

E-commerce Analysis: Final Scripts & Documentation

CMO / CX

Category 1: CMO (Chief Marketing Officer) Analysis

Goal:

To measure the effectiveness of marketing channels, understand true customer value, and track loyalty and retention. These queries analyze acquisition, customer lifetime value (CLV), and cohort behavior to guide strategic marketing spend.

Query 1.1: Channel Acquisition Summary

Description:

This view provides a high-level summary of acquisition performance by traffic source. It counts the total unique sessions, the total orders generated (by counting unique sessions with a 'purchase' event), and segments this traffic into "Guest" vs. "Logged-in" sessions. It's the first step in identifying which channels drive high-volume traffic versus high-value, purchasing traffic.

```
24      -- CMO ANALYSIS -----
25      -- Acquisition Analysis -----
26 •  SELECT
27          traffic_source,
28          COUNT(DISTINCT session_id) AS total_sessions,
29          COUNT(CASE
30              WHEN event_type = 'purchase' THEN event_type
31              ELSE NULL
32          END) total_orders,
33          COUNT(DISTINCT CASE
34              WHEN user_id IS NOT NULL THEN session_id
35              END) AS logged_in_customers,
36          COUNT(DISTINCT CASE
37              WHEN user_id IS NULL THEN session_id
38              END) AS guest_sessions
39      FROM
40          events
41      GROUP BY traffic_source;
42
```

Result Grid Filter Rows: Export: Wrap Cell Content:					
	traffic_source	total_sessions	total_orders	logged_in_customers	guest_sessions
▶	adwords	205010	54542	54542	150468
	email	306313	81706	81706	224607
	facebook	67933	18305	18305	49628
	organic	34301	9124	9124	25177
	youtube	68202	18082	18082	50120

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Query 1.2: Channel Quality (CVR)

Description:

This query calculates the precise Conversion Rate (CVR) for each traffic source. A low CVR from a specific source (e.g., 'Facebook') may indicate low-intent traffic or misaligned ad expectations.

```
43      -- ----- Channel Quality -----
44      -- CVR traffic source
45 •  SELECT
46          e.traffic_source,
47          COUNT(DISTINCT session_id) AS total_sessions,
48          COUNT(DISTINCT CASE
49              WHEN event_type = 'purchase' THEN session_id
50              END) AS total_order_sessions,
51          round(CAST(COUNT(DISTINCT CASE
52              WHEN event_type = 'purchase' THEN session_id
53              END)
54              AS DECIMAL (18 , 4 )) / COUNT(DISTINCT session_id), 4) AS CVR
55      FROM
56          events AS e
57      GROUP BY e.traffic_source
58      ORDER BY total_sessions DESC;
```

Result Grid | Filter Rows: Export: Wrap Cell Content:

traffic_source	total_sessions	total_order_sessions	CVR
email	306313	81706	0.2667
adwords	205010	54542	0.2660
youtube	68202	18082	0.2651
facebook	67933	18305	0.2695
organic	34301	9124	0.2660

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Query 1.3: Browser Quality (CVR)

Description:

This query calculates the CVR for each browser. This is a key diagnostic tool. A low CVR from a specific browser (e.g., 'Safari') indicates a critical bug or poor user experience on that platform (e.g., iPhones) that needs immediate attention from the CX and development teams.

```
61
62      -- CVR browser
63 *   SELECT
64         e.browser,
65         COUNT(DISTINCT session_id) AS total_sessions,
66         COUNT(DISTINCT CASE
67             WHEN event_type = 'purchase' THEN session_id
68             END) AS total_order_sessions,
69         CAST(COUNT(DISTINCT CASE
70             WHEN event_type = 'purchase' THEN session_id
71             END)
72             AS DECIMAL (18 , 4 )) / COUNT(DISTINCT session_id) AS CVR
73     FROM
74         events AS e
75     GROUP BY e.browser
76     ORDER BY total_sessions DESC;
77
```

Result Grid				
	browser	total_sessions	total_order_sessions	CVR
▶	chrome	341142	91377	0.26785620
	firefox	136615	36457	0.26685942
	safari	135817	35932	0.26456187
	ie	34312	8957	0.26104570
	other	33873	9036	0.26676114

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Query 1.4: Customer Lifetime Value (CLV)

Description:

