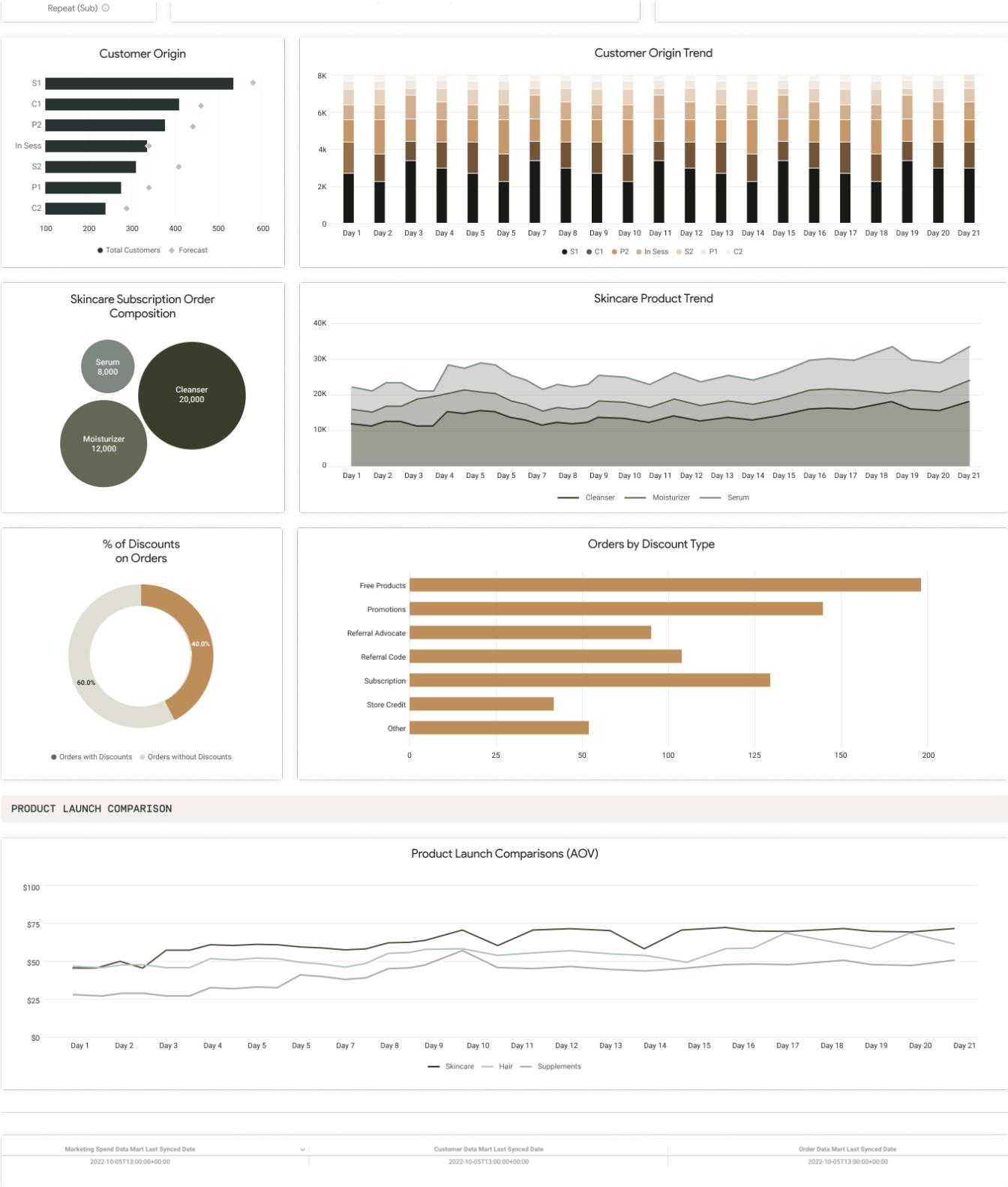
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Overview Tab High Fidelity Design









finance cohort

File Edit View Insert Format Data Tools Extensions Help

Last edit was made 7 days ago by Katharine Nicosia

100% View only

A1	A	B	C	D	E	F	G	H	I	J	K	L	M	N	O	P	Q
1	last_order_coun	cohort_level	metric	cohort_finance	cohort_name	order_placed_at	DISCOUNT										
2	CA	cohort	total_discount	new_non_sub		7/1/2018	0										
3	CA	cohort	total_discount	new_non_sub		12/1/2018	0										
4	CA	cohort	total_discount	new_non_sub		1/1/2019	0										
5	CA	cohort	total_discount	new_non_sub		2/1/2019	0										
6	CA	cohort	total_discount	new_non_sub		4/1/2019	0										
7	CA	cohort	total_discount	new_non_sub		5/1/2019	0										
8	CA	cohort	total_discount	new_non_sub		7/1/2019	10										
9	CA	cohort	total_discount	new_non_sub		8/1/2019	0										
10	CA	cohort	total_discount	new_non_sub		10/1/2019	0										
11	CA	cohort	total_discount	new_non_sub		11/1/2019	0										
12	CA	cohort	total_discount	new_non_sub		12/1/2019	0										
13	CA	cohort	total_discount	new_non_sub		2/1/2020	0										
14	CA	cohort	total_discount	new_non_sub		6/1/2020	0										
15	CA	cohort	total_discount	new_non_sub		8/1/2020	0										
16	CA	cohort	total_discount	new_non_sub		10/1/2020	0										
17	CA	cohort	total_discount	new_non_sub		1/1/2021	0										
18	CA	cohort	total_discount	new_non_sub		4/1/2021	0										
19	CA	cohort	total_discount	new_non_sub		5/1/2021	0										
20	CA	cohort	total_discount	new_non_sub		7/1/2021	0										
21	CA	cohort	total_discount	new_non_sub		9/1/2021	0										
22	CA	cohort	total_discount	new_non_sub		12/1/2021	0										
23	CA	cohort	total_discount	new_non_sub		1/1/2022	326										
24	CA	cohort	total_discount	new_non_sub		2/1/2022	1037										
25	CA	cohort	total_discount	new_non_sub		3/1/2022	1276										
26	CA	cohort	total_discount	new_non_sub		4/1/2022	1500										
27	CA	cohort	total_discount	new_non_sub		5/1/2022	1905										
28	CA	cohort	total_discount	new_non_sub		6/1/2022	1198										
29	CA	cohort	total_discount	new_non_sub		7/1/2022	668										
30	CA	cohort	total_discount	new_non_sub		8/1/2022	1134										
31	CA	cohort	total_discount	new_non_sub		9/1/2022	0										
32	CA	cohort	total_discount	new_sub_at_first		3/1/2020	11										
33	CA	cohort	total_discount	new_sub_at_first		5/1/2020	8										
34	CA	cohort	total_discount	new_sub_at_first		7/1/2020	13										
35	CA	cohort	total_discount	new_sub_at_first		9/1/2020	28										
36	CA	cohort	total_discount	new_sub_at_first		10/1/2020	26										
37	CA	cohort	total_discount	new_sub_at_first		1/1/2021	30										
38	CA	cohort	total_discount	new_sub_at_first		2/1/2021	13										
39	CA	cohort	total_discount	new_sub_at_first		3/1/2021	13										
40	CA	cohort	total_discount	new_sub_at_first		4/1/2021	13										
41	CA	cohort	total_discount	new_sub_at_first		5/1/2021	13										
42	CA	cohort	total_discount	new_sub_at_first		9/1/2021	45										
43	CA	cohort	total_discount	new_sub_at_first		10/1/2021	29										
44	CA	cohort	total_discount	new_sub_at_first		11/1/2021	25										
45	CA	cohort	total_discount	new_sub_at_first		12/1/2021	27										
46	CA	cohort	total_discount	new_sub_at_first		1/1/2022	10651										
47	CA	cohort	total_discount	new_sub_at_first		2/1/2022	20963										
48	CA	cohort	total_discount	new_sub_at_first		3/1/2022	12463										
49	CA	cohort	total_discount	new_sub_at_first		4/1/2022	13702										
50	CA	cohort	total_discount	new_sub_at_first		5/1/2022	12473										
51	CA	cohort	total_discount	new_sub_at_first		6/1/2022	8979										
52	CA	cohort	total_discount	new_sub_at_first		7/1/2022	11216										
53			total_discount	new_sub_at_first		8/1/2022	18810										
54			total_discount	repeat_non_sub 2018-07		9/1/2018	0										
55	CA	cohort	total_discount	repeat_non_sub 2018-07		11/1/2018	0										

Updated Sep 1

QueryOrders ExtractOrders QuerySales ExtractSales QueryUSChurned ExtractUSChurned QueryDiscount ExtractDiscount Explore

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Overview of Dashboards and Components

The goal of this section is to provide a high-level overview of each dashboard created during SOW#1, providing insight into functionality and origins of each Look.

During this SOW, Contractor agreed to deliver the two beginning tabs of the Company Report dashboard (Daily Flash & Overview), the New Product Launch dashboard and the Finance Cohort Export. Below will be a high-level, technical introduction to each deliverable:

Daily Flash

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The Daily Flash tab of the Company Report is intended to give high-level insight into how Client is performing across its topline metrics. Contents include 'Yesterday's Topline Metrics' and multiple visuals on Net Sales, Orders, AOV, New Customers, CPA, and Media Spend per the date range bucket (last 7/30/90 days, or ytd), shipping country and hair type.

Filters

Below are the current filters employed:

Filter	Scope
--------	-------

Filter	Scope
Trend Line Date Range	Affects all trendline tiles; thus, it does not affect the five topline metric tiles, KPI Performance over Time or Last Synced Dates.
Shipping Country	Affects all tiles except Last Synced Dates.

Explores and Tiles

Below are the involved explores in each Look within the dashboard:

Explore	Tiles
order	Tile #1 Net Sales - [Single Value] Tile #2 Customers - [Single Value] Tile #3 Orders - [Single Value] Tile #4 AOV - [Single Value] Tile #5 Sales - [looker_line] Tile #6 Orders - [looker_line] Tile #7 New Customers - [looker_line] Tile #8 AOV (2) - [looker_line]
marketing_spend_and_orders	Tile #1 CPA & Media Spend - [looker_column] Tile #2 CPA Change - [Single Value]
drv_pdt_daily_flash	Tile #1 KPI Performance over Time - [looker_grid]
last_synced	Tile #1 Last Synced Dates - [looker_grid]

Dashboard Components

Sub-Navigation Pane

At the top of the dashboard, one finds a sub-navigation panel to navigate across the different tabs within the Company Report:

Daily flash Overview

YESTERDAY'S TOPLINE METRICS The date range filter does not apply to the topline metrics

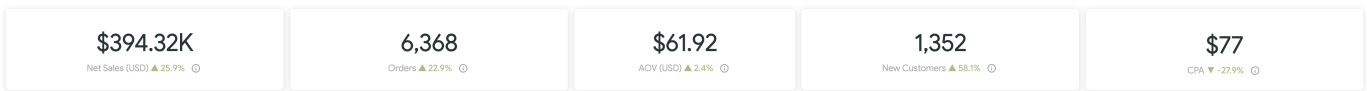
This tile consists of the following HTML code:

```
<div style="border:solid 0px #000000; border-radius: 4px; padding-top: 10px; padding: 12px;
background: #283333; height: 100%">
<a href="https://Client.cloud.looker.com/dashboards/14"></a><a
href="https://Client.cloud.looker.com/dashboards/23"></a></div>
<div style="height:10px;"><br></div>
```

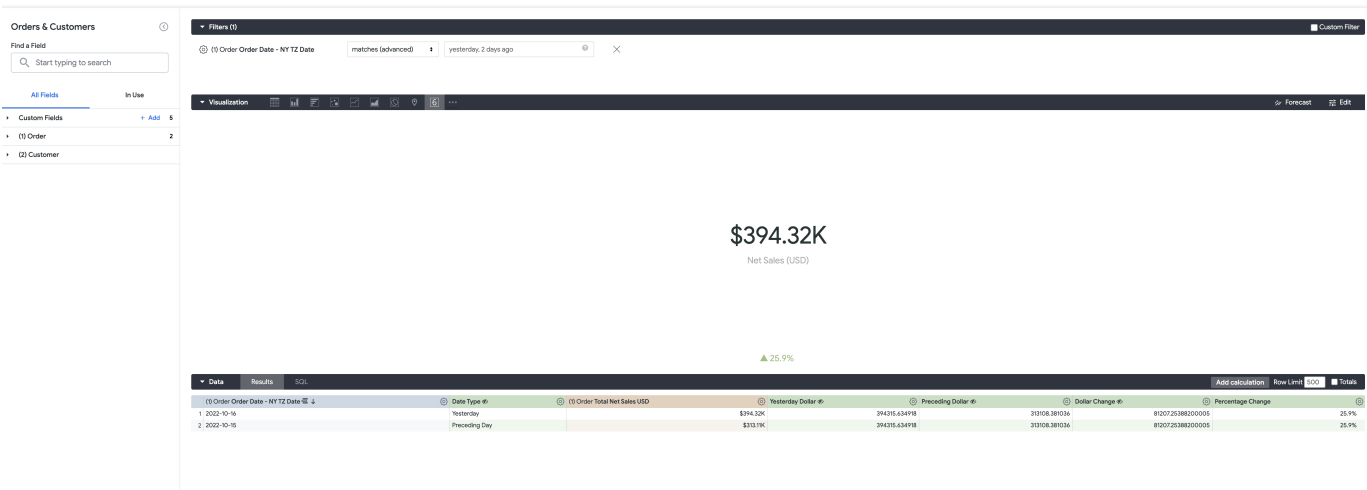

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The tile renders custom assets housed within the **Client LOOKER STATIC ASSETS** shared Google Drive hosted by Client, as well as makes an HTML clarification of the limits of the date range filter.

