

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;
```

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   Filter Rows: Export:  Wrap Cell Content: 

	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,
```

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

	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
```

Result Grid Filter Rows: Export: Wrap Cell Content:				
	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_re	month_6_reten	month_7_reter	month_8_rete	month_9_rete	month_10_r	month_11_re	month_12_retention
2019-01-01	19	0.00000000	0.00000000	0.00000000	0.00000000	0.00000000	0.00000000	0.00000000	5.26315780	0.00000000	0.00000000	0.00000000	0.00000000
2019-02-01	34	0.00000000	0.00000000	0.00000000	5.88235290	5.88235290	5.88235290	0.00000000	5.88235290	0.00000000	2.94117640	5.88235290	2.94117640
2019-03-01	75	1.33333330	0.00000000	1.33333330	1.33333330	1.33333330	2.66666660	1.33333330	1.33333330	5.33333330	1.33333330	0.00000000	0.00000000
2019-04-01	97	1.03092780	1.03092780	1.03092780	0.00000000	2.06185560	1.03092780	0.00000000	1.03092780	3.09278350	0.00000000	1.03092780	0.00000000
2019-05-01	144	0.69444440	0.69444440	2.08333330	1.38888880	1.38888880	1.38888880	1.38888880	2.08333330	0.00000000	0.69444440	0.69444440	0.00000000
2019-06-01	171	0.58479530	1.75438590	1.16959060	1.16959060	1.75438590	0.58479530	0.58479530	1.16959060	0.58479530	1.16959060	1.75438590	0.00000000
2019-07-01	194	2.57731950	1.03092780	1.54639170	0.00000000	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	0.00000000	4.68750000	1.56250000	3.12500000	2.08333330
2019-09-01	280	1.42857140	2.14285710	2.14285710	2.50000000	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	0.00000000	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 Content: [IA](#)

	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,
```

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

	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 )
```

step1_viewed_product	step2_added_to_cart	product_to_cart_rate	step3_purchased	cart_to_purchase_rate
681759	432146	63.39	181759	42.06

```
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 (
488     SELECT
489         over_years,
490         SUM(viewed_product) AS step1_viewed_product,
491         SUM(CASE
492             WHEN viewed_product = 1 AND added_to_cart = 1 THEN 1
493             ELSE 0
494         END) AS step2_added_to_cart,
495         SUM(CASE
496             WHEN viewed_product = 1 AND added_to_cart = 1 AND did_purchase = 1 THEN 1
497             ELSE 0
498         END) AS step3_purchased
499 )
```

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 ($\text{total_exits} / \text{total_views}$) 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 (
```

Result Grid

Filter Rows:

Exports:

Wrap Cell Content:

	uri	total_views	total_exits	exit_rate_percent
▶	cancel	125568	125568	100.00000000
	purchase	181759	181759	100.00000000
	cart	432146	124819	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: | Export: | Wrap Cell Content: [IA](#)

	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
	Ninaxia 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
```

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  )
```

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 )
```

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 (
```

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