This query calculates the total *profit* generated by each customer. It calculates the historical CLV for *each customer* by summing (*sale_price* - *cost*) for all their non-returned orders ('Complete' or 'Shipped'). This identifies the company's VIP customers.

```
79      -- CLV
80  WITH OrderProfit AS (
81      SELECT
82          o.user_id,
83          oi.order_id,
84          SUM(oi.sale_price) AS total_revenue_per_order,
85          SUM(p.cost) AS total_cost_per_order,
86          (SUM(oi.sale_price) - SUM(p.cost)) AS total_profit_per_order
87      FROM
88          order_items oi
89      JOIN
90          products p ON oi.product_id = p.id
91      JOIN
92          orders o ON oi.order_id = o.order_id
93      WHERE
94          o.user_id IS NOT NULL
95          AND o.status in('Complete', 'Shipped')
96      GROUP BY
97          o.user_id, oi.order_id
98  )
99
100     SELECT
101         user_id,
102         COUNT(order_id) AS total_orders,
103         round(SUM(total_revenue_per_order),2) AS total_lifetime_revenue,
104         round(SUM(total_profit_per_order),2) AS total_lifetime_profit
```

	user_id	total_orders	total_lifetime_revenue	total_lifetime_cost	historical_clv
▶	1	1	159.99	81.13	78.86
	2	1	22	10.38	11.62
	3	2	233.46	119.37	114.09
	4	1	148	61.57	86.43
	6	1	144.82	62.79	82.03

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Query 1.5: Channel CLV (First-Touch Attribution)

Description:

This query attributes CLV back to the acquiring channel. It uses a UserFirstTouch CTE to identify the *first-ever* traffic source for each customer. It then calculates the average_clv_per_customer for each channel. This answers the most important marketing question: "Which channels acquire our most profitable customers long-term?"

```
113      -- Channels CLV
114 *  WITH CustomerCLV AS (
115     SELECT
116       o.user_id,
117       (SUM(oi.sale_price) - SUM(p.cost)) AS historical_clv
118     FROM
119       order_items oi
120     JOIN
121       products p ON oi.product_id = p.id
122     JOIN
123       orders o ON oi.order_id = o.order_id
124     WHERE
125       o.user_id IS NOT NULL
126       AND o.status in('Complete', 'Shipped')
127     GROUP BY
128       o.user_id
129   ),
130 *  UserFirstTouch AS (
131     SELECT
132       user_id,
133       traffic_source
134   FROM (
135     SELECT
```

	traffic_source	total_acquired_customers	average_clv_per_customer	total_clv_from_channel
▶	email	23842	58.77	1401185.08
	organic	2577	58.77	151456.28
	facebook	5215	58.69	306051.82
	adwords	15976	58.66	937082.3
	youtube	5301	57.36	304073.52

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Query 1.6: New vs. Returning Customer Analysis

Description:

This view measures customer loyalty by comparing the purchasing power of new vs. returning customers. It uses ROW_NUMBER() to tag each order as a user's 1st purchase ("New Customer") or a subsequent purchase ("Returning Customer"). It then calculates the AOV and revenue share for each group to prove the value of retention (e.g., "Returning customers spend 20% more per order").

```
163      -- New vs. Returning
164 *  WITH OrderRevenue AS (
165     SELECT
166       o.order_id,
167       o.user_id,
168       o.created_at,
169       SUM(oi.sale_price) AS order_revenue
170     FROM
171       orders o
172     JOIN
173       order_items oi ON o.order_id = oi.order_id
174     WHERE
175       o.status in('Complete', 'Shipped')
176     GROUP BY
177       o.order_id, o.user_id, o.created_at
178   ),
179 *  TaggedOrders AS (
180     SELECT
181       order_id,
182       user_id,
183       order_revenue,
184       ROW_NUMBER() OVER(PARTITION BY user_id ORDER BY created_at ASC) AS order_rank
185     FROM
```

Result Grid | Filter Rows: | Export: | Wrap Cell Content: |

	customer_type	total_orders	total_revenue	AOV	revenue_percentage
▶	New Customer	52911	4577136.59	86.51	76.64
	Returning Customer	16020	1394868.62	87.07	23.36

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Query 1.7: 12-Month Retention Cohort Analysis

Description:

This is the most powerful retention metric. It groups customers into "cohorts" based on their first-ever purchase month. It then tracks what percentage of each cohort returns to make another purchase in Month 1, Month 2... up to Month 12. This query was critical in discovering that our business has a long-term (7-12 month) repurchase cycle, not a seasonal one.

```
215 -- Cohort Analysis
216 WITH ValidPurchases AS (
217     SELECT DISTINCT
218         o.user_id,
219         DATE_FORMAT(o.created_at, '%Y-%m-01') AS purchase_month
220     FROM
221         orders o
222     JOIN
223         order_items oi ON o.order_id = oi.order_id
224     WHERE
225         o.status <> 'returned'
226         AND o.user_id IS NOT NULL
227 ),
228 UserCohorts AS (
229     SELECT
230         user_id,
231         MIN(purchase_month) AS cohort_month
232     FROM
233         ValidPurchases
```

cohort_month	cohort_size	month_1_retention	month_2_retention	month_3_retention	month_4_retention	month_5_reb	month_6_retent	month_7_rete	month_8_rete	month_9_rete	month_10_re	month_11_re	month_12_retention
2019-01-01	19	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	5.26315780	0.0000000	0.0000000	0.0000000	0.0000000	0.0000000
2019-02-01	34	0.0000000	0.0000000	0.0000000	0.0000000	5.88235290	5.88235290	0.0000000	5.88235290	0.0000000	2.94117640	5.88235290	2.94117640
2019-03-01	75	1.33333330	0.0000000	1.33333330	1.33333330	1.33333330	1.33333330	2.66666660	1.33333330	1.33333330	5.33333330	1.33333330	0.0000000
2019-04-01	97	1.03092780	1.03092780	1.03092780	0.0000000	2.06185560	1.03092780	0.0000000	1.03092780	3.09278350	0.0000000	1.03092780	0.0000000
2019-05-01	144	0.69444440	0.69444440	2.08333330	1.38888880	1.38888880	1.38888880	2.08333330	0.0000000	0.69444440	0.69444440	0.0000000	0.0000000
2019-06-01	171	0.58479530	1.75438590	1.16959060	1.16959060	1.75438590	0.58479530	0.58479530	1.16959060	1.75438590	0.0000000	0.0000000	0.0000000
2019-07-01	194	2.57731950	1.03092780	1.54639170	0.0000000	1.03092780	4.12371130	1.54639170	1.03092780	1.54639170	2.06185560	2.57731950	1.54639170
2019-08-01	192	0.52083330	1.56250000	1.04166660	1.04166660	1.56250000	1.56250000	1.04166660	0.0000000	4.68750000	1.56250000	3.12500000	2.08333330
2019-09-01	280	1.42857140	2.14285710	2.14285710	2.5000000	1.07142850	1.42857140	2.14285710	1.42857140	1.07142850	1.42857140	3.57142850	1.07142850
2019-10-01	287	2.43902430	2.43902430	2.78745640	2.43902430	1.04529610	1.39372820	1.04529610	2.78745640	1.74216020	0.69686410	0.0000000	1.74216020

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Query 1.8: Average Order Value (AOV) Trend

Description:

This query tracks the health of the Average Order Value (AOV) over time (both monthly and yearly). This was used to identify the strategic shift in our business.

```

296 •   SELECT
297     DATE_FORMAT(o.created_at, '%Y-%m') AS purchase_month_year,
298     COUNT(DISTINCT o.order_id) AS total_orders,
299     AVG(oi.order_revenue) AS average_order_value
300   FROM
301     orders o
302     JOIN
303       (SELECT
304         order_id, SUM(sale_price) AS order_revenue
305       FROM
306         order_items
307       GROUP BY order_id) oi ON o.order_id = oi.order_id
308   WHERE
309     o.status <> 'returned'
310   GROUP BY purchase_month_year
311   ORDER BY purchase_month_year;

```

	purchase_month_year	total_orders	average_order_value
▶	2019-01	19	102.76789479506643
	2019-02	34	107.87176488427555
	2019-03	75	95.27760021527608
	2019-04	98	100.62795959200177
	2019-05	146	89.94075331132706
	2019-06	177	80.78412453064138
	2019-07	201	85.09323383207938
	2019-08	207	95.0164735155981
	2019-09	291	89.14223368634883
	2019-10	310	84.66503246907266
	2019-11	320	80.96165630146861
	2019-12	415	84.77074706640589
	2020-01	411	91.56107048222619

```

308      -- AOV over years
309 •   SELECT
310     DATE_FORMAT(o.created_at, '%Y') AS purchase_month_year,
311     COUNT(DISTINCT o.order_id) AS total_orders,
312     AVG(oi.order_revenue) AS average_order_value
313   FROM
314     orders o
315     JOIN
316       (SELECT
317         order_id, SUM(sale_price) AS order_revenue
318       FROM
319         order_items
320       GROUP BY order_id) oi ON o.order_id = oi.order_id
321   WHERE
322     o.status <> 'returned'
323   GROUP BY purchase_month_year;

