

LLM-Powered
Review Analysis:
Optimising Data
Engineering using Airflow



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About me



Naseem Shah

- Head of Engineering at Xena Intelligence
- Loves to build products and engineer solutions
- Contributor to Airflow
- Fun Fact: I love airplanes; I'm a big-time aviation fan :)

Context



Amazon APIs

Walmart APIs

Shopify APIs

Public Web

User-facing Products/Tools:

Xena 360: Multichannel tool

Foresight: Listing Management tool BrandVoice: Market Research tool

Internal:

Our Backend + Data

Pipelines

Reports, Internal Dashboards, Alerts,

Enterprise Services:

Market Research, Shopper Preference Reports

Review Analysis Requirement



It all started with one big question from a long-term enterprise customer

"Should we launch this new product in this market?"

"What do shoppers truly value across marketplaces?"

"What key risks and success factors must we plan for?"

Solution: Market Research



We analyzed thousands of real shopper reviews to uncover what buyers care about most and turned those insights into a shopper preference report.

Identify Best-Selling Products Analyze Key Features & Quality

Map Direct Competitors

Collect user reviews

Shopper Preference Report



Review Analysis Workflow



Multiple Data Analysts, Manually Copy Paste Reviews

Analyzing them one by one and grouping insights

Building the final PDF

Downsides/challenges



Manual workflows quickly became unsustainable as projects and data grew.

Efficiency

Accuracy

Scale

Time-intensive

Error-prone

Doesn't Scale Across
Products

Weeks of Manual Work

Heavy Human Involvement Hard to Scale Across Marketplaces

From Manual to Automated



To overcome these challenges, we automated the entire workflow using Airflow + LLMs.









Reviews Analysis: Challenges & Insights



To overcome these challenges, we automated the entire workflow using Airflow + LLMs.

Text Mining

Lack of Structure in Reviews

Difficulty in understanding context

Missing Key Sentiments

Lack of Structured Data for Actionable Insights

Review Analysis Tasks



Summarisation

Aspect-Based Sentiment Analysis

Use cases

Buyer Audience Identification

Keywords Extraction

Improvement Suggestions

Translation

Themes & Topic Identification

Review Summarization

- Reading Lengthy reviews sucks!
- Easiest task in the review analysis
- Reviews have no standard structure
- Do we really need to read the full review to get the context?





Review Summarization



Simple prompt

You are an expert in analyzing e-commerce product reviews.

Summarise the following product review

Output is also a string, which is what we need



The reviewer appreciates the watch for its lightweight design, affordability, useful functions, and long battery life. However, they note issues with the analog setup, glass quality, and synchronization between analog and digital displays.



Traditional Sentiment Analysis

Keyword-based:

"good" = positive, "bad" = negative (no nuance).

Shallow context:

can't capture sarcasm, irony, or aspect-level meaning.

Domain-specific models:

Needed retraining for every product/industry.

Bag-of-Words:

ignores word order \rightarrow "not good" still looks positive.



Why does context matter in sentiment analysis?

LLMs understand context = no more retraining.

Same word, different meaning:

"This phone is light and easy to carry" (positive).

"This phone feels light and cheap" (negative).

Without context, the model misclassifies.



Basic Sentiment Analysis using LLMs

Read the review and classify the sentiment as Positive, Negative, or Neutral.

Review: 'The product quality is great but delivery was late.'

✓ Sentiment: Neutral

Reason: The review contains both a **positive** remark about the product quality and a **negative** remark about the delivery delay. Since the positive and negative sentiments balance each other, the overall tone is **neutral/mixed**.



Downsides to this approach

Read the review and classify the sentiment as Positive, Negative, or Neutral.

Review:

I really like the phone's camera and the battery lasts all day, but the screen scratches way too easily and the charger stopped working in a week.

d Overall Sentiment: Negative

because the frustrations outweigh the positives when considering overall user experience).

Aspect-Based Sentiment Analysis



By treating a review as a single sentiment, we miss the rich layers of insight each line contains.

Adjusting the prompt gives us this:

Analyze the following review and identify all the key aspects mentioned. For each aspect, classify the sentiment as Positive, Negative, or Neutral, and explain briefly why.