Topline Metrics Snapshot



These topline metrics are aggregates of yesterday's performance; the comparisons directly below them compare to aggregates from the preceding day. Below exposes the filters and table calculations involved in generating these aggregates and comparisons:



KPI Performance over Time Crosstab

Below the topline metrics holds the KPI performance over time crosstab. See screenshot below:

KPI Performance over Time 🕒										
Time Frame	Net Sales	Net Sales Δ	Orders	Orders Δ	Net AOV	Net AOV Δ	New Customers	New Customers Δ	CPA	CPA Δ
Yesterday	\$394.32K	▲ 32.9%	6,368	▲ 31.5%	\$61.92	▲ 2.1%	1,352	▲ 12.1%	\$79	▼ 19.3%
Last 7 Days	\$2.28M	▲ 22.0%	38,632	▲ 18.7%	\$59.04	▲ 4.0%	6,429	▼ 27.5%	\$105	▲ 8.7%
Last 30 Days	\$10.17M	▲ 29.5%	169,437	▲ 25.4%	\$60.05	▲ 5.5%	29,906	▼ 11.6%	\$102	▲ 9.8%
Last 365 Days	\$115.92M	▲ 32.5%	1,923,978	▲ 33.0%	\$60.25	▼ 0.8%	410,629	▼ 3.0%	\$91	▲ 11.9%

This crosstab addresses KPI performance in across different time bins (yesterday, last 7 days, last 30 days and last 365 days). Additionally, the delta between each KPI is calculated against the same KPI's performance in the preceding year. The following were the challenges face in the development of this rollup report:

- Binary limitations in **case when** statements when binning data
- Calculating deltas to compare against same day of the week in the preceding year instead of same date last year
- Performance speed

These challenges were cared for by creating separate dimensions for each time bin for each dimension/measure for the calculation of each field within the Look. Below is a table of each dimension/measure with its corresponding value to the Look:

Field	View	Description
yesterday last_7_days last_30_days last_365_days	order & marketing_spend	yesno dimensions flagging if a date is within the respective time range. These are utilized in fields described later below.
yesterday_prior_year last_7_days_prior_year last_30_days_prior_year last_365_days_prior_year	order & marketing_spend	yesno dimensions used to perform similar tasks to above, except for the same day last year. These are utilized in fields below to calculate the deltas.
order_pubkey_yesterday order_pubkey_last_7_days order_pubkey_last_30_days order_pubkey_last_365_days	order	These dimensions care for the count distinct data issue caused by the binary classification employed by a simple case when solution; a simple running total will not suffice. Thus, creating separate dimensions for each time bin, with case when statements within these dimensions to isolate orders within a given time frame, preserved order pubkeys to accurately perform a count distinct in a later measure.
customer_pubkey_yesterday customer_pubkey_last_7_days customer_pubkey_last_30_days customer_pubkey_last_365_days	order	These dimensions perform the same functions as above. They dedicate separate fields for different time bins to accurately care for distinct aggregations upon customer pubkey.
order_date_yesterday order_date_last_7_days order_date_last_30_days order_date_last_365_days	order	These dimensions isolate dates within the specified time bin. Within the dimension holds a case when statement, returning the order date if within the respective date range, and null if not.
order_date_yesterday_prior_year order_date_last_7_days_prior_year order_date_last_30_days_prior_year order_date_last_365_days_prior_year	order	Similar to above, these dimensions isolate dates within the specified time bin. Within the dimension holds a case when statement, returning the order date if within the respective date range, and null if not.
daily_flash_times	order	Returns 'Yesterday', 'Last 7 Days', 'Last 30 Days' or 'Last 365 Days' strings if the order date bins above fall within the respective time bins.

Field	View	Description
flash_rank	order	A simple <code>case when</code> statement that convertes <code>daily_flash_times</code> to integers to support chronological sort (since <code>daily_flash_times</code> is a string and would therefore sort incorrectly).
total_number_orders_daily_flash	order	<code>count_distinct</code> measure with a <code>case when</code> statement in the <code>sql</code> parameter. In this <code>case when</code> statement, it evaluates which bin type <code>daily_flash_times</code> is returning, then pulling the corresponding <code>order_pubkey</code> dimension to perform a count distinct. Thus, if <code>daily_flash_times = 'Last 30 Days'</code> then <code>order_pubkey_last_30_days</code> is returned.
total_number_orders_yesterday_py total_number_orders_last_7_days_py total_number_orders_last_30_days_py total_number_orders_last_365_days_py	order	<code>count_distinct</code> measures that calculate the prior year values of total orders. These had to be separated into separate measures because the primary dates were already assumed in the <code>daily_flash_times</code> ; creating a <code>daily_flash_times_prior_year</code> would double the crosstab size and not allow side-by-side comparison. Thus, separate, filtered measures were created, later being aggregated via a table calculation in the final Look.
distinct_new_customer_count_daily_flash	order	Same logic for <code>total_number_orders_daily_flash</code> , different pubkey returns (<code>customer_pubkey</code> return instead of <code>order_pubkey</code>).
distinct_new_customer_count_yesterday_py distinct_new_customer_count_last_7_days_py distinct_new_customer_count_last_30_days_py distinct_new_customer_count_last_365_days_py	order	Same logic for <code>total_number_orders_[time_bin]_py</code> measures above, except calculating the total number of new distinct customers instead of distinct order counts.
total_net_sales_USD_running_total	order	A simple running total upon <code>total_net_sales_USD</code> . This cares for the binary classification issue originally grappled with the <code>daily_flash_times</code> .

Field	View	Description
total_net_sales_USD_yesterday_py total_net_sales_USD_last_7_days_py total_net_sales_USD_last_30_days_py total_net_sales_USD_last_365_days_py	order	<p><code>sum_distinct</code> measures that calculate the prior year values of total net sales. These had to be separated into separate measures because the primary dates were already assumed in the <code>daily_flash_times</code>; creating a <code>daily_flash_times_prior_year</code> would double the crosstab size and not allow side-by-side comparison. Thus, separate, filtered measures were created, later being aggregated via a table calculation in the final Look.</p>
total_media_spend_running_total	marketing_spend	<p>A simple running total upon <code>total_media_spend</code>. This cares for the binary classification issue originally grappled with the <code>daily_flash_times</code>.</p>
total_media_spend_yesterday_py total_media_spend_last_7_days_py total_media_spend_last_30_days_py total_media_spend_last_365_days_py	marketing_spend	<p><code>sum</code> measures that calculate the prior year values of total marketing spend. These had to be separated into separate measures because the primary dates were already assumed in the <code>daily_flash_times</code>; creating a <code>daily_flash_times_prior_year</code> would double the crosstab size and not allow side-by-side comparison. Thus, separate, filtered measures were created, later being aggregated via a table calculation in the final Look.</p>
cpa_yesterday_py cpa_last_7_days_py cpa_last_30_days_py cpa_last_365_days_py	marketing_spend	<p>Measures that calculate the prior year values of CPA. These had to be separated into separate measures because the primary dates were already assumed in the <code>daily_flash_times</code>; creating a <code>daily_flash_times_prior_year</code> would double the crosstab size and not allow side-by-side comparison. Thus, separate, filtered measures were created, later being aggregated via a table calculation in the final Look.</p>

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While these dimensions were validated and finalized, they were employed within the crosstab, along with some table calculations to calculate the deltas. However, performance issues were encountered that made the Look run upwards towards 10 minutes. To care for this, we created a PDT (`drv_pdt_daily_flash`) that is managed by the `datagroup_orders` datagroup, which has a max cache age of 24 hours and a `sql_trigger` of

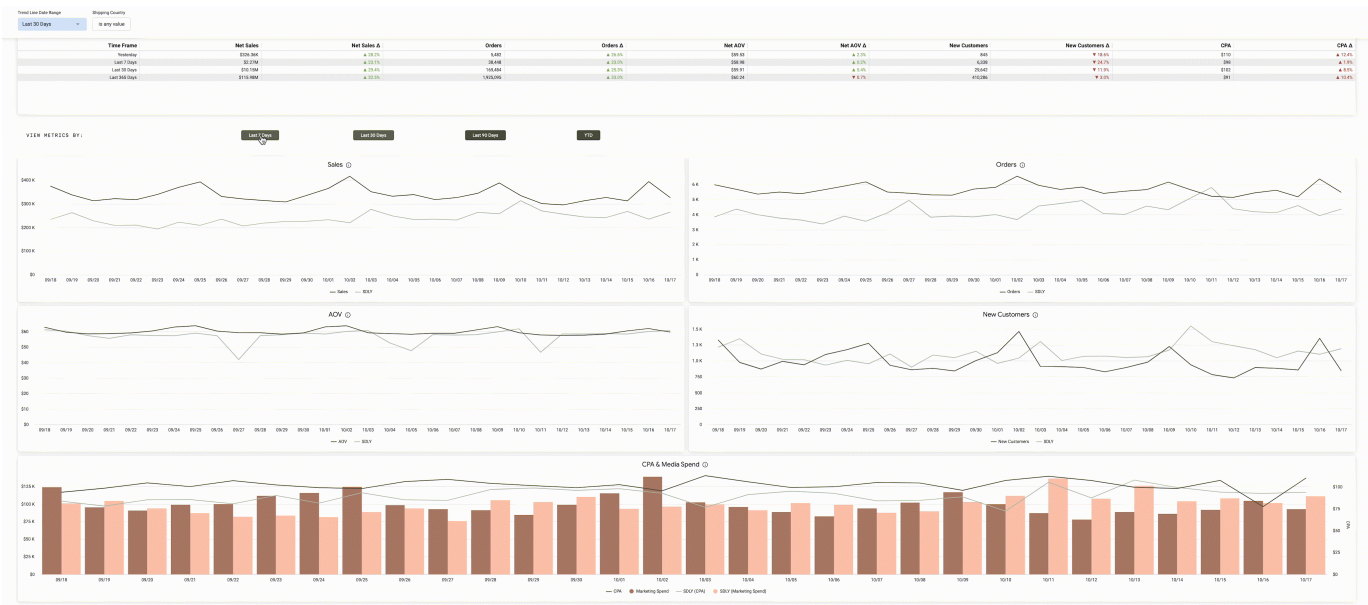
```
SELECT max(order_pubkey) FROM
`Clienthair.Client_bi_datamart_growth_customer_ltv_customer_order.order
```

with an `interval_trigger` of 12 hours.

After the PDT, the crosstab was refactored to point towards `drv_pdt_daily_flash`, and it is significantly more performant.

Topline Metric Trendlines

Below the KPI crosstab is the trend lines for the five topline metrics over time. The default filter for these is to show the last 30 days, and one may toggle between alternate views to view the trends by the last 7 days, the last 30 days, the last 90 days or YTD. Below is a demo of the UX functionality:



In order to allow for dynamic date range renderings in comparison with the prior year, a number of custom dimensions and measures had to be employed. Below is a list of all the involved fields, as well as descriptions of their utility:

Field	View	Description
<code>select_timeframe</code>	order	A parameter with values of 'Last 7 Days', 'Last 30 Days', 'Last 90 Days' and 'YTD'. This parameter is manipulated when buttons are clicked on the dashboard to change the date range and date formatting.
<code>current_vs_previous_period</code>	order	A dynamic dimension that allows for comparison between current year and last year. This dimension employs a warpper of liquid logic around a <code>case when</code> statement, and it is heavily dependent on the value of <code>select_timeframe</code> .
<code>selected_dynamic_day_of</code>	order	A dynamic dimension used to change the date formatting depending upon the value for <code>select_timeframe</code> .
<code>selected_dynamic_day_of_sort</code>	order	A dynamic dimension used to determine sort order based upon <code>select_timeframe</code> to enforce chronological sorting. This could be an opportunity for simplification.
<code>selected_dynamic_duration</code>	order	Returns the week index or month number based on the selection of <code>select_timeframe</code> ; this could be an opportunity for deprecation.

Overview

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The Overview tab is intended to provide a level more detail than the Daily Flash tab. The date filter affects all trend lines similar to the dynamic trend line date filter in the Daily Flash tab. This dashboard displays Yesterday's Topline Metrics and multiple visuals on Net Sales, Orders, and AOV per the date range bucket (last 7/30/90 days, or ytd), shipping country and hair type.

The looks mainly employs simple aggregations that Client pre-supplied on view/data mart creation. However, each Look (besides the topline metrics) uses the `current_vs_previous_period` dimensions to filter by date. Some Looks incorporate prior year comparison; others do not. The ones that do not compare the metrics to prior year have "Previous" filtered out. is in the use of the Sales Type Breakout & Monthly Subscription Revenue area charts, as well as the Cumulative Sales column/line combination chart.

Filters

Below are the current filters employed:

Filter	Scope
Date Range	This filter applies to all tiles except the five 'Yesterday's Topline Metrics' tiles and the Last Synced Dates tile.
Shipping Country	Applies to all tiles except the Last Synced Dates tile.
Hair Type	Applies to all tiles except the Last Synced Dates tile.