```

	purchase_month_year	total_orders	average_order_value
▶	2019	2293	87.26734418833666
	2020	7772	86.61021238514829
	2021	15031	87.12830358782925
	2022	26312	85.24134322560151
	2023	53072	86.48417724191617

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Query 1.9: Cart Abandonment by Channel

Description:

This query analyzes "cart abandonment" by traffic source. It isolates sessions that successfully *reached* the cart and calculates the percentage of those sessions that *failed* to purchase. This helps the CMO identify channels (e.g., 'Facebook') that drive low-intent "window-shopping" traffic, or where promotional offers may be misleading.

```
313      -- ----- C.Marketing Behavior -----
314
315 •      WITH SessionFacts AS (
316     SELECT
317         session_id,
318         MAX(traffic_source) AS traffic_source,
319         MAX(CASE
320             WHEN event_type = 'cart' THEN 1
321             ELSE 0
322             END) AS reached_cart,
323         MAX(CASE
324             WHEN event_type = 'purchase' THEN 1
325             ELSE 0
326             END) AS did_purchase
327     FROM
328         events
329     GROUP BY
330         session_id
331     )
332     SELECT
333         traffic_source,
334         COUNT(session_id) AS total_sessions_reached_cart,
```

Result Grid	Filter Rows:	Export:	Wrap Cell Contents:	
▶	traffic_source	total_sessions_reached_cart	total_sessions_purchased	cart_conversion_rate_percent
▶	youtube	43182	18082	41.87392890
	adwords	130001	54542	41.95506180
	email	193988	81706	42.11910010
	facebook	43372	18305	42.20464810
	organic	21603	9124	42.23487470

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Query 1.10: Top Regions by Revenue & AOV

Description:

This query identifies the most valuable geographic markets (country, state, and city) by ranking them based on total revenue and AOV. This provides a clear guide for regional marketing spend, logistics planning, and expansion strategy.

```
346      -- Top region by Revenue & AOV
347
348  WITH OrderRevenue AS (
349      SELECT
350          o.order_id,
351          o.user_id,
352          SUM(oi.sale_price) AS order_revenue
353      FROM
354          orders o
355      JOIN
356          order_items oi ON o.order_id = oi.order_id
357      WHERE
358          o.status <> 'returned'
359      GROUP BY
360          o.order_id, o.user_id
361    ),
362  UserOrders AS (
363      SELECT
364          u.country,
365          u.state,
366          u.city,
367          o.order_id,
```

Result Grid						
Filter Rows:			Export:		Wrap Cell Content:	
country	state	city	total_orders	total_revenue	AOV	
South Korea	Seoul	Seoul	1673	149635.17	89.44	
United States	New York	New York	954	80778.99	84.67	
Japan	Tokyo	Tokyo	846	78160.29	92.39	
Brasil	Bahia	Salvador	624	49679.8	79.62	
United Kingdom	England	London	557	48132.81	86.41	
Brasil	São Paulo	São Paulo	538	46970.97	87.31	
South Korea	Busan	Busan	486	45384.85	93.38	

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Category 2: CX (Customer Experience) Analysis

Goal:

To identify and measure points of "friction" in the customer journey. These views analyze on-site behavior, quantify drop-off points, and track customer dissatisfaction through detailed return metrics.

Query 2.1: Funnel Diagnostics

Description:

This is a diagnostic query used to discover the *true* user journey. It buckets every session into one of 16 possible combinations of actions (e.g., visited_home=0, viewed_product=1, added_to_cart=1, did_purchase=0). This query was critical in proving that our user funnel *does not* start at the homepage, but at the product page.

```
391      -- CX ANALYSIS -----
392      -- ----- A.On-Site Friction Analysis ---
393      -- Funnel Drop-off -----
394
395      -- identifying the right schema
396  • WITH SessionFacts AS (
397          SELECT
398              session_id,
399              MAX(CASE WHEN event_type = 'home' THEN 1 ELSE 0 END) AS visited_home,
400              MAX(CASE WHEN event_type = 'product' THEN 1 ELSE 0 END) AS viewed_product,
401              MAX(CASE WHEN event_type = 'cart' THEN 1 ELSE 0 END) AS added_to_cart,
402              MAX(CASE WHEN event_type = 'purchase' THEN 1 ELSE 0 END) AS did_purchase
403          FROM
404              events
405          GROUP BY
406              session_id
407      )
408
409      SELECT
410          visited_home,
411          viewed_product,
412          added_to_cart,
```