Review: '[insert review text here]

Aspect Based Sentiment Analysis



■ Aspect-Wise Analysis						
Aspect	Sentiment	Explanation				
Camera quality	Positive	The user says "I really like the phone's camera," showing clear satisfaction with this feature.				
Battery life	Positive	"The battery lasts all day" indicates a good experience with battery performance.				
Screen durability	Negative	"The screen scratches way too easily" highlights a clear complaint about build quality/durability.				
Charger reliability	Negative	"The charger stopped working in a week" points to a failure of the accessory, leading to a negative experience.				

The challenge: Strucuture



We get the response from the LLM — but it's just a string of text.

As mentioned earlier, text mining is hard.

Why do we need structure?



We need the output in a format we can easily parse, store, and work with.

That's where JSON comes in.

Why JSON?



Structured & flexible

Web-native

Supports nesting

Dashboard-ready

Modelling the JSON



Considerations:

UI/UX

Handling multiple languages

Context aware

SQL Query Friendly

Support Nested Structures

Extendable

Adjusted Prompt



Prompt

```
You are an expert in analyzing e-commerce product
reviews, focusing on sentiment analysis. Analyze and
return the following information in JSON format:
positive_sentiments: Extract the positive sentiments
from the review and organize them as an array of
objects. Each object should contain an aspect, a
comment and a sentimentScore (0-100).
"positive_sentiments": [
   "aspect": "Add the positive sentiment aspect here",
   "comment": "Add the direct mention here",
   "sentimentScore": "sentiment score out of 100"
```

Output

```
"positive sentiments": [
    "aspect": "Comfort",
    "comment": "Super comfortable shoes",
    "sentimentScore": 95
    "aspect": "Support",
    "comment": "excellent support",
    "sentimentScore": 92
 },{...},{...}
```

Handling Foreign Languages



Original Prompt

```
You are an expert in analyzing e-commerce product
reviews, focusing on sentiment analysis. Analyze and
return the following information in JSON format:
positive_sentiments: Extract the positive sentiments
from the review and organize them as an array of
objects. Each object should contain an aspect, a
comment and a sentimentScore (0-100).
"positive_sentiments": [
   "aspect": "Add the positive sentiment aspect here",
   "comment": "Add the direct mention here",
   "sentimentScore": "sentiment score out of 100"
```

Output

```
"positive_sentiments": [
      "aspect": "Product quality",
      "comment": "Buena calidad!!",
      "sentimentScore": 90
      "aspect": "Design/Functionality",
      "comment": "Para ajustar la hora
analógica, se hace desde los botones, lo
que da mayor seguridad al uso en el agua.",
      "sentimentScore": 85
```

Handling Foreign Languages



Adding a comment translation key

```
6 6 6
You are an expert in analyzing e-commerce product reviews,
focusing on sentiment analysis. Analyze and return the
following information in JSON format:
positive_sentiments: Extract the positive sentiments from
the review and organize them as an array of objects. Each
object should contain an aspect, a comment, a
commentTranslated (if the review is in a non-English
language, include the corresponding direct mention from the
translatedReview key; otherwise, set this to null) and a
sentimentScore (0-100).
"positive_sentiments": [
   "aspect": "Add the positive sentiment aspect here",
   "comment": "Add the direct mention from the review here",
   "commentTranslated": "If the review is in a non-English
language, add the corresponding direct mention from the
translatedReview key here; otherwise, this becomes null",
   "sentimentScore": "sentiment score out of 100"
```

Output

```
. .
  "positive_sentiments": [
      "aspect": "Product quality",
      "comment": "Buenísimo el reloi. Buena
calidad!!",
      "commentTranslated": "The watch is great.
Good quality!!",
      "sentimentScore": 90
      "aspect": "Product functionality",
      "comment": "Para ajustar la hora
analógica, se hace desde los botones, lo que da
mayor seguridad al uso en el agua.",
      "commentTranslated": "To adjust the analog
time, it is done from the buttons, which
provides greater safety for use in water.",
      "sentimentScore": 85
```

Themes & Topics



Themes and **topics** help organize insights in a structured, actionable way.

Themes

- High-level category aspect
- Think of it as the "big idea" behind what reviewers are talking about.
- Examples:
 Product Quality, Packaging,
 Delivery Experience, Ease of Use,

Topics

- A topic is a specific sub-aspect within a theme.
- It captures exact details or points mentioned in the review.
- Examples: Durability, Comfort, Material, and Color Accuracy.