Explores and Tiles

Below are the involved explores in each Look within the dashboard:

Explore	Tiles
---------	-------

Explore	Tiles
order	Tile #1 Net Sales - [Single Value]
	Tile #2 Customers - [Single Value]
	Tile #3 Orders - [Single Value]
	Tile #4 AOV - [Single Value]
	Tile #5 Net Sales (2) - [looker_column]
	Tile #6 Sales Type Breakout - [looker_area]
	Tile #7 New vs Repeat Net Sales - [looker_bar]
	Tile #8 Sub vs Non-Sub Net Sales - [looker_bar]
	Tile #9 Orders by Country - [looker_column]
	Tile #10 New vs Repeat - [looker_donut_multiples]
	Tile #11 Sub vs Non-Sub - [looker_donut_multiples]
	Tile #12 Total Discount Amount - [single_value]
	Tile #13 Average Discount Value - [single_value]
	Tile #14 Discount Trend - [looker_column]
	Tile #15 AOV by Order Type - [looker_line]
	Tile #16 Sub Sales - [single_value]
	Tile #17 First Order AOV (Copy 2) - [single_value]
	Tile #18 First Order AOV - [single_value]
	Tile #19 First Order AOV (Copy) - [single_value]
	Tile #20 First Order AOV (Copy 3) - [single_value]
	Tile #22 Cumulative Orders - [looker_column]
	Tile #24 Cumulative Sales - [looker_line]
	Tile #25 New Sales - [single_value]
	Tile #26 Repeat Sales - [single_value]
	Tile #27 Non-Sub Sales - [single_value]
	Tile #28 AOV by Order Instance - [looker_bar]
marketing_spend_and_orders	Tile #1 CPA Change - [Single Value]
last_synced	Tile #1 Last Synced Dates - [looker_grid]

Sub-Navigation Pane

The Sub-Navigation pane is similar to the Daily Flash tab, with the exception being that the "Overview" is underlined to signal selection as opposed to "Daily Flash". The rendering and HTML are below:

Daily flash

Overview

YESTERDAY'S TOPLINE METRICS

The date range filter does not apply to the topline metrics

```
<div style="border:solid 0px #000000; border-radius: 4px; padding: 10px 12px;
background: #283333; height: 100%">
  <a href="https://Client.cloud.looker.com/dashboards/14"></a><a
href="https://Client.cloud.looker.com/dashboards/23"></a></div>
```

Topline Metrics Snapshot

\$394.32K	6,368	\$61.92	1,352	\$77
Net Sales (USD) ▲ 25.9% ⓘ	Orders ▲ 22.9% ⓘ	AOV (USD) ▲ 2.4% ⓘ	New Customers ▲ 58.1% ⓘ	CPA ▼ -27.9% ⓘ

Orders & Customers

Find a Field

Start typing to search

All Fields

In Use

Custom Fields

Add

1

Order

2

Customer

Filters (1)

Order Order Date - NY TZ Date

matches (advanced)

yesterday, 2 days ago

Visualization

Forecast

Edit

\$394.32K

Net Sales (USD)

▲ 25.9%

Date	Results	SQL	Add calculation	Row Limit	SQL	Totals
Order Order Date - NY TZ Date	Date Type	Order Total Net Sales USD	Yesterday Dollar	Preceding Dollar	Dollar Change	Percentage Change
2022-10-16	Yesterday	\$394.32K	\$394.316K	\$394.316K	\$394.316K	25.9%
2022-10-15	Preceding Day	\$313.19K	\$313.19K	\$313.19K	\$313.19K	25.9%

NET SALES

\$2.66M

New ⓘ

\$7.49M

Repeat ⓘ

New vs Repeat Net Sales ⓘ

26.2%

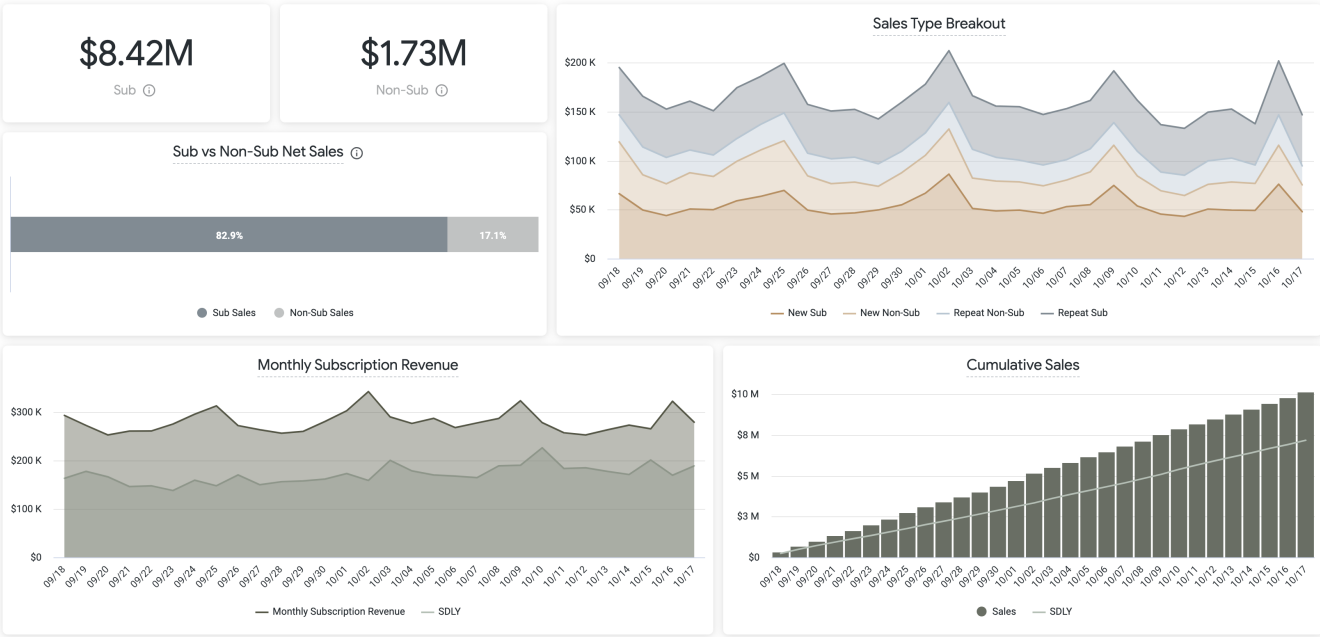
73.8%

● New Sales ● Repeat Sales

Net Sales

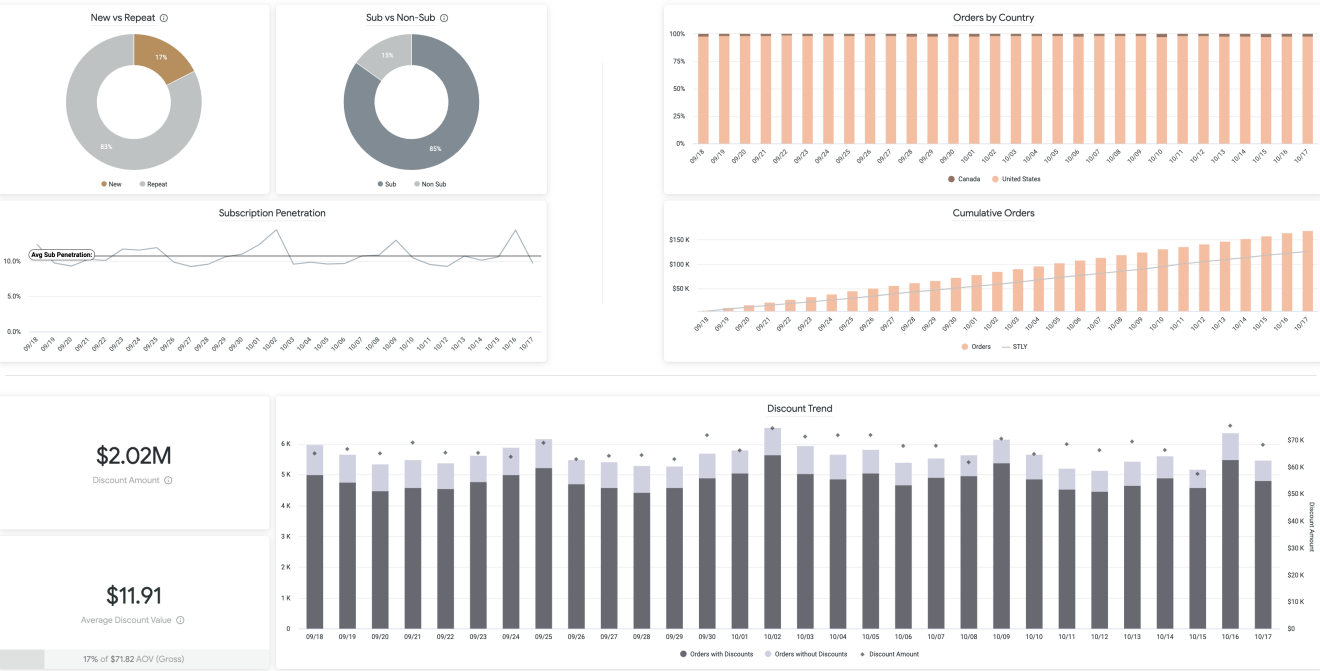
Month	New Sales (\$K)	Repeat Sales (\$K)	Total Net Sales (\$K)
9/1/18	115	255	370
9/1/19	80	250	330
9/1/20	75	235	310
9/1/21	80	230	310
9/1/22	80	230	310
9/1/23	95	245	340
9/1/24	110	260	370
9/1/25	115	285	400
9/1/26	80	245	325
9/1/27	75	240	315
9/1/28	75	230	305
9/1/29	70	230	300
9/1/30	85	245	330
10/1/01	100	260	360
10/1/02	125	280	405
10/1/03	80	270	350
10/1/04	75	250	325
10/1/05	75	255	330
10/1/06	70	240	310
10/1/07	75	250	325
10/1/08	85	260	345
10/1/09	115	275	390
10/1/10	75	260	335
10/1/11	65	240	305
10/1/12	60	230	290
10/1/13	70	240	310
10/1/14	70	250	320
10/1/15	75	235	310
10/1/16	110	280	390
10/1/17	70	250	320

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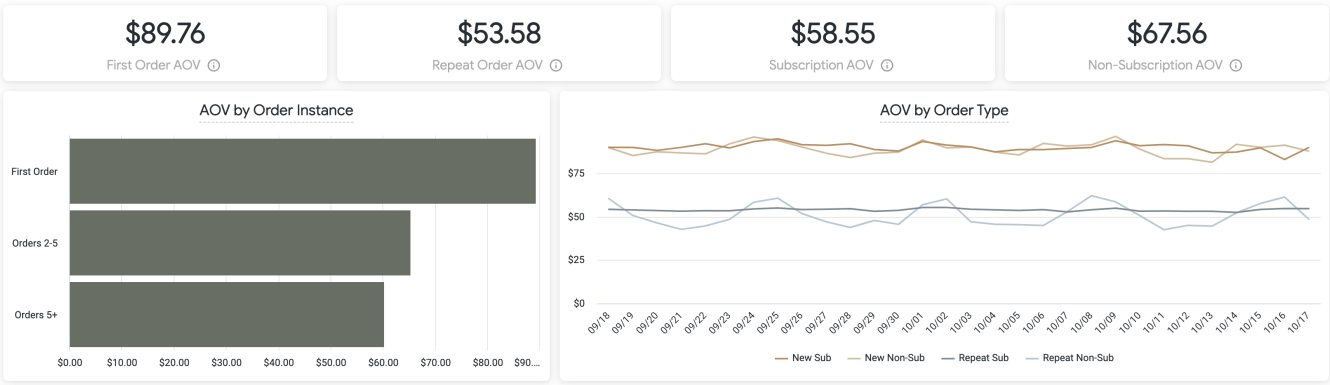
Orders

The Orders section addresses order driven metrics. It provides insight into number and types of orders, along with discount data. Below is a snapshot of the section:



AOV

The AOV section provides insight into various AOV contexts: first order, repeat order, subscription, non-subscription, as well as AOV by order tier. Below is a snapshot of the section:



New Product Launch

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The New Product Launch Dashboard provides a "Days Since Launch" perspective on the selected new product. The Days since launch parameter provides 30/60/90/1 yr/ and unlimited options for selection. Components required to support this parameter are detailed at the end of this section.

The looks mainly employs simple aggregations that Client pre-supplied on view/data mart creation. However, each Look (besides the topline metrics) uses the `current_vs_previous_period` dimensions to filter by date. Some Looks incorporate prior year comparison; others do not. The ones that do not compare the metrics to prior year have "Previous" filtered out. is in the use of the Sales Type Breakout & Monthly Subscription Revenue area charts, as well as the Cumulative Sales column/line combination chart.

Days Since Launch Components

In order to support the requirement for a variety of 'days since launch' options and correspondingly adjust the 'tick density' on the x-axis of the impacted trend charts, number of dimension were created to complement the 'select days since launch' parameter. These fields are described in the table below, with hyperlinks on the fields to LookML snippets detailing their construction:

Field	View	Description
launch_date	drv_launch_date_by_product	A dimension group based on min(order_date) by product_name used to calculate launch duration.
launch	order_items	A duration dimension group surfacing number of days since launch based on launch_date as start and order_date as duration end.
select_days_since_launch	order_items	The parameter for providing options for 30,60,90, 1 yr, and 'no limit' time frame options for days since launch
selected_dynamic_launch_days_label	order_items	A dynamic dimension used to aggregate tick density from daily to 7 or 30 day increments based on the selected parameter value.

Field	View	Description
selected_dynamic_launch_days_duration	order_items	This dimension is used to translate the string parameter value to a number for the custom filter below. Values returned are 31,61,91, 366, or 9999.
[custom filter](#custom filter)	NA - specific tiles	A custom filter was applied on the impacted tiles= to translate the parameter to the requisite filter.