Result Grid					
	visited_home	viewed_product	added_to_cart	did_purchase	session_count
▶	0	1	1	0	250387
	0	1	0	0	249613
	0	1	1	1	94047
	1	1	1	1	87712

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Query 2.2: Funnel Performance & Trend

Description: This query calculates the *real* user funnel (Product -> Cart -> Purchase), which was discovered in the diagnostic view. It calculates the product_to_cart_rate and cart_to_purchase_rate, identifying the two biggest drop-off points. It also trends this performance year-over-year to measure the impact of site improvements.

```
424 -- funnel
425 WITH SessionFacts AS (
426     SELECT
427         session_id,
428         MAX(CASE WHEN event_type = 'product' THEN 1 ELSE 0 END) AS viewed_product,
429         MAX(CASE WHEN event_type = 'cart' THEN 1 ELSE 0 END) AS added_to_cart,
430         MAX(CASE WHEN event_type = 'purchase' THEN 1 ELSE 0 END) AS did_purchase
431     FROM
432         events
433     GROUP BY
434         session_id
435     ),
436     FunnelCounts AS (
437         SELECT
438             SUM(viewed_product) AS step1_viewed_product,
439             SUM(CASE
440                 WHEN viewed_product = 1 AND added_to_cart = 1 THEN 1
441                 ELSE 0
442             END) AS step2_added_to_cart,
443             SUM(CASE
444                 WHEN viewed_product = 1 AND added_to_cart = 1 AND did_purchase = 1 THEN 1
445                 ELSE 0
446             END) AS step3_purchased
447     )
448     SELECT
449         step1_viewed_product,
450         step2_added_to_cart,
451         product_to_cart_rate,
452         step3_purchased,
453         cart_to_purchase_rate
454     FROM
455         FunnelCounts
```

```
474 -- funnel trend
475 WITH SessionFacts AS (
476     SELECT
477         DATE_FORMAT(created_at, '%Y') over_years,
478         session_id,
479         MAX(CASE WHEN event_type = 'product' THEN 1 ELSE 0 END) AS viewed_product,
480         MAX(CASE WHEN event_type = 'cart' THEN 1 ELSE 0 END) AS added_to_cart,
481         MAX(CASE WHEN event_type = 'purchase' THEN 1 ELSE 0 END) AS did_purchase
482     FROM
483         events
484     GROUP BY
485         session_id, DATE_FORMAT(created_at, '%Y')
486     ),
487     FunnelCounts AS (

```

Result Grid | Filter Rows: Export: Wrap Cell Content:

	over_years	step1_viewed_product	step2_added_to_cart	product_to_cart_rate	step3_purchased	cart_to_purchase_rate
▶	2019	102094	52938	51.85	3646	6.89
	2020	112190	62417	55.64	12668	20.30
	2021	123608	74134	59.98	24254	32.72
	2022	141246	91921	65.08	42149	45.85
	2023	185097	135434	73.17	85064	62.81
	2024	17525	15303	87.32	13166	86.04

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Query 2.3: Page Exit Rate

Description:

This query creates an "action list" for the CX Manager by identifying the "leakiest" pages on the site. It calculates the exit rate (total_exits / total_views) for every URI, ordered from worst to best. A high exit rate on a page like /cart would indicate a clear friction point driving customers away.

```
460      -- Exit Rate
461  WITH RankedEvents AS (
462      SELECT
463          session_id,
464          uri,
465          ROW_NUMBER() OVER(PARTITION BY session_id ORDER BY sequence_number DESC) AS event_rank
466      FROM
467          events
468  ),
469  ExitCounts AS (
470      SELECT
471          uri,
472          COUNT(*) AS total_exits
473      FROM
474          RankedEvents
475      WHERE
476          event_rank = 1
477          AND uri IS NOT NULL
478      GROUP BY
479          uri
480  ),
481  PageViews AS (
482      SELECT
483          uri,
484          COUNT(*) AS total_views
485      FROM
486          events
487      WHERE
488          event_rank = 1
489          AND uri IS NOT NULL
490      GROUP BY
491          uri
492  )
493  SELECT
494      e.uri,
495      e.total_exits,
496      p.total_views,
497      (e.total_exits / p.total_views) * 100 AS exit_rate_percent
498  FROM
499      ExitCounts e
500      JOIN PageViews p ON e.uri = p.uri
501  ORDER BY
502      exit_rate_percent DESC;
```