Modeling Themes & Topics



Themes and **topics** help organize insights in a structured, actionable way.

```
You are an expert in analyzing e-commerce product reviews, focusing on sentiment
analysis, keyword optimization, and actionable insights. For each product
review, analyze and return the following information in JSON format:
themes: Organize the review insights into themes, where each theme contains a
list of topics and associated mentions. Each theme represents a general
category, and each topic within the theme represents a specific aspect of the
review with direct mentions from the review body.
"themes": [{
    "theme": "Add the theme here",
    "topics": [{
        "topic": "Add the topic here",
        "mentions": [
            "Add mention from the review here"
       "translatedMentions": [
            "Add mention from the translatedReview here"
```

```
"theme": "Product Quality",
"topics": [
    "topic": "Comfort",
    "mentions": ["The shoes feel
     soft and light on my feet"],
    "translatedMentions": [null]
```

Other Analysis tasks



```
Buyer Audience

"buyer_audience": [
    "runners",
    "athletes",
    "people with pronation issues"
]
```

```
Use cases

"usecases": [
    "running",
    "long-distance running",
    "athletic training"
],
```

```
Improvements
```

```
"improvements": [
    "Improve toe box fit",
    "Enhance arch support",
],
```

```
Keywords Extraction
   . .
     "keywords": [
       "pronated stride shoes",
       "GTS running shoes",
       "cushioning running shoes",
       "best running shoes",
       "running shoes for pronation"
```

The result?



Unstructured Text



rive had my snare of both casio regular and Genote wateres and petry familiar with relatively consistent programming and functions of their other designs. This one turned out to be an outlier. It was supposed to be an upgrade from my last Casio, AQ-5810W, but in many ways it was a downgrade.

I like how it tooks, at least in tan/black color scheme. It supposedly has a 10-year battery, which I prefer over the Solar charging. With solar, when the watch wasn't exposed to enough light, it'd go into a "hibernation" mode and effectively shut off the display to preserve and extend battery life, which happened on more than one occasion and usually at night. So a 10-year battery is a plus.

The LCD display on this model shows both date and day of the week in the same view (on AQS810W you had to change the view, which was really annoying)

HDC700 also has a dual clock, but it's implemented very poorly- to the point where it's practically useless. It seemed like a good idea to have the analog hands show the time in the current/home location and for the LCD to show the time in another time zone, however there's only one calendar and it's tied to the LCD. Read more

10 people found this helpful

Out-

Report

Summarisation

Buyer Audience Identification

Translation

Aspect-Based Sentiment Analysis

Keywords Extraction

Themes & Topic Identification

Use cases

Improvement Suggestions

Structured Data

```
"positive sentiments": |
      "aspect": "Product quality",
      "comment": "Buenísimo el reloj. Buena
calidad!!",
      "commentTranslated": "The watch is great.
Good quality!!",
      "sentimentScore": 90
      "aspect": "Product functionality",
      "comment": "Para aiustar la hora
analógica, se hace desde los botones, lo que da
mayor seguridad al uso en el agua.",
      "commentTranslated": "To adjust the analog
time, it is done from the buttons, which
provides greater safety for use in water.",
      "sentimentScore": 85
```

Putting it all together









SQL queries



Orchestrator |

Database



Interactive Dashboard

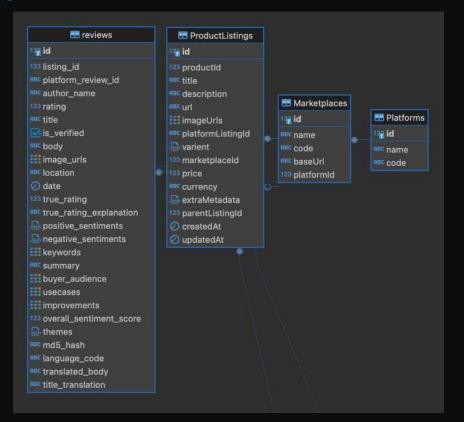




Entity Relationship Modeling



model reviews {		
id	Int	@id @default(a
language_code	String?	
author_name	String?	
rating	Int?	
title	String?	
is_verified	Boolean?	
body	String?	
translated_body	String?	
title_translation	String?	
date	DateTime	
positive_sentiments	Json?	@db.JsonB
negative_sentiments	Json?	@db.JsonB
keywords	String[]	
summary	String?	
buyer_audience	String[]	
usecases	String[]	
improvements	String[]	
themes	Json?	@db.JsonB
md5_hash	String	
listing_id	Int	
product_listing	ProductListings	@relation(fiel
@@unique([listing_id,	md5_hash])	
}		