Filters

Below are the current filters employed:

Filter	Scope
Days Since Launch	All Trend Tiles: Sales Trend (with measure substitution for Items, AOV, and Customers) Customer Origin Trend Product Trend - Gross Sales % Discount on Orders Orders by Discount Type Prod Launch Comparisons (AOV)
Shipping Country	All tiles except Last Synced Dates tile.
Hair Type	All tiles except Last Synced Dates tile.
Measure Selector	Sales Trend tile
New Products	All tiles except Last Synced Dates tile.

Explores and Tiles

Below are the involved explores in each Look within the dashboard:

Explore	Tiles
---------	-------

Explore	Tiles
order_items	Tile #1 Gross Sales - [Single Value]
	Tile #2 Orders - [Single Value]
	Tile #3 Items - [Single Value]
	Tile #4 AOV - [Single Value]
	Tile #5 New Customers - [Single Value]
	Tile #6 Sales Trend - [looker_line]
	Tile #7 New(Non-Sub) - [Single Value]
	Tile #8 New(Sub) - [Single Value]
	Tile #9 Repeat(Non-Sub) - [Single Value]
	Tile #10 Repeat(Sub) - [Single Value]
	Tile #11 Subscription vs Non-Subscription - [Donut Multiples]
	Tile #12 Customer Origin - [looker_bar]
	Tile #13 Customer Origin Trend - [looker_column_stacked]
	Tile #14 Subscription Order Composition - [Packed Bubble]
	Tile #15 Product Trend (Gross Sales) - [looker_line]
	Tile #16 Product Launch Comparison Trend - [looker_line]
order	Tile #1 Orders by Customer Type - [looker_bar]
	Tile #2 % of Discount on Orders - [Donut Multiples]
	Tile #3 Orders by Discount Type - [looker_bar]
customer	Tile #1 Customer Origin - [looker_bar]
	Tile #2 Customer Origin Trend - [looker_column_stacked]
cohort_finance_name	Tile #1 Orders by Customer Type - [looker_bar]
last_synced	Tile #1 Last Synced Dates - [looker_grid]

Sub-Navigation Pane

The Sub-Navigation pane provides a hyperlink to the Company Report.

Launch highlights

Additional product insight can be found in the [Company Report](#)

YESTERDAY'S TOPLINE METRICS

The date range filter does not apply to the topline metrics

This tile consists of the following html code:

```
<div style="border:solid 0px #000000; border-radius: 4px; padding: 10px 12px;
background: #283333; height: 100%">
  <a
href="https://Client.cloud.looker.com/dashboards/14"></a></div>
<div style="height:10px;"><br></div>
<div style="border:solid 0px #000000; border-radius: 4px;
background: #FBF6F2; padding: 12px; height: 100%">
<div><img style="padding-right:20px;" src="https://drive.google.com/uc?
```

Topline Metrics Snapshot

\$394.32K

Net Sales (USD) ▲ 25.9%

6,368

Orders ▲ 22.9%

\$61.92

AOV (USD) ▲ 2.4%

1,352

New Customers ▲ 58.1%

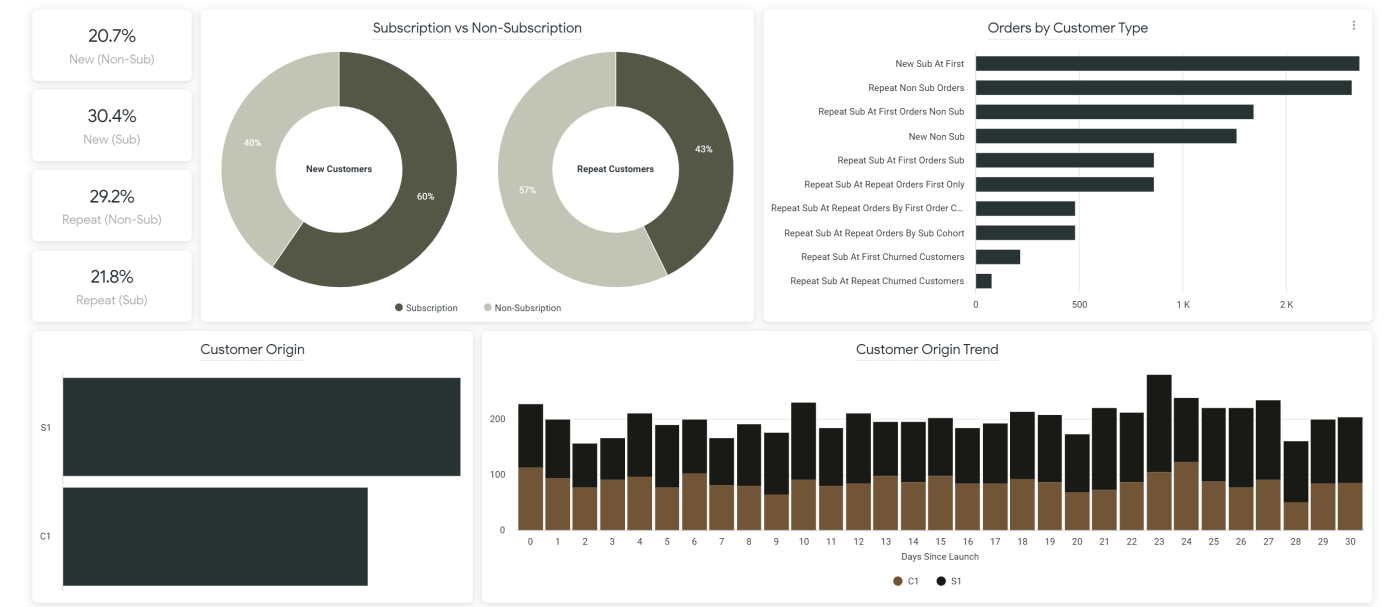
\$77

CPA ▼ -27.9%

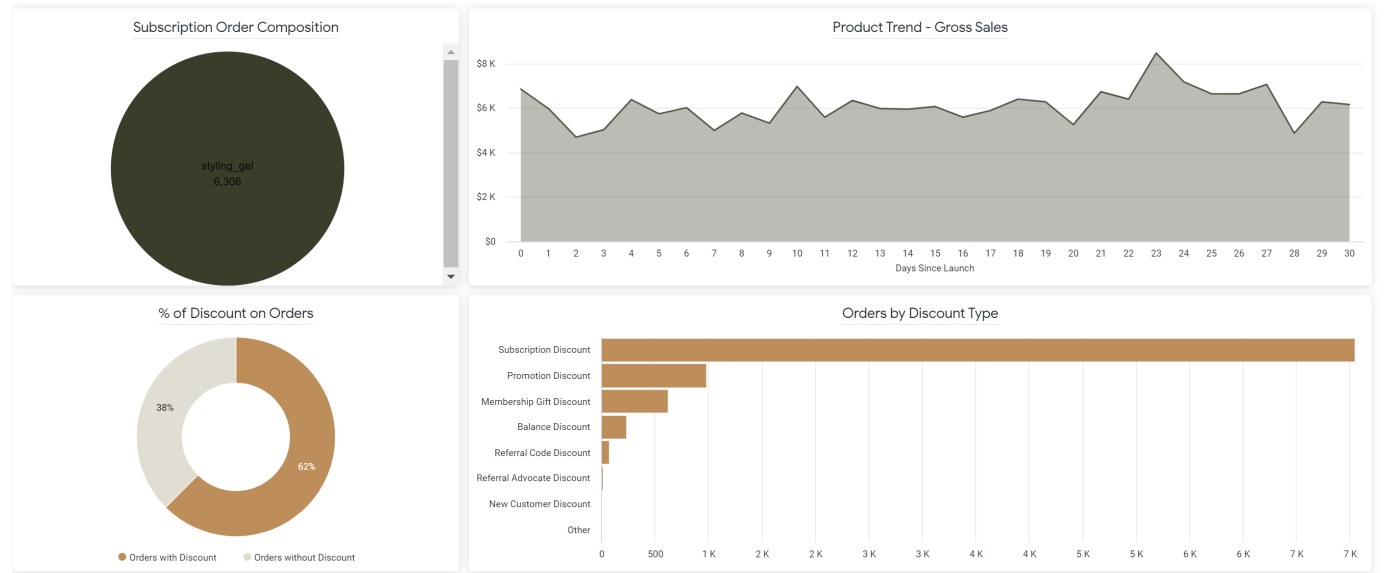
This section provides the option to select a trend view of one of four measures determined by the button selector. The timespan is controlled by the days since launch parameter selected in the filter section. The x-axis tick density is dynamic, changing from daily, 7-day, and 30-day increments based on the selected timespan. The following is a snapshot of this section with Gross Sales selected:



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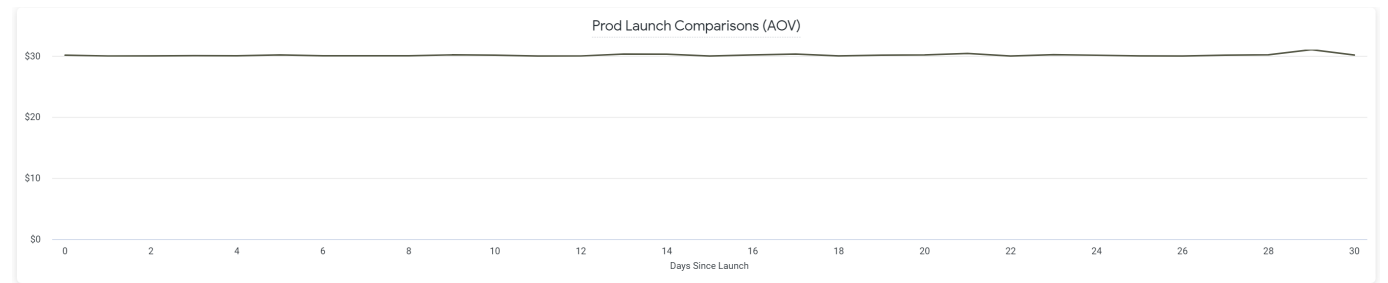


Customer orders second section



Product Launch Comparison

This section is to provide a product component comparison by AOV across the selected days since launch timespan:



Finance Cohort Looks

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Five Looks were produced to support the 'Finance Cohort Name' reporting requirement. The Looks themselves are straightforward, though some have large result sets that will require elevated user permissions for large record set downloads.

Three views were required to support these looks and are detailed in the view section of this document. Listed here for convenience/context.

View	Description
cohort_finance_name	This view is used to generate a list of 16 [cohort_finance_names](#Cohort Finance Name) as identified in the requirements from Finance. Corresponding filtered measures are mapped to these names via case statements in the order_cohort_extension and orer_items_cohort_extension views.
order_cohort_extension	This view extends the order view, adding cohort_finance_name from the required join to the cohort_finance_names view. This is required to accomplish the cohort_name mapping to the appropriate measure. Example of this mapping [here.](#Cohort Finance Units)The net sales and discount measures are redefined in this view, as sum_distinct is required as measure type due to the required cross join to cohort_finance_name.
order_items_cohort_extension	This view extends the order_items view, adding cohort_finance_name from the required join to the cohort_finance_names view required to accomplish the cohort_name mapping at product level to the appropriate measure. Measure duplication was not required for this view in that the measures are limited to type count_distinct.

Screen grabs of the five Looks are provided below:

Extract Churned by Finance Cohort

1759 rows - from cache · 45m ago

Filters (3)

(2) Customer Last Order Shipping Country

is equal to

US

+

Cohort Finance Name Cohort Finance Name

is equal to

repeat_sub_at_repeat_churned_customers

repeat_sub_at_first_churned_customers

+

(1) Order Cohort Finance Orders

is greater than

0

+

Visualization

	Last Order Shipping Country	Cohort Finance Name	First Order - Company Month	First Subscription Canceled at Date - Company - NY TZ Month	Cohort Finance Churned Customer Counts
1	US	Repeat Sub At First Churned Customers	2019-10	2021-05	1
2	US	Repeat Sub At First Churned Customers	2019-11	2021-11	1
3	US	Repeat Sub At First Churned Customers	2019-12	2020-10	1
4	US	Repeat Sub At First Churned Customers	2019-12	2021-01	1
5	US	Repeat Sub At First Churned Customers	2019-12	2021-03	1
6	US	Repeat Sub At First Churned Customers	2020-01	2020-06	2
7	US	Repeat Sub At First Churned Customers	2020-01	2021-01	1
8	US	Repeat Sub At First Churned Customers	2020-02	2020-02	5
9	US	Repeat Sub At First Churned Customers	2020-02	2020-03	22
10	US	Repeat Sub At First Churned Customers	2020-02	2020-04	15
11	US	Repeat Sub At First Churned Customers	2020-02	2020-05	16
12	US	Repeat Sub At First Churned Customers	2020-02	2020-06	11
13	US	Repeat Sub At First Churned Customers	2020-02	2020-08	4
14	US	Repeat Sub At First Churned Customers	2020-02	2020-09	4
15	US	Repeat Sub At First Churned Customers	2020-02	2020-10	5
16	US	Repeat Sub At First Churned Customers	2020-02	2020-11	2
17	US	Repeat Sub At First Churned Customers	2020-02	2020-12	6
18	US	Repeat Sub At First Churned Customers	2020-02	2021-01	3
19	US	Repeat Sub At First Churned Customers	2020-02	2021-03	1
20	US	Repeat Sub At First Churned Customers	2020-02	2021-04	3
21	US	Repeat Sub At First Churned Customers	2020-02	2021-05	3
22	US	Repeat Sub At First Churned Customers	2020-02	2021-06	2

Extract Discounts by Finance Cohort

▼ Filters (3)