Result Grid | Filter Rows: _____ | Export: _____ | Wrap Cell Content: _____

uri	total_exits	total_views	exit_rate_percent
cancel	125568	125568	100.00000000
purchase	181759	181759	100.00000000
cart	124819	432146	28.88352540

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Query 2.4: State Return Rates

Description:

This view identifies high-risk geographic areas by calculating the order return rate for each state. It uses the safe SUM(CASE... logic to count returns (year(returned_at) <> '0000'). A high rate in a specific state can point to issues with a regional shipping carrier or local fraud.

```
508      -- ----- B.Dissatisfaction Analysis -----
509      -- states return rates
510  • WITH StateOrders AS (
511      SELECT
512          u.state,
513          o.order_id,
514          o.returned_at
515      FROM
516          orders o
517      JOIN
518          users u ON o.user_id = u.id
519      WHERE
520          u.state IS NOT NULL
521          AND year(o.shipped_at) <> '0000'
522      )
523      SELECT
524          state,
525          COUNT(order_id) AS total_shipped_orders,
526          SUM(CASE
527              WHEN year(returned_at) <> '0000' THEN 1
528              ELSE 0
529          END) AS total_returned_orders,
```

Result Grid Filter Rows: <input type="text"/> Export: Wrap Cell Content:				
	state	total_shipped_orders	total_returned_orders	return_rate_percent
▶	Dolnośląskie	75	17	22.66666660
	Sergipe	171	36	21.05263150
	Daejeon	100	21	21.00000000
	Brussels	67	14	20.89552230
	Schleswig-Holstein	97	20	20.61855670
	Gyeongsangbuk-do	64	13	20.31250000
	Ninolia Hui Autonomous Region	181	36	19.88950270

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Query 2.5: Order Return Rate (ORR) Trend

Description:

This is a CX metric that tracks the % of *orders* that experienced *any* return (even a partial one). A high ORR means many customers are being forced to interact with the returns system, indicating high operational friction. This query tracks the trend yearly.

```
559      -- Order Return Rate - ORR
560  • WITH OrderReturnStatus AS (
561      SELECT
562          order_id,
563          MAX(CASE
564              WHEN status = 'Returned' THEN 1
565              ELSE 0
566          END) AS has_a_return
567      FROM
568          order_items
569      GROUP BY
570          order_id
571  )
572  SELECT
573      DATE_FORMAT(o.created_at, '%Y') over_years,
```

over_years	total_shipped_orders	orders_with_a_return	order_return_rate_percent
2019	1674	245	14.63560330
2020	5607	870	15.51631880
2021	10868	1673	15.39381670
2022	19017	2859	15.03391700
2023	38328	5969	15.57347100
2024	5967	914	15.31758000

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Query 2.6: Item Return Rate (IRR) Trend

Description:

This is a financial/quality metric that tracks the % of *individual items* that were returned. This is the most accurate measure of the *cost of returns* and is a key performance indicator for the Head of Products and CFO. This query tracks the trend yearly.

```
586 -- Item Return Rate - IRR
587 • SELECT
588     DATE_FORMAT(o.created_at, '%Y') over_years,
589     COUNT(oi.id) AS total_shipped_items,
590     SUM(CASE
591         WHEN oi.status = 'Returned'
592             THEN 1
593         ELSE 0
594     END) AS total_returned_items,
595     (CAST(SUM(CASE WHEN oi.status = 'Returned' THEN 1 ELSE 0 END) AS DECIMAL(18,4)) / COUNT(oi.id)) * 100 AS item_return_rate_percent
596     FROM
597     order_items oi
598     JOIN
599     orders o ON oi.order_id = o.order_id
600     WHERE
```