Advantages of JsonB in Postgres



Flexibility

Faster Queries

Querying Specific Parts of JSON

Partial Updates

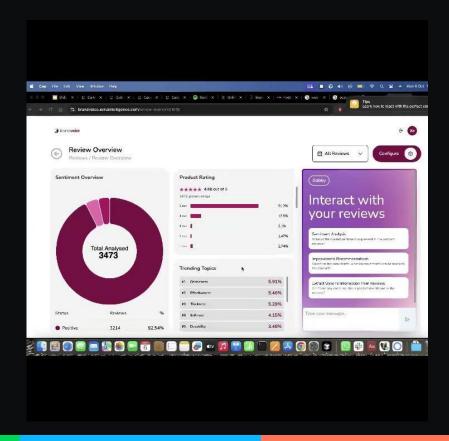
Evolving Schema

Example Query

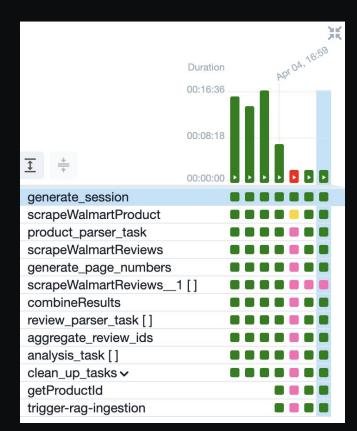
Groups all the positive sentiment

```
. .
SELECT
  aspect data ->> 'aspect' AS aspect,
 COUNT(*)::int AS count,
  jsonb_agg(
    isonb build object(
      'mention', aspect data ->> 'comment',
      'review', jsonb_build_object(
        'id', r.id,
        'rating', r.rating,
        'title', r.title,
        'body', r.body,
        -- ... other review fields ...
        'themes', r.themes,
        'md5 hash', r.md5 hash
  ) AS comments
FROM reviews r
CROSS JOIN LATERAL
isonb array elements(r.positive sentiments) AS
aspect data
GROUP BY aspect
ORDER BY count DESC
LIMIT 100;
```