🔍 (2) Customer Last Order Shipping Country

is equal to

US

+

🔍 Cohort Finance Name Cohort Finance Name

is equal to

+

🔍 (1) Order Cohort Finance Discounts

is greater than

0

+

▼ Visualization

	Last Order Shipping Country	Cohort Finance Name	First Order - Company Month	Order Date - NY TZ Month	Cohort Finance Discounts
1	US	Repeat Sub At Repeat Orders First Only	2020-02	2022-08	\$1,551
2	US	Repeat Sub At Repeat Orders First Only	2021-05	2022-05	\$38,075
3	US	Repeat Sub At Repeat Orders First Only	2020-11	2021-10	\$22,925
4	US	Repeat Sub At Repeat Orders First Only	2021-12	2022-07	\$48,719
5	US	Repeat Sub At Repeat Orders First Only	2021-06	2022-08	\$24,629
6	US	Repeat Sub At Repeat Orders First Only	2020-04	2020-09	\$27,283
7	US	Repeat Sub At Repeat Orders First Only	2020-06	2020-06	\$1,389
8	US	Repeat Sub At Repeat Orders First Only	2021-12	2022-03	\$98,789
9	US	Repeat Sub At Repeat Orders First Only	2022-06	2022-06	\$6,156
10	US	Repeat Sub At Repeat Orders First Only	2020-06	2021-05	\$15,036
11	US	Repeat Sub At Repeat Orders First Only	2019-11	2020-05	\$33
12	US	Repeat Sub At Repeat Orders First Only	2021-06	2021-07	\$53,624
13	US	Repeat Sub At Repeat Orders First Only	2021-06	2022-06	\$26,832
14	US	Repeat Sub At Repeat Orders First Only	2020-07	2021-07	\$16,199
15	US	Repeat Sub At Repeat Orders First Only	2020-05	2021-09	\$8,924
16	US	Repeat Sub At Repeat Orders First Only	2020-06	2021-10	\$10,198
17	US	Repeat Sub At Repeat Orders First Only	2020-02	2021-05	\$1,989
18	US	Repeat Sub At Repeat Orders First Only	2021-08	2022-04	\$34,052
19	US	Repeat Sub At Repeat Orders First Only	2020-03	2021-07	\$7,310
20	US	Repeat Sub At Repeat Orders First Only	2021-07	2021-10	\$60,895
21	US	Repeat Sub At Repeat Orders First Only	2021-04	2021-07	\$89,448
22	US	Repeat Sub At Repeat Orders First Only	2020-05	2022-10	\$4,418
23	US	Repeat Sub At Repeat Orders First Only	2021-03	2022-05	\$37,580
24	US	Repeat Sub At Repeat Orders First Only	2019-12	2021-04	\$93

Extract Orders by Finance Cohort ♡

▼ Filters (3)

🔗 (2) Customer Last Order Shipping Country

is equal to

US x CA x

+

🔗 Cohort Finance Name Cohort Finance Name

is equal to

+

🔗 (1) Order Cohort Finance Orders

is greater than

0

+

▼ Visualization

	Last Order Shipping Country ^	Cohort Finance Name ^	First Order - Company Month ^	Order Date - NY TZ Month ^	Cohort Finance Orders
1	CA	New Non Sub	2018-07	2018-07	1
2	CA	New Non Sub	2018-10	2018-10	2
3	CA	New Non Sub	2018-12	2018-12	1
4	CA	New Non Sub	2019-01	2019-01	1
5	CA	New Non Sub	2019-02	2019-02	1
6	CA	New Non Sub	2019-04	2019-04	1
7	CA	New Non Sub	2019-05	2019-05	1
8	CA	New Non Sub	2019-07	2019-07	1
9	CA	New Non Sub	2019-08	2019-08	1
10	CA	New Non Sub	2019-10	2019-10	2
11	CA	New Non Sub	2019-11	2019-11	1
12	CA	New Non Sub	2019-12	2019-12	1
13	CA	New Non Sub	2020-01	2020-01	1
14	CA	New Non Sub	2020-02	2020-02	2
15	CA	New Non Sub	2020-04	2020-04	1
16	CA	New Non Sub	2020-06	2020-06	1
17	CA	New Non Sub	2020-08	2020-08	1
18	CA	New Non Sub	2020-10	2020-10	1
19	CA	New Non Sub	2021-01	2021-01	1
20	CA	New Non Sub	2021-03	2021-03	1
21	CA	New Non Sub	2021-04	2021-04	4
22	CA	New Non Sub	2021-05	2021-05	2
23	CA	New Non Sub	2021-07	2021-07	2
24	CA	New Non Sub	2021-08	2021-08	1
25	CA	New Non Sub	2021-09	2021-09	1
26	CA	New Non Sub	2021-12	2021-12	2
27	CA	New Non Sub	2022-01	2022-01	740

Extract Units by Finance Cohort by Product ♡

2536 rows - 1s - 22m ago

Run

Edit

⌵

▼ Filters (3)

🔗 (2) Customer Last Order Shipping Country

is equal to

US x

+

🔗 Cohort Finance Name Cohort Finance Name

is equal to

new_non_sub x new_sub_at_first x repeat_sub_at_repeat_orders_by_first_order... x repeat_sub_at_repeat_orders_first_only x

+

🔗 (1) Order Cohort Finance Units

is greater than

0

+

▼ Visualization

	Last Order Shipping Country	Order Date - NY TZ Month	Cohort Finance Name	Product Name	Cohort Finance Units
1	US	2022-10	Repeat Sub At Repeat Orders First Only	Brush Agave	9
2	US	2022-10	Repeat Sub At Repeat Orders First Only	Brush Boar	19
3	US	2022-10	Repeat Sub At Repeat Orders First Only	Brush Detangling	24
4	US	2022-10	Repeat Sub At Repeat Orders First Only	Brush Mixed	69
5	US	2022-10	Repeat Sub At Repeat Orders First Only	Complete Set	13
6	US	2022-10	Repeat Sub At Repeat Orders First Only	Conditioner	41,188
7	US	2022-10	Repeat Sub At Repeat Orders First Only	Curl Cream	12,387
8	US	2022-10	Repeat Sub At Repeat Orders First Only	Dry Shampoo	2,594
9	US	2022-10	Repeat Sub At Repeat Orders First Only	Essentials Set	18
10	US	2022-10	Repeat Sub At Repeat Orders First Only	File Bag Grp	4

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Below addresses custom LookML implementations throughout the dashboards. General dimensions/measures will not be discussed here; components that provided creative solutions to business-specific problems will be reserved for this section.

Manifest File

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The project's manifest file ([manifest.lkml](#)) was used to define many constants that are subsequently referenced throughout the projects LookML files using Looker's `@{constant_name}` conventions. Variables include a variety of SQL date range expressions/declarations, dynamic currency formatting, and even a variable for parsing in HTML to accomplish format conversions.

Below addresses the two general types of constants in the manifest file: Formatting Blocks and SQL Constants.

Formatting Blocks

[Return to TOC](#)

A number of liquid formatting and logic components are stored within the manifest file. Below details each component and its purpose:

Dynamic Currency Formatting

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Looker does not have a native functionality to dynamically format currency with a `value_format`; that is, a way to give currency summaries if something is in the millions or thousands with 'M' or 'K', respectively.

Thus, this necessitated utilizing Liquid for dynamic currency formatting via a dimension/measure's `html` parameter. Below is the constant that houses this Liquid logic:

```
constant: dynamic_currency_formatting_us {
  value: "{% assign abs_value = value | abs %}
{% if abs_value >= 1000 and abs_value < 1000000 %}
${{value | divided_by: 1000 | round: 2 }}K
{% elsif abs_value >= 1000000 and abs_value < 1000000000 %}
${{value | divided_by: 1000000 | round: 2 }}M
{% elsif abs_value >= 1000000000 %}
${{value | divided_by: 1000000000 | round: 2 }}B
{% else %}
${{value | round: 2 }}
{% endif %}"}
```

Proper Casing of Snakecase Strings

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The below Liquid logic splits strings with the `'_'` delimiter and capitalizes each word in the returned split:

```
constant: proper_case_snakecase {
  value: "{% assign words = value | split: '_' %}"}
```

```

    {% for word in words %}
    {% assign down_word = word | downcase %}
    {% if down_word == 'gcs' %}
    {{word | upcase}}
    {% elsif down_word == 'ssn' %}
    {{word | upcase}}
    {% elsif down_word contains 'https' %}
    {{word | downcase }}
    {% elsif down_word contains '@' %}
    {{word | downcase }}
    {% elsif down_word == 'n/a' %}
    {{word | upcase }}
    {%else %}
    {{ word | capitalize }}
    {% endif %}
    {% endfor %}"
}

```

SQL Date Expressions

[Return to TOC](#)

A number of date logic subqueries are stored and utilized from the manifest file. These were instantiated within the manifest file due to the reusability throughout the various dashboards.

Below tackles each date logic subquery and explains the reasoning behind its implementation:

Yesterday SQL

[Return to TOC](#)

The below constant houses SQL that grabs yesterday's date:

```

constant: yesterday_sql {
  value: "date_add(current_date, interval -1 day)"
}

```

Yesterday Prior Year SQL

[Return to TOC](#)

The below constant houses SQL that grabs yesterday's date in the prior year:

```

constant: yesterday_prior_year {
  value: "date(extract(year from @{yesterday_sql}) - 1, extract(month from
@{yesterday_sql}), extract(day from @{yesterday_sql}))"
}

```

Yesterday Same Day Prior Year SQL

[Return to TOC](#)

The below constant houses SQL that grabs the date of the same day of yesterday in the prior year. This is not the same date last year; this is the same day of the week ~365 days prior. Thus, this would not be 10/01/2021 for a comparison with 10/01/2022; the correct value would be 10/02/2021, since both 10/02/2021 and 10/01/2022 both fall on Saturdays.

Also note: the following SQL takes into account if the current or compared-to year are leap years.

```
constant: yesterday_same_day_prior_year {
  value: "case
    when extract(day from (last_day(date(extract(year from
      @{{yesterday_sql}}),
      2, 1)))) = 29
    then date_add(date(extract(year from @{{yesterday_sql}}) - 1,
      extract(month from @{{yesterday_sql}}),
      extract(day from @{{yesterday_sql}}), interval 2 day)
    else
      date_add(date(extract(year from @{{yesterday_sql}}) - 1,
        extract(month from @{{yesterday_sql}}),
        extract(day from @{{yesterday_sql}}), interval 1 day)
    end"
}
```

Beginning of Last 7 days SQL

[Return to TOC](#)

The below constant houses SQL that grabs the date from 7 days ago:

```
constant: beginning_of_last_7_days {
  value: "date_add(current_date, interval -7 day)"
}
```

Beginning of Last 7 Days Prior Year SQL

[Return to TOC](#)

The below constant houses SQL that grabs the date from 7 days ago in the prior year.

```
constant: beginning_of_last_7_days_prior_year {
  value: "date(extract(year from @{{beginning_of_last_7_days}}) - 1, extract(month from
    @{{beginning_of_last_7_days}}), extract(day from @{{beginning_of_last_7_days}})"
}
```

Beginning of Last 7 Days Same Day Prior Year SQL

[Return to TOC](#)

The below constant houses SQL that grabs the date of the same day of 7 days ago in the prior year. This is not the same date last year; this is the same day of the week ~365 days prior. Thus, this would not be 10/01/2021 for a comparison with 10/01/2022; the correct value would be 10/02/2021, since both 10/02/2021 and 10/01/2022 both fall on Saturdays.

Also note: the following SQL takes into account if the current or compared-to year are leap years.

```
constant: beginning_of_last_7_days_same_day_prior_year {
  value: "case
  when extract(day from (last_day(date(extract(year from
    @beginning_of_last_7_days)),
    2, 1)))) = 29
  then date_add(date(extract(year from @beginning_of_last_7_days)) - 1,
    extract(month from @beginning_of_last_7_days),
    extract(day from @beginning_of_last_7_days)), interval 2 day)
  else
  date_add(date(extract(year from @beginning_of_last_7_days)) - 1,
    extract(month from @beginning_of_last_7_days),
    extract(day from @beginning_of_last_7_days)), interval 1 day)
  end"
}
```

Beginning of Last 30 Days Prior Year SQL

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The below constant houses SQL that grabs the date from 30 days ago in the prior year.

```
constant: beginning_of_last_30_days {
  value: "date_add(current_date, interval -30 day)"
}
```

Beginning of Last 30 Days Prior Year SQL

[Return to TOC](#)

The below constant houses SQL that grabs the date from 30 days ago in the prior year.

```
constant: beginning_of_last_30_days_prior_year {
  value: "date(extract(year from @beginning_of_last_30_days)) - 1, extract(month from
    @beginning_of_last_30_days), extract(day from @beginning_of_last_30_days))"
}
```

Beginning of Last 30 Days Same Day Prior Year SQL

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The below constant houses SQL that grabs the date of the same day of 30 days ago in the prior year. This is not the same date last year; this is the same day of the week ~365 days prior. Thus, this would not be 10/01/2021 for a comparison with 10/01/2022; the correct value would be 10/02/2021, since both 10/02/2021 and 10/01/2022 both fall on Saturdays.