Result Grid | Filter Rows: Export: Wrap Cell Content: ▾

over_years	total_shipped_items	total_returned_items	item_return_rate_percent
2019	2409	355	14.73640510
2020	8236	1309	15.89363760
2021	15798	2419	15.31206480
2022	27551	4106	14.90327020
2023	55615	8707	15.65584820
2024	8672	1336	15.40590400

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Query 2.7: Toxic Products

Description:

This query identifies the "Top Toxic Products with 100% return rate" that cause the most customer dissatisfaction. This separates high-volume products with acceptable returns from low-volume products that are truly problematic and damaging to the brand.

```
647      -- Toxic Products
648 • WITH ItemProductStatus AS (
649       SELECT
650         p.name,
651         oi.id AS item_id,
652       (CASE
653         WHEN oi.status = 'Returned' THEN 1
654         ELSE 0
655       END) AS is_returned
656     FROM
657       order_items oi
658     JOIN
659       orders o ON oi.order_id = o.order_id
660     JOIN
```

	name	total_shipped_items	total_returned_items	return_rate_percent
▶	Jones New York Women's Petite 3/4 Sleeve Blouse	2	2	100.00000000
	Democracy Women's Triple Collar Woven Surplu...	1	1	100.00000000
	Calvin Klein Women's Peasant Blouse	1	1	100.00000000
	Allegra K Ladies Button Detail V Neck Close-fittin...	1	1	100.00000000
	Foxcroft Women's 3/4 Sleeve Fitted Blouse	1	1	100.00000000
	Calvin Klein Jeans Women's Convertible Henley ...	1	1	100.00000000
	Gildan G240L Ladies 6.1 oz. Ultra Cotton Long-S...	1	1	100.00000000

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Query 2.8: Return Rate by Category

Description:

This query pinpoints systemic product issues by calculating the item return rate (%) for each product *category*. This guides the Head of Products on where to focus quality control efforts.

```
600      -- Return Rate by Category
601 • WITH ItemCategoryStatus AS (
602     SELECT
603       p.category,
604       oi.id AS item_id,
605       (CASE
606         WHEN oi.status = 'Returned' THEN 1
607         ELSE 0
608       END) AS is_returned
609     FROM
610       order_items oi
611     JOIN
612       orders o ON oi.order_id = o.order_id
613     JOIN
614       products p ON oi.product_id = p.id
615     WHERE
616       year(o.shipped_at) <> '0000'
617   )
```

Result Grid				
	category	total_shipped_items	total_returned_items	return_rate_percent
▶	Intimates	8784	1364	15.52823310
	Jeans	8234	1237	15.02307500
	Fashion Hoodies & Sweatshirts	7766	1217	15.67087300
	Tops & Tees	7720	1172	15.18134710
	Shorts	7174	1139	15.87677720
	Swim	7398	1133	15.31494990
	Sweaters	7322	1130	15.43294180
	Sleep & Lounge	7206	1060	14.70996390
	Active	5979	947	15.83876900
	Accessories	6249	945	15.12241950

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Query 2.9: Return Rate by Distribution Center

Description:

This query identifies operational issues at the warehouse level. It calculates the item return rate (%) for each *Distribution Center*. A high rate from a specific DC (e.g., "New Orleans LA ") while other DCs are normal points to an operational problem (e.g., poor packaging, bad quality control) rather than a product-specific problem.

```
630      -- Return Rate by Geography
631  • WITH ShippedItemStatus AS (
632      SELECT
633          oi.id AS item_id,
634          oi.product_id,
635          (CASE
636              WHEN oi.status = 'Returned' THEN 1
637              ELSE 0
638          END) AS is_returned
639      FROM
640          order_items oi
641      JOIN
642          orders o ON oi.order_id = o.order_id
643      WHERE
644          year(o.shipped_at) <> '0000'
645      ),
646
647  • ItemDistribution AS (
```

Result Grid				
	distribution_center_name	total_shipped_items	total_returned_items	return_rate_percent
▶	New Orleans LA	8660	1428	16.48960730
	Houston TX	14663	2300	15.68573960
	Port Authority of New York/New Jersey NY/NJ	10508	1644	15.64522260
	Charleston SC	10951	1682	15.35932790
	Los Angeles CA	11262	1729	15.35251280
	Memphis TN	15640	2400	15.34526850
	Philadelphia PA	11058	1692	15.30113940
	Chicago IL	15657	2379	15.19448170
	Savannah GA	7786	1181	15.16825070
	Mobile AL	12096	1797	14.85615070