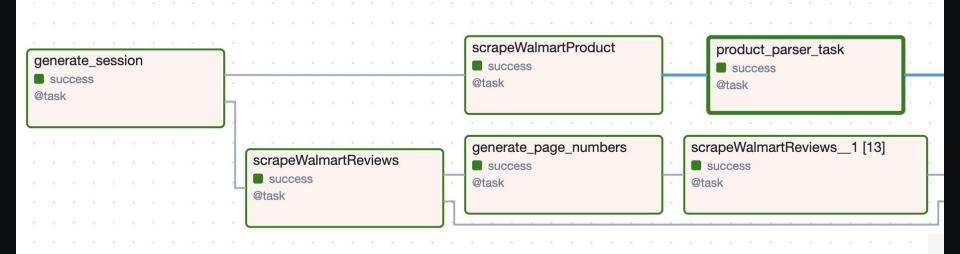






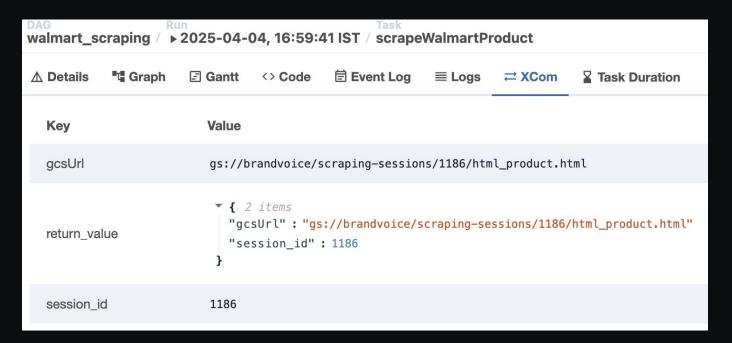




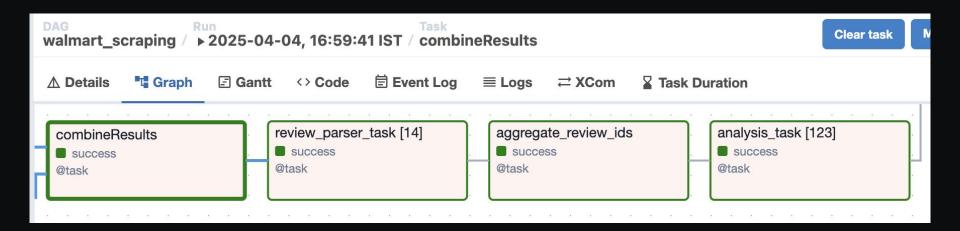




XComs are powerful, but use them wisely









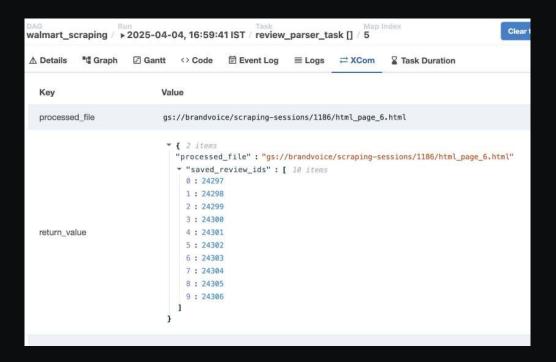
Mapped tasks make multi-output workflows seamless

Key	Value
return_value	<pre>" [14 items 0 : "gs://brandvoice/scraping-sessions/1186/html_page_1.html" 1 : "gs://brandvoice/scraping-sessions/1186/html_page_2.html" 2 : "gs://brandvoice/scraping-sessions/1186/html_page_3.html" 3 : "gs://brandvoice/scraping-sessions/1186/html_page_4.html" 4 : "gs://brandvoice/scraping-sessions/1186/html_page_5.html" 5 : "gs://brandvoice/scraping-sessions/1186/html_page_6.html" 6 : "gs://brandvoice/scraping-sessions/1186/html_page_7.html" 7 : "gs://brandvoice/scraping-sessions/1186/html_page_8.html" 8 : "gs://brandvoice/scraping-sessions/1186/html_page_10.html" 10 : "gs://brandvoice/scraping-sessions/1186/html_page_11.html" 11 : "gs://brandvoice/scraping-sessions/1186/html_page_12.html" 12 : "gs://brandvoice/scraping-sessions/1186/html_page_13.html" 13 : "gs://brandvoice/scraping-sessions/1186/html_page_14.html"</pre>

walmart_scraping / > 2025-04-04, 16:59:41 IST / review_parser_task []							
△ Details	*⊈ Graph	☑ Gantt	<> Code	Event Log	[] Mapped Tasks	☑ Task Dur	
MAP INDEX	•	STATE \$		DURATION #	START DATE \$		
0		success		00:00:01	2025-04-04, 17:	03:11 IST	
1		success		00:00:02	2025-04-04, 17:03:10 IST		
2		success		00:00:02	2025-04-04, 17:03:10 IST		
3		success 00:00:02 2025-04-04, 17:03:		03:10 IST			
4		success		00:00:01	2025-04-04, 17:	03:11 IST	
5		success		00:00:02	2025-04-04, 17:03:07 IST		
6		success		00:00:01	2025-04-04, 17:03:11 IST		
7		success		00:00:02	2025-04-04, 17:03:10 IST		
8		success		00:00:02	2025-04-04, 17:	03:10 IST	
9		SUCCESS		00-00-02	2025-04-04 17:	03:10 IST	

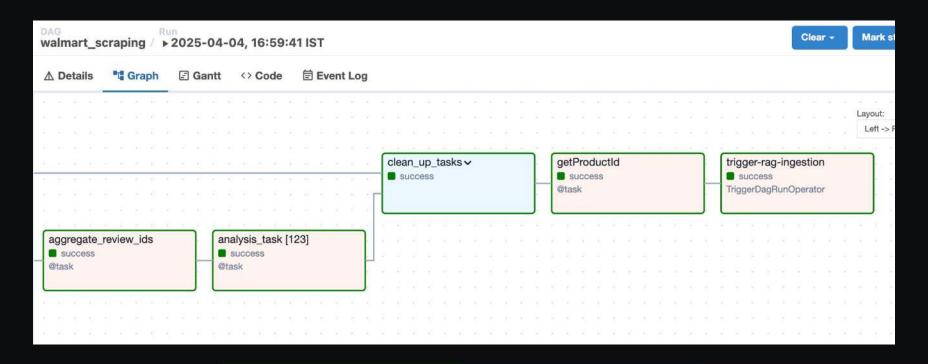


Each parser task processes reviews, saves them to the database, and returns their IDs.