Also note: the following SQL takes into account if the current or compared-to year are leap years.

```
constant: beginning_of_last_30_days_same_day_prior_year {
  value: "case
  when extract(day from (last_day(date(extract(year from
    @beginning_of_last_30_days)),
```

```

2, 1)))) = 29
then date_add(date(extract(year from @beginning_of_last_30_days) - 1,
extract(month from @beginning_of_last_30_days),
extract(day from @beginning_of_last_30_days)), interval 2 day)
else
date_add(date(extract(year from @beginning_of_last_30_days) - 1,
extract(month from @beginning_of_last_30_days),
extract(day from @beginning_of_last_30_days)), interval 1 day)
end"
}

```

Beginning of Last 90 days SQL

[Return to TOC](#)

The below constant houses SQL that grabs the date from 90 days ago:

```

constant: beginning_of_last_90_days {
  value: "date_add(current_date, interval -90 day)"
}

```

Beginning of Last 90 Days Prior Year SQL

[Return to TOC](#)

The below constant houses SQL that grabs the date from 90 days ago in the prior year.

```

constant: beginning_of_last_90_days_prior_year {
  value: "date(extract(year from @beginning_of_last_90_days) - 1, extract(month from
@beginning_of_last_90_days), extract(day from @beginning_of_last_90_days))"
}

```

Beginning of Last 365 days SQL

[Return to TOC](#)

The below constant houses SQL that grabs the date from 365 days ago:

```

constant: beginning_of_last_365_days {
  value: "date_add(current_date, interval -365 day)"
}

```

Beginning of Last 365 Days Same Day Prior Year SQL

[Return to TOC](#)

The below constant houses SQL that grabs the date of the same day of 365 days ago in the prior year. This is not the same date last year; this is the same day of the week ~365 days prior. Thus, this would not be 10/01/2021 for a comparison with 10/01/2022; the correct value would be 10/02/2021, since both 10/02/2021 and 10/01/2022 both fall on Saturdays.

Also note: the following SQL takes into account if the current or compared-to year are leap years.

```
constant: beginning_of_last_365_days_same_day_prior_year {
  value: "case
    when extract(day from (last_day(date(extract(year from
      @beginning_of_last_365_days)),
      2, 1)))) = 29
    then date_add(date(extract(year from @beginning_of_last_365_days)) - 1,
      extract(month from @beginning_of_last_365_days),
      extract(day from @beginning_of_last_365_days)), interval 2 day)
    else
      date_add(date(extract(year from @beginning_of_last_365_days)) - 1,
        extract(month from @beginning_of_last_365_days),
        extract(day from @beginning_of_last_365_days)), interval 1 day)
    end"
}
```

Beginning of Last Year SQL

[Return to TOC](#)

The below constant houses SQL that grabs the date for the beginning of the year:

```
constant: beginning_of_last_year {
  value: "date(extract(year from (current_date())), 1, 1)"
}
```

Beginning of Last Year Prior Year SQL

[Return to TOC](#)

The below constant houses SQL that grabs the date for the beginning of last year.

```
constant: beginning_of_last_year_prior_year {
  value: "date(extract(year from @beginning_of_last_year)) - 1, 1, 1)"
}
```

Custom Dimensions

[Return to TOC](#)

In this section, any dimensions developed that fall outside what would be considered standard dimensions will be addressed. Thus, any dimensions provided by Client upon view creation/dimensions that perform modest calculations will not be addressed in this section.

Yesterday YesNo Dimension

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The following is a simple **yesno** dimension that returns **true** if an order date is the same date as yesterday's date. Note that it references the **yesterday_sql** SQL constant stored in the manifest.

This dimension is referenced throughout other fields in the LookML as a foundational building block.

```
dimension: yesterday {  
  hidden: yes  
  type: yesno  
  sql: case when ${order_date_at_tz_ny_date} = @{{yesterday_sql}}  
        then true  
        else false  
        end;;  
  group_label: "Relative to Today"  
}
```

Last 7 Days YesNo Dimension

[Return to TOC](#)

The following is a simple **yesno** dimension that returns **true** if an order date is between the last 7 days and yesterday's dates. Note that it references the **beginning_of_last_7_days** & **yesterday_sql** SQL constants stored in the manifest.

This dimension is referenced throughout other fields in the LookML as a foundational building block.

```
dimension: last_7_days {  
  hidden: yes  
  type: yesno  
  sql: case when ${order_date_at_tz_ny_date} between @{{beginning_of_last_7_days}} and  
        @{{yesterday_sql}}  
        then true  
        else false  
        end;;  
  group_label: "Relative to Today"  
}
```

Last 30 Days YesNo Dimension

[Return to TOC](#)

The following is a simple **yesno** dimension that returns **true** if an order date is between the last 30 days and yesterday's dates. Note that it references the **beginning_of_last_30_days** & **yesterday_sql** SQL constants stored in the manifest.

This dimension is referenced throughout other fields in the LookML as a foundational building block.

```
dimension: last_30_days {  
  hidden: yes  
  type: yesno  
  sql: case when ${order_date_at_tz_ny_date} between @{{beginning_of_last_30_days}} and  
        @{{yesterday_sql}}  
        then true  
        else false  
        end;;  
  group_label: "Relative to Today"  
}
```


Last 365 Days YesNo Dimension

[Return to TOC](#)

The following is a simple **yesno** dimension that returns **true** if an order date is between the last 365 days and yesterday's dates. Note that it references the **beginning_of_last_365_days** & **yesterday_sql** SQL constants stored in the manifest.

This dimension is referenced throughout other fields in the LookML as a foundational building block.

```
dimension: last_365_days {  
  hidden: yes  
  type: yesno  
  sql: case when ${order_date_at_tz_ny_date} between @{beginning_of_last_365_days} and  
    @{yesterday_sql}  
    then true  
    else false  
    end;;  
  group_label: "Relative to Today"  
}
```

Yesterday Prior Year YesNo Dimension

[Return to TOC](#)

The following is a simple **yesno** dimension that returns **true** if an order date is the same day as yesterday in the prior year's date. Note that it references the **yesterday_same_day_prior_year** SQL constant stored in the manifest.

This dimension is referenced throughout other fields in the LookML as a foundational building block.

```
dimension: yesterday_prior_year {  
  hidden: yes  
  type: yesno  
  sql: case when ${order_date_at_tz_ny_date} = @{yesterday_same_day_prior_year}  
    then true  
    else false  
    end;;  
  group_label: "Relative to Prior Year"  
}
```

Last 7 Days Prior Year YesNo Dimension

[Return to TOC](#)

The following is a simple **yesno** dimension that returns **true** if an order date is between the same day as 7 days ago in the prior year and yesterday's same day prior year dates. Note that it references the **beginning_of_last_7_days_same_day_prior_year** & **yesterday_same_day_prior_year** SQL constants stored in the manifest.

This dimension is referenced throughout other fields in the LookML as a foundational building block.

```
dimension: last_7_days_prior_year {
  hidden: yes
  type: yesno
  sql: case when ${order_date_at_tz_ny_date} between
    @{beginning_of_last_7_days_same_day_prior_year} and @{yesterday_same_day_prior_year}
    then true
    else false
    end;;
  group_label: "Relative to Prior Year"
}
```

Last 30 Days Prior Year YesNo Dimension

[Return to TOC](#)

The following is a simple **yesno** dimension that returns **true** if an order date is between the same day as 30 days ago in the prior year and yesterday's same day prior year dates. Note that it references the **beginning_of_last_30_days_same_day_prior_year** & **yesterday_same_day_prior_year** SQL constants stored in the manifest.

This dimension is referenced throughout other fields in the LookML as a foundational building block.

```
dimension: last_30_days_prior_year {
  hidden: yes
  type: yesno
  sql: case when ${order_date_at_tz_ny_date} between
    @{beginning_of_last_30_days_same_day_prior_year} and @{yesterday_same_day_prior_year}
    then true
    else false
    end;;
  group_label: "Relative to Prior Year"
}
```

Last 365 Days Prior Year YesNo Dimension

[Return to TOC](#)

The following is a simple **yesno** dimension that returns **true** if an order date is between the same day as 365 days ago in the prior year and yesterday's same day prior year dates. Note that it references the **beginning_of_last_365_days_same_day_prior_year** & **yesterday_same_day_prior_year** SQL constants stored in the manifest. This dimension is referenced throughout other fields in the LookML as a foundational building block.

```
dimension: last_365_days_prior_year {
  hidden: yes
  type: yesno
  sql: case when ${order_date_at_tz_ny_date} between
    @{beginning_of_last_365_days_same_day_prior_year} and @{yesterday_same_day_prior_year}
    then true
    else false
    end;;
  group_label: "Relative to Prior Year"
}
```

Order Pubkey (Yesterday) String Dimension

[Return to TOC](#)

The following string dimension isolates order pubkeys that involve orders placed yesterday. This is integral to allowing count distinct integrity across various time binnings in the Daily Flash dashboard.

```
dimension: order_pubkey_yesterday {  
  hidden: yes  
  type: string  
  sql: case when ${yesterday}  
    then ${TABLE}.order_pubkey  
    else null  
    end;;  
  group_label: "ID Fields"  
}
```

Order Pubkey (Last 7 Days) String Dimension

[Return to TOC](#)

The following string dimension isolates order pubkeys that involve orders placed within the last 7 days. This is integral to allowing count distinct integrity across various time binnings in the Daily Flash dashboard.

```
dimension: order_pubkey_last_7_days {  
  hidden: yes  
  type: string  
  sql: case when ${yesterday} or ${last_7_days}  
    then ${TABLE}.order_pubkey  
    else null  
    end;;  
  group_label: "ID Fields"  
}
```

Order Pubkey (Last 30 Days) String Dimension

[Return to TOC](#)

The following string dimension isolates order pubkeys that involve orders placed within the last 30 days. This is integral to allowing count distinct integrity across various time binnings in the Daily Flash dashboard.

```
dimension: order_pubkey_last_30_days {  
  hidden: yes  
  type: string  
  sql: case when ${yesterday} or ${last_7_days} or ${last_30_days}  
    then ${TABLE}.order_pubkey  
    else null  
    end;;  
  group_label: "ID Fields"  
}
```

Order Pubkey (Last 365 Days) String Dimension

[Return to TOC](#)

The following string dimension isolates order pubkeys that involve orders placed within the last 365 days. This is integral to allowing count distinct integrity across various time binnings in the Daily Flash dashboard.

```
dimension: order_pubkey_last_365_days {
  hidden: yes
  type: string
  sql: case when ${yesterday} or ${last_7_days} or ${last_30_days} or ${last_365_days}
        then ${TABLE}.order_pubkey
        else null
        end;;
  group_label: "ID Fields"
}
```

Customer Pubkey (Yesterday) String Dimension

[Return to TOC](#)

The following string dimension isolates customer pubkeys that involve orders placed yesterday. This is integral to allowing count distinct integrity across various time binnings in the Daily Flash dashboard.

```
dimension: customer_pubkey_yesterday {
  hidden: yes
  type: string
  sql: case when ${yesterday}
        then ${TABLE}.customer_pubkey
        else null
        end;;
  group_label: "ID Fields"
}
```

Customer Pubkey (Last 7 Days) String Dimension

[Return to TOC](#)

The following string dimension isolates customer pubkeys that involve orders placed within the last 7 days. This is integral to allowing count distinct integrity across various time binnings in the Daily Flash dashboard.

```
dimension: customer_pubkey_last_7_days {
  hidden: yes
  type: string
  sql: case when ${yesterday} or ${last_7_days}
        then ${TABLE}.customer_pubkey
        else null
        end;;
  group_label: "ID Fields"
}
```

Customer Pubkey (Last 30 Days) String Dimension

[Return to TOC](#)

The following string dimension isolates customer pubkeys that involve orders placed within the last 30 days. This is integral to allowing count distinct integrity across various time binnings in the Daily Flash dashboard.

```
dimension: customer_pubkey_last_30_days {
  hidden: yes
  type: string
  sql: case when ${yesterday} or ${last_7_days} or ${last_30_days}
    then ${TABLE}.customer_pubkey
    else null
    end;;
  group_label: "ID Fields"
}
```

Customer Pubkey (Last 365 Days) String Dimension

[Return to TOC](#)

The following string dimension isolates customer pubkeys that involve orders placed within the last 365 days. This is integral to allowing count distinct integrity across various time binnings in the Daily Flash dashboard.

```
dimension: customer_pubkey_last_365_days {
  hidden: yes
  type: string
  sql: case when ${yesterday} or ${last_7_days} or ${last_30_days} or ${last_365_days}
    then ${TABLE}.customer_pubkey
    else null
    end;;
  group_label: "ID Fields"
}
```

Order Date (Yesterday) String Dimension

[Return to TOC](#)

The following is a simple **string** dimension that returns **order_date_at_tz_ny_date** if an order date is the same as yesterday's date. Note that it uses the **yesterday yesno** dimension for logic.

```
dimension: order_date_yesterday {
  hidden: yes
  type: string
  sql: case when ${yesterday}
    then ${order_date_at_tz_ny_date}
    else null
    end;;
  group_label: "Daily Flash Calculation"
}
```

Order Date (Last 7 Days) String Dimension

[Return to TOC](#)

The following is a simple **string** dimension that returns **order_date_at_tz_ny_date** if an order date is between the last 7 days and yesterday's dates. Note that it uses the **last_7_days** & **yesterday yesno** dimensions for logic.

```
dimension: order_date_last_7_days {  
  hidden: yes  
  type: string  
  sql: case when ${yesterday} or ${last_7_days}  
    then ${order_date_at_tz_ny_date}  
    else null  
    end;;  
  group_label: "Daily Flash Calculation"  
}
```

Order Date (Last 30 Days) String Dimension

[Return to TOC](#)

The following is a simple **string** dimension that returns **order_date_at_tz_ny_date** if an order date is between the last 30 days and yesterday's dates. Note that it uses the **last_30_days**, **last_7_days** & **yesterday yesno** dimensions for logic.

```
dimension: order_date_last_30_days {  
  hidden: yes  
  type: string  
  sql: case when ${yesterday} or ${last_7_days} or ${last_30_days}  
    then ${order_date_at_tz_ny_date}  
    else null  
    end;;  
  group_label: "Daily Flash Calculation"  
}
```

