# **Economic Analysis of LLM Inference Costs** (2023–2026)

## Introduction

Large Language Models (LLMs) such as those developed by **OpenAl** and **Anthropic** are experiencing massive adoption. Yet, their operation relies on an expensive infrastructure: high-end GPUs, high electricity consumption, and limited energy optimization in single-agent usage.

The objective of this project is to:

- Assess the evolution of inference costs over time.
- Identify **break-even thresholds** for different subscription profiles.
- Propose a **5–10 year projection** of costs and the subscription prices required for profitability.

## Methodology

#### 1. Data sources:

- **EIA** (U.S. Energy Information Administration) for commercial electricity prices.
- GPU pricing estimates (H100, L4) via **public markets** and manual overrides.
- Internal calculations: energy consumption, PUE (Power Usage Effectiveness), throughput (tokens/sec).

#### 2. ETL pipeline:

- Extraction (Python, EIA API + GPU scrapers).
- Transformation (cleaning, temporal harmonization, GPU/electricity cost integration).
- Loading into **PostgreSQL** via \copy.

#### 3. Analysis:

- Calculation of **cost per million tokens** (electricity + GPU).
- Definition of subscription profiles:
  - Lite: 200k tokens/month.
  - **Standard**: 1M tokens/month.
  - **Pro**: 5M tokens/month.
- Application of a **70% target margin** to estimate break-even prices.

# **Analysis**

## 1. Observed costs (2023-2026)

- The average cost per million tokens ranges between 1.8 and 2.2 USD depending on the period.
- Costs fluctuate with:
  - Changes in electricity prices (EIA source).
  - Adjustments to GPU hourly rates (H100, L4).

### 2. Break-even thresholds

- Lite (200k tokens/month): break-even ≈ 1.2–1.4 USD/month.
- Standard (1M tokens/month): break-even ≈ 6-7 USD/month.
- Pro (5M tokens/month): break-even ≈ 31–34 USD/month.

These levels remain **far below current subscription prices** (ChatGPT Plus: 20 USD/month, Claude Pro: 20 USD/month).

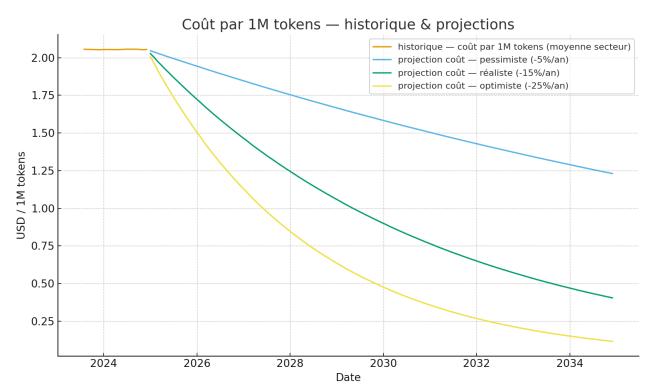
This indicates that companies are massively subsidizing access, despite high costs (and here we only consider electricity—other factors also come into play).

#### 3. Structural deficit

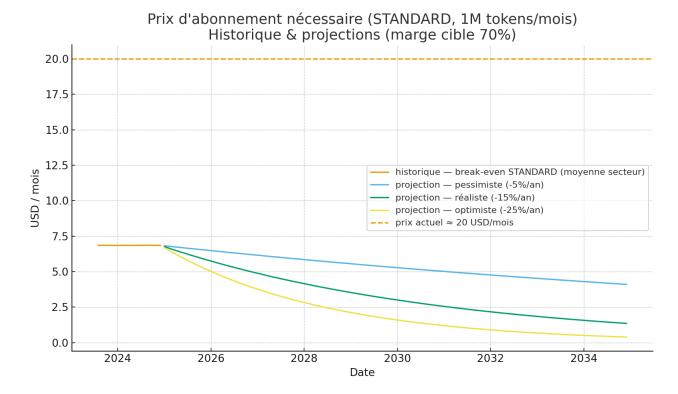
- Even without including **R&D**, **salaries**, **servers**, **and storage**, operations remain **unprofitable**.
- Current subscription prices do not cover the real marginal cost of inference.

# **Visualizations (5–10 years)**

1. Cost per million tokens curve (2023–2026).



2. **Hypothetical subscription price curve** required to reach profitability.



(Charts generated via matplotlib, available in the Python notebook.)

## **Discussion**

Why do OpenAI and Anthropic continue despite the losses?

- **Network effect**: more users = more data = better models.
- Strategic race: Al is a winner-takes-all sector; leadership matters more than immediate profitability.
- Massive subsidies: Microsoft, Google, and Amazon heavily fund these players.
- Bet on the future: GPU/energy costs may decrease, or subscription prices may rise.
- Market entrenchment: becoming indispensable.

In the long run, it is likely that:

- Subscriptions will **gradually