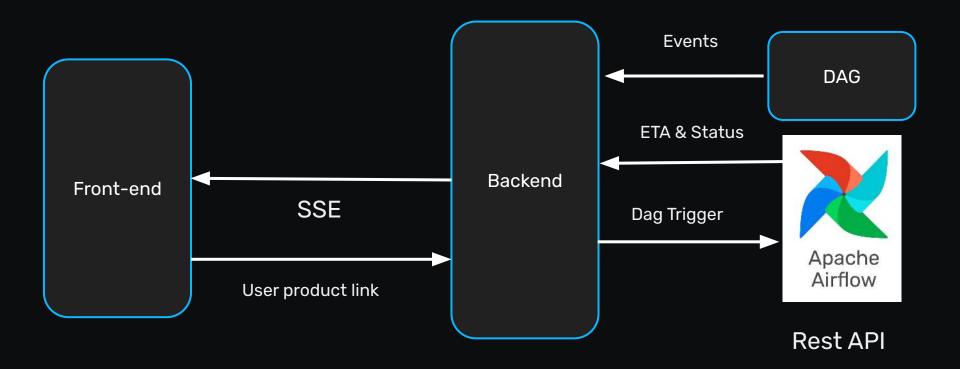


The output from all parser tasks is aggregated and passed into an analysis step, mapped task pulls reviews by their IDs and sends them to an LLM for analysis.



Getting progress, ETAs, etc.





Getting progress, ETAs, etc.



```
default_args = {
    'owner': 'airflow',
    'start_date': datetime(2025, 3, 1),
    'retries': 5,
    'retry_delay': timedelta(seconds=20),
    "on_success_callback": send_eta_and_progress_to_webhook
}

%L to chat, %K to generate

# Define a separate function for extracting reviews

@dag(default_args=default_args, schedule_interval=None, catchupted for walmart_brandvoice_lite(**context):
    dag_conf = extract_dag_conf()
```

```
def send_eta_and_progress_to_webhook(context):
    elapsed, remaining, eta, progress = compute dag eta and proc
    product id = context["dag run"].conf.get("product id")
    payload = {
        "event_type": "overview",
        "elapsed":
                     str(elapsed).
        "eta":
                    eta.
        "progress": progress,
    base url = Variable.get("BV LITE API BASE URL")
    url = f"{base_url}/events/stream/product_fproduct_id}"
    try:
        # Ensure all datetime objects in the payload are convert
        serialized payload = {k: (v.isoformat() if isinstance()
            v, datetime) else v) for k, v in payload.items()}
        requests.post(url, ison=serialized payload, timeout=5)
    except Exception as e:
        context["task_instance"].log.error(f"Failed to send ETA:
```

Getting progress, ETAs, etc.

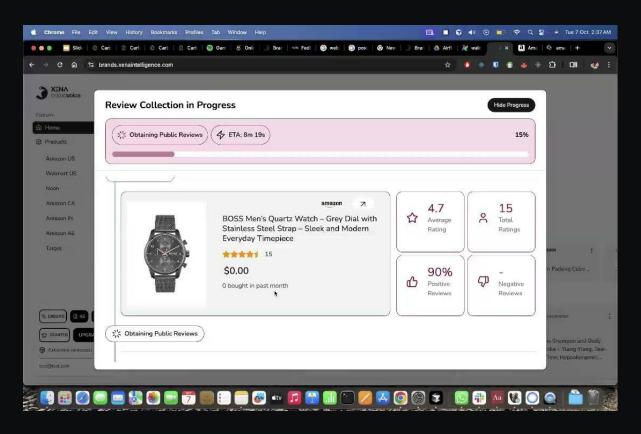


```
def analysis_task(productData, parsedReview, **kwargs):
    print(f"Analysis for review_id {review_id or "(No id)"}:
    analysis = get_analysis_from_llm(
        product_title=product_title,
        review_title=review_title,
        review_body=review_body
    send_event_to_webhook(
        payload={
            "event_type": "analysis_data",
            "data": analysis["micro_insight"]
        },
        product_id=product_id
```

Sending SSE Events

Integrating it all in UI/UX





The Impact: Stats



2.3+ Million reviews

2500+ reports generated

From
2-3
weeks/report
to 10 mins

I'm Excited for



New features of Airflow 3

Migration to Airflow 3

Human-in-the-Loop workflows



Questions?