Order Date (Last 365 Days) String Dimension

[Return to TOC](#)

The following is a simple **string** dimension that returns **order_date_at_tz_ny_date** if an order date is between the last 365 days and yesterday's dates. Note that it uses the **last_365_days**, **last_30_days**, **last_7_days** & **yesterday yesno** dimensions for logic.

```
dimension: order_date_last_365_days {  
  hidden: yes  
  type: string  
  sql: case when ${yesterday} or ${last_7_days} or ${last_30_days} or ${last_365_days}  
    then ${order_date_at_tz_ny_date}  
    else null  
    end;;  
  group_label: "Daily Flash Calculation"  
}
```

Order Date (Yesterday Prior Year) String Dimension

[Return to TOC](#)

The following is a simple **string** dimension that returns **order_date_at_tz_ny_date** if an order date as the same yesterday prior year's date. Note that it uses the **yesterday_prior_year yesno** dimension for logic.

```
dimension: order_date_yesterday_prior_year {
  hidden: yes
  type: string
  sql: case when ${yesterday_prior_year}
    then ${order_date_at_tz_ny_date}
    else null
    end;;
  group_label: "Daily Flash Calculation"
}
```

Order Date (Last 7 Days Prior Year) String Dimension

[Return to TOC](#)

The following is a simple **string** dimension that returns **order_date_at_tz_ny_date** if an order date is between the last 7 days in the prior year and yesterday in the prior year's dates. Note that it uses the **last_7_days_prior_year** & **yesterday_prior_year yesno** dimensions for logic.

```
dimension: order_date_last_7_days_prior_year {
  hidden: yes
  type: string
  sql: case when ${yesterday_prior_year} or ${last_7_days_prior_year}
    then ${order_date_at_tz_ny_date}
    else null
    end;;
  group_label: "Daily Flash Calculation"
}
```

Order Date (Last 30 Days Prior Year) String Dimension

[Return to TOC](#)

The following is a simple **string** dimension that returns **order_date_at_tz_ny_date** if an order date is between the last 30 days in the prior year and yesterday in the prior year's dates. Note that it uses the **last_30_days_prior_year**, **last_7_days_prior_year** & **yesterday_prior_year yesno** dimensions for logic.

```
dimension: order_date_last_30_days_prior_year {
  hidden: yes
  type: string
  sql: case when ${yesterday_prior_year} or ${last_7_days_prior_year} or
    ${last_30_days_prior_year}
    then ${order_date_at_tz_ny_date}
    else null
  end;;
  group_label: "Daily Flash Calculation"
}
```

```

        end;;
    group_label: "Daily Flash Calculation"
}

```

Order Date (Last 365 Days Prior Year) String Dimension

[Return to TOC](#)

The following is a simple **string** dimension that returns **order_date_at_tz_ny_date** if an order date is between the last 365 days in the prior year and yesterday in the prior year's dates. Note that it uses the **last_365_days_prior_year**, **last_30_days_prior_year**, **last_7_days_prior_year** & **yesterday_prior_year yesno** dimensions for logic.

```

dimension: order_date_last_365_days_prior_year {
    hidden: yes
    type: string
    sql: case when ${yesterday_prior_year} or ${last_7_days_prior_year} or
    ${last_30_days_prior_year} or ${last_365_days_prior_year}
        then ${order_date_at_tz_ny_date}
        else null
        end;;
    group_label: "Daily Flash Calculation"
}

```

Daily Flash Times String Dimesion

[Return to TOC](#)

The following dimension bins time according to date criteria, returning a bin string if an order date meets the corresponding criteria. This dimension is used in the KPI Performance over Time Look on the Daily Flash tab of the Company Report.

Note that this dimension is ordered by the **flash_rank** dimension described directly after this dimension.

```

dimension: daily_flash_times {
    description: "Buckets for different timeframes to compare by in the Daily Flash dashboard"
    hidden: yes
    sql: case when ${order_date_yesterday} is not null
        then 'Yesterday'
        when ${order_date_last_7_days} is not null
        then 'Last 7 Days'
        when ${order_date_last_30_days} is not null
        then 'Last 30 Days'
        when ${order_date_last_365_days} is not null
        then 'Last 365 Days'
        else null
        end;;
    order_by_field: flash_rank
    group_label: "Daily Flash Calculation"
}

```

Flash Rank Number Dimesion

[Return to TOC](#)

This dimension ensures chronological order sorting of the `daily_flash_times` dimension. This dimension recasts the logic in `daily_flash_times` into a ranking of integers.

```
dimension: flash_rank {
  description: "Hidden field used to rank the daily_flash_times dimension
chronologically."
  hidden: yes
  sql: case when ${daily_flash_times} = 'Yesterday'
    then 1
    when ${daily_flash_times} = 'Last 7 Days'
    then 2
    when ${daily_flash_times} = 'Last 30 Days'
    then 3
    when ${daily_flash_times} = 'Last 365 Days'
    then 4
    when ${daily_flash_times} is null
    then 5
    end;;
}
```

Current vs Previous Period Dynamic Dimension[Return to TOC](#)

This dynamic dimension employs a mixture of Liquid logic and SQL logic. This dimension is dependent upon the selection of the `select_timeframe` parameter, which branches the Liquid logic accordingly. Within each Liquid block is a different `case when` criterion addressing order date logic.

This dimension returns either 'Current Year' or 'Prior Year' after the logic has been evaluated. This is a dynamic dimension; any measure stacked against it will be cast to compare the current vs prior year metrics.

```
dimension: current_vs_previous_period {
  description: "Use this dimension along with the Period over Period Scope Filter for dynamic
Period of Period analysis"
  hidden: yes
  type: string
  sql:
    case
      {% if select_timeframe._parameter_value == "'Last 7 Days'" %}
        when ${order_date_at_tz_ny_date}
          between @{{beginning_of_last_7_days}} and @{{yesterday_sql}}
      {% elsif select_timeframe._parameter_value == "'Last 30 Days'" %}
        when ${order_date_at_tz_ny_date}
          between @{{beginning_of_last_30_days}} and @{{yesterday_sql}}
      {% elsif select_timeframe._parameter_value == "'Last 90 Days'" %}
        when ${order_date_at_tz_ny_date}
          between @{{beginning_of_last_90_days}} and @{{yesterday_sql}}
      {% elsif select_timeframe._parameter_value == "'YTD'" %}
        when ${order_date_at_tz_ny_date}
          between @{{beginning_of_last_year}} and @{{yesterday_sql}}
      {% endif %}
      then 'Current Year'
    {% if select_timeframe._parameter_value == "'Last 7 Days'" %}
```

```

        when ${order_date_at_tz_ny_date}
        between @{{beginning_of_last_7_days_prior_year}} and @{{yesterday_prior_year}}
    {% elsif select_timeframe._parameter_value == "'Last 30 Days'" %}
        when ${order_date_at_tz_ny_date}
        between @{{beginning_of_last_30_days_prior_year}} and @{{yesterday_prior_year}}
    {% elsif select_timeframe._parameter_value == "'Last 90 Days'" %}
        when ${order_date_at_tz_ny_date}
        between @{{beginning_of_last_90_days_prior_year}} and @{{yesterday_prior_year}}
    {% elsif select_timeframe._parameter_value == "'YTD'" %}
        when ${order_date_at_tz_ny_date}
        between @{{beginning_of_last_year_prior_year}} and @{{yesterday_prior_year}}
    {% endif %}
    then 'Prior Year'
    else null
end
;;
}

```

Selected Dynamic Duration Number Dynamic Dimension

[Return to TOC](#)

This dynamic dimension is Liquid dependent. The key evaluator is the `select_timeframe` parameter value. Depending on the value, a different sorting index is returned to aide in sorting.

```

dimension: selected_dynamic_duration {
  description: "Dynamic date column for Period over Period analysis"
  hidden: yes
  type: number
  sql:
    {% if select_timeframe._parameter_value == "'Last 7 Days'" %}
      ${order_date_at_tz_ny_day_of_week_index}
    {% elsif select_timeframe._parameter_value == "'Last 30 Days'" %}
      ${order_date_at_tz_ny_month_num}
    {% elsif select_timeframe._parameter_value == "'Last 90 Days'" %}
      ${order_date_at_tz_ny_month_num}
    {% elsif select_timeframe._parameter_value == "'YTD'" %}
      ${order_date_at_tz_ny_month_num}
    {% endif %}
  ;;
}

```

Selected Dynamic Day Of String Dynamic Dimension

[Return to TOC](#)

This dynamic dimension is Liquid dependent. The key evaluator is the `select_timeframe` parameter value. Depending on the value, a different date formatting is returned. This formatting is what is rendered on the x-axis of dynamic trend visualizations.

Note in the formatting that year is formatted out. This is to allow comparison between current and previous periods simultaneously.

```

dimension: selected_dynamic_day_of {
  label: "Trend Line Date"
  description: "Dynamic date formatting column for Period over Period analysis"
  hidden: yes
  order_by_field: selected_dynamic_day_of_sort
  type: string
  sql:
    {% if select_timeframe._parameter_value == "'Last 7 Days'" %}
      format_date("%m/%d", ${order_date_at_tz_ny_date})
    {% elsif select_timeframe._parameter_value == "'Last 30 Days'" %}
      format_date("%m/%d", ${order_date_at_tz_ny_date})
    {% elsif select_timeframe._parameter_value == "'Last 90 Days'" %}
      format_date("%m/%d", ${order_date_at_tz_ny_date})
    {% elsif select_timeframe._parameter_value == "'YTD'" %}
      ${order_date_at_tz_ny_month_name}
    {% endif %}
  ;;
}

```

Selected Dynamic Day Of Sort String Dynamic Dimension

[Return to TOC](#)

This dynamic dimension is Liquid dependent. The key evaluator is the `select_timeframe` parameter value. Depending on the value, a different date formatting is returned. This formatting is used to sort values chronologically.

Note in the formatting that year is formatted out. This is to allow comparison between current and previous periods simultaneously.

```

dimension: selected_dynamic_day_of_sort {
  description: "Sorting column to dynamically sort Period over Period analysis in chronological order"
  hidden: yes
  type: string
  sql:
    {% if select_timeframe._parameter_value == "'Last 7 Days'" %}
      format_date("%m/%d", ${order_date_at_tz_ny_date})
    {% elsif select_timeframe._parameter_value == "'Last 30 Days'" %}
      format_date("%m/%d", ${order_date_at_tz_ny_date})
    {% elsif select_timeframe._parameter_value == "'Last 90 Days'" %}
      format_date("%m/%d", ${order_date_at_tz_ny_date})
    {% elsif select_timeframe._parameter_value == "'YTD'" %}
      format_date("%m", ${order_date_at_tz_ny_date})
    {% endif %};;
}

```

Parameters

[Return to TOC](#)

`select_timeframe` | `order` | A parameter with values of 'Last 7 Days', 'Last 30 Days', 'Last 90 Days' and 'YTD'. This parameter is manipulated when buttons are clicked on the dashboard to change the date range and date formatting.

Custom Measures

[Return to TOC](#)

| `total_number_orders_daily_flash` | `order` | `count_distinct` measure with a `case when` statement in the `sql` parameter. In this `case when` statement, it evaluates which bin type `daily_flash_times` is returning, then pulling the corresponding `order_pubkey` dimension to perform a count distinct. Thus, if `daily_flash_times` = 'Last 30 Days' then `order_pubkey_last_30_days` is returned. || `total_number_orders_yesterday_py`
`total_number_orders_last_7_days_py`
`total_number_orders_last_30_days_py`
`total_number_orders_last_365_days_py` | `order` | `count_distinct` measures that calculate the prior year values of total orders. These had to be separated into separate measures because the primary dates were already assumed in the `daily_flash_times`; creating a `daily_flash_times_prior_year` would double the crosstab size and not allow side-by-side comparison. Thus, separate, filtered measures were created, later being aggregated via a table calculation in the final Look. || `distinct_new_customer_count_daily_flash` | `order` | Same logic for `total_number_orders_daily_flash`, different pubkey returns (`customer_pubkey` return instead of `order_pubkey`). || `distinct_new_customer_count_yesterday_py`
`distinct_new_customer_count_last_7_days_py`
`distinct_new_customer_count_last_30_days_py`
`distinct_new_customer_count_last_365_days_py` | `order` | Same logic for `total_number_orders_[time_bin].py` measures above, except calculating the total number of new distinct customers instead of distinct order counts. || `total_net_sales_USD_running_total` | `order` | A simple running total upon `total_net_sales_USD`. This cares for the binary classification issue originally grappled with the `daily_flash_times`. || `total_net_sales_USD_yesterday_py`
`total_net_sales_USD_last_7_days_py`
`total_net_sales_USD_last_30_days_py`
`total_net_sales_USD_last_365_days_py` | `order` | `sum_distinct` measures that calculate the prior year values of total net sales. These had to be separated into separate measures because the primary dates were already assumed in the `daily_flash_times`; creating a `daily_flash_times_prior_year` would double the crosstab size and not allow side-by-side comparison. Thus, separate, filtered measures were created, later being aggregated via a table calculation in the final Look. || `total_media_spend_running_total` | `marketing_spend` | A simple running total upon `total_media_spend`. This cares for the binary classification issue originally grappled with the `daily_flash_times`. || `total_media_spend_yesterday_py`
`total_media_spend_last_7_days_py`
`total_media_spend_last_30_days_py`
`total_media_spend_last_365_days_py` | `marketing_spend` | `sum` measures that calculate the prior year values of total marketing spend. These had to be separated into separate measures because the primary dates were already assumed in the `daily_flash_times`; creating a `daily_flash_times_prior_year` would double the crosstab size and not allow side-by-side comparison. Thus, separate, filtered measures were created, later being aggregated via a table calculation in the final Look. || `cpa_yesterday_py`
`cpa_last_7_days_py`
`cpa_last_30_days_py`
`cpa_last_365_days_py` | `marketing_spend` | Measures that calculate the prior year values of CPA. These had to be separated into separate measures because the primary dates were already assumed in the `daily_flash_times`; creating a `daily_flash_times_prior_year` would double the crosstab size and not allow side-by-side comparison. Thus, separate, filtered measures were created, later being aggregated via a table calculation in the final Look. |

Product Launch Components Reference

[Return to TOC](#)

Launch Date

[return](#)

```

view: drv_launch_date_by_product {
  derived_table: {
    sql:
      select distinct product_name, product_target_sub_category, target_sub_category,
min(order_date_at_tz_ny) over (partition by product_name order by product_name) as
launch_date FROM
`Clienthair.Client_bi_datamart_growth_customer_ltv_customer_order.order_items`
      where item_type ='formula' ;;
  }
  dimension: product_name{
    primary_key: yes
    type: string
    sql: ${TABLE}.product_name ;;
  }
  dimension: target_sub_category {
    type: string
    sql: ${TABLE}.target_sub_category ;;
  }
  dimension: product_target_sub_category {
    type: string
    sql: ${TABLE}.product_target_sub_category ;;
  }
  dimension_group: launch_date {
    type: time
    timeframes: [
      raw,
      date,
      day_of_week,
      day_of_week_index,
      week,
      month,
      month_name,
      month_num,
      quarter,
      year
    ]
    convert_tz: no
    datatype: date
    sql: ${TABLE}.launch_date ;;
  }
}

```

Launch

return

```

dimension_group: launch {
  type: duration
  sql_start: ${drv_launch_date_by_product.launch_date_raw} ;;
  sql_end: ${order_date_at_tz_ny_raw} ;;
}

```

Select Days Since Launch

return

```
parameter: select_days_since_launch {
  hidden: no
  type: string
  default_value: "< 31"
  allowed_value: {
    label: "30 days"
    value: "< 31"
  }
  allowed_value: {
    label: "60 Days"
    value: "< 61"
  }
  allowed_value: {
    label: "90 Days"
    value: "< 91"
  }
  allowed_value: {
    label: "1 yr"
    value: "< 366"
  }
  allowed_value: {
    label: "no limit"
    value: "< 99999"
  }
}
```

Selected Dynamic Launch Days Label

return

```
dimension: selected_dynamic_launch_days_label {
  description: "Dynamic Number of Launch days grouping"
  label: "Days Since Launch"
  type: number
  sql:
    {% if select_days_since_launch._parameter_value == "< 31" %}
      ${days_launch}
    {% elsif select_days_since_launch._parameter_value == "< 61" %}
      ${days_launch}
    {% elsif select_days_since_launch._parameter_value == "< 91" %}
      DIV(${days_launch},7)* 7
    {% elsif select_days_since_launch._parameter_value == "< 366" %}
      DIV(${days_launch},7)* 7
    {% elsif select_days_since_launch._parameter_value == "< 99999" %}
      DIV(${days_launch},30)* 30
    {% else %}
      DIV(${days_launch},7)* 7
    {% endif %}
    ;;
}
```

Selected Dynamic Launch Days Duration

[return](#)

```
dimension: selected_dynamic_launch_days_duration {
  description: "Dynamic Number of Launch days grouping"
  type: number
  sql:
    {% if select_days_since_launch._parameter_value == "'< 31'" %}
      31
    {% elsif select_days_since_launch._parameter_value == "'< 61'" %}
      61
    {% elsif select_days_since_launch._parameter_value == "'< 91'" %}
      91
    {% elsif select_days_since_launch._parameter_value == "'< 366'" %}
      366
    {% elsif select_days_since_launch._parameter_value == "'< 99999'" %}
      9999
    {% else %}
      31
    {% endif %}
  ;;
}
```

Custom Filter

[return](#custom filter_ref)

```
${order_items.days_launch}<${order_items.selected_dynamic_launch_days_duration}
```

Finance Cohort Components Reference

[Return to TOC](#)

Cohort Finance Name

[return](#Cohort Finance Name_ref)

```
view: cohort_finance_name {
  derived_table: {
    sql:
      Select "repeat_sub_at_first_orders_non_sub_supplements" as cohort_finance_name
      union all
      Select "repeat_sub_at_first_orders_sub_haircare" as cohort_finance_name
      union all
      Select "repeat_sub_at_first_orders_non_sub_haircare" as cohort_finance_name
      union all
      Select "repeat_sub_at_first_orders_sub_supplements" as cohort_finance_name
      union all
      Select "repeat_sub_at_repeat_churned_customers" as cohort_finance_name
      union all
      Select "repeat_sub_at_first_churned_customers" as cohort_finance_name
      union all
      Select "new_non_sub" as cohort_finance_name
```

```

    union all
    Select "repeat_sub_at_repeat_orders_first_only" as cohort_finance_name
    union all
    Select "repeat_non_sub_orders" as cohort_finance_name
    union all
    Select "repeat_sub_at_repeat_orders_by_sub_cohort" as cohort_finance_name
    union all
    Select "repeat_sub_at_repeat_orders_by_sub_cohort_supplements" as cohort_finance_name
    union all
    Select "repeat_sub_at_repeat_orders_by_sub_cohort_haircare" as cohort_finance_name
    union all
    Select "repeat_sub_at_first_orders_non_sub" as cohort_finance_name
    union all
    Select "repeat_sub_at_first_orders_sub" as cohort_finance_name
    union all
    Select "repeat_sub_at_repeat_orders_by_first_order_cohort" as cohort_finance_name
    union all
    Select "new_sub_at_first" as cohort_finance_name
  ;;
}
dimension: cohort_finance_name {
  primary_key: yes
  type: string
  sql: ${TABLE}.cohort_finance_name ;;
  html: @proper_case_snakecase ;;
}
}

```

Cohort Finance Units

[return](#Cohort Finance Units_ref)

```

measure: cohort_finance_units {
  type: number
  sql: CASE
    WHEN ${cohort_finance_name} = "new_non_sub"
    THEN ${new_non_subscription_at_first_units}

    WHEN ${cohort_finance_name} = "new_sub_at_first"
    THEN ${new_subscription_at_first_units}

    WHEN ${cohort_finance_name} = "repeat_non_sub_orders"
    THEN ${repeat_nonsub_units}

    WHEN ${cohort_finance_name} = "repeat_sub_at_repeat_orders_by_first_order_cohort"
    --same as above, grouping different on report??
    THEN ${repeat_subscription_units}

    WHEN ${cohort_finance_name} = "repeat_sub_at_repeat_orders_first_only"
    THEN ${repeat_sub_at_first_units}

    WHEN ${cohort_finance_name} = "repeat_sub_at_first_orders_non_sub"
    THEN ${repeat_sub_at_first_orders_non_sub_units}

    WHEN ${cohort_finance_name} = "repeat_sub_at_first_orders_sub"
    THEN ${repeat_sub_at_first_sub_units}
  
```



```
        WHEN ${cohort_finance_name} = "repeat_sub_at_repeat_churned_customers" -- best
guess on this - need confirmation
        THEN ${repeat_subscription_units_churned}

        WHEN ${cohort_finance_name} = "repeat_sub_at_first_churned_customers"
        THEN ${repeat_sub_at_first_sub_units_churned}

        else
        0 end ;;
    }
```

Explores

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Explores used in the application are identified below along with their component views with additional commentary where applicable.

Explore Name: name	Model File	Notes
Explore #1 customer: Contained View File #1 customer	Test_model	Customer view, no joins
Explore #2 customer_subscriber: Contained View File #1 customer_subscriber	Test_model	Customer subscriber view, no joins
Explore #3 order: Contained View File #1 order Contained View File #2 customer	Test_model	Orders & Customers
Explore #4 marketing_spend_and_orders: Contained View File #1 order Contained View File #2 customer Contained View File #3 marketing_spend	Test_model	Orders & Customers
Explore #5 order_items: Contained View File #1 order_items Contained View File #2 customer Contained View File #3 order Contained View File #4 drv_launch_date_by_produc	Test_model	Created to support product level reporting by days since launch parameter

Explore Name: name	Model File	Notes
Explore #6 last_synced: Contained View File #1 order		
Contained View File #2 customer		
Contained View File #3 marketing_spend	Test_model	Obtains last synced date from respective views
Contained View File #4 order_items		
Explore #7 finance_cohort: Contained View File #1 order_cohort_extensions		
Contained View File #2 customer	Test_model	Created for Finance Cohorts reporting, supports reporting against derived cohort finance name dimension
Contained View File #3 cohort_finance_name		
Explore #8 finance_cohort_products: Contained View File #1 order_items_cohort_extension		
Contained View File #2 customer	Test_model	Created for Finance Cohort product units reporting
Contained View File #3 cohort_finance_name		
Contained View File #4 drv_launch_date_by_product		
Explore #9 drv_pdt_daily_flash:	Test_model	Single view explore on pdt for daily flash table visual
Explore #10 subscription:	Test_model	Single view explore on subscription view

View Files

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The views required for the application are identified below, with a brief description of purpose and identification of referenced BigQuery tables.

The views are categorized as follows: -DB Table/view -Derived (Native) Table -Derived (SQL) Table -Extension

The lone persistent derived table, is identified as such, along with it's triggering datagroup. LookML objects with embedded logic in the SQL definition or LookML objects with unique parameters are listed with a brief description.

View File #1 customer (DB Table/view)

Only minor changes were required in the customer view, consisting of pulling in a couple of additional flags and correctinng a misspeling on leavin_conditioner, changed to leavein_conditioner.

BigQuery Tables Referenced: Clienthair.Client_bi_datamart_growth_customer_ltv_customer_order.customer

View File #2 customer_subscriber (Derived (SQL) Table)

This view is not used in the application.

BigQuery Tables Referenced:

Clienthair.Client_bi_datamart_growth_customer_ltv_customer_order.order_items

View File #3 marketing_spend (DB Table/view)

This view captures media spend by country, platform, channel group, and calendar date in which the media dollars were spent / recorded. Filtered measures are used to "bin" prior year measures into 4 time period buckets (yesterday_py, last_7_days_prior_year, last_30_days_prior_year, and last_365_days_prior_year).)

BigQuery Tables Referenced: Clienthair.Client_bi_datamart_marketing_spend.marketing_spend

View File #4 order (DB Table/view)

The order view serves as one of the primary fact tables for the application, with heavy dependency on customer attributes from the customer view which is a required join to this view. A number of dimensions and measures were created to support "binning" of results by This year/last year grouping in to time period buckets of yesterday, last 7 days, last 30 days and last 365 days. In addition to the basic measure computation, dimensions were created to support dynamic labeling of the time period axis.

BigQuery Tables Referenced: Clienthair.Client_bi_datamart_growth_customer_ltv_customer_order.order

View File #5 order_items (DB Table/view)

This view pulls data from the order items data mart. There are many hidden fields within this view. This is to support efforts to provide visibility into rollup aggregation bucketed by timeframes.

BigQuery Tables Referenced:

Clienthair.Client_bi_datamart_growth_customer_ltv_customer_order.order_items

View File #6 subscription (DB Table/view)

No changes were made to this view

BigQuery Tables Referenced: Clienthair.Client_bi_datamart_subscription.subscription

View File #7 cohort_finance_name (Derived (SQL) Table)

This view is used to generate a list of 16 cohort_finance_names as identified in the requirements from Finance. Corresponding filtered measures are mapped to these names via case statements in the order_cohort_extension and order_items_cohort_extension views.

BigQuery Tables Referenced: N/A

View File #8 order_cohort_extension (Extension)

This view extends the order view, adding cohort_finance_name from the required join to the cohort_finance_names view required to accomplish the cohort_name mapping to the appropriate measure. The measures themselves are redefined in this view, as sum_distinct is required as measure type due to the required cross join to cohort_finance_name.

BigQuery Tables Referenced: Clienthair.Client_bi_datamart_growth_customer_ltv_customer_order.order

View File #9 order_items_cohort_extension (Extension)

This view extends the order_items view, adding cohort_finance_name from the required join to the cohort_finance_names view required to accomplish the cohort_name mapping to the appropriate measure. Measure duplication was not required for this view in that the measures are limited to type count_distinct.

BigQuery Tables Referenced:

`Clienthair.Client_bi_datamart_growth_customer_ltv_customer_order.order_items`

View File #10 drv_launch_date_by_product (Derived (SQL) Table)

This view was required to get a distinct launch date (min(order_date_at_tz_ny)) by product, which was used in launch duration dimension in order_items. The duration dimension is a filter in the Product Launch dashboard. This construction requires that drv_launch_date_by_product be joined in explores referencing order_items.

BigQuery Tables Referenced:

`Clienthair.Client_bi_datamart_growth_customer_ltv_customer_order.order_items`

View File #11 drv_pdt_daily_flash (Derived (Native) Table)

This persistent derived table was constructed to address performance issues with the daily flash table visual on the dashboard of the same name. The base lookml was generated from the non-pdt version of the visual, with additional measures added to support calculation of ratio (i.e. non-aggregatable) measures in the view post aggregation.

Explore Source: `marketing_spend_and_orders`

Persistent Derived Table with datagroup_trigger: datagroup_orders [Return to TOC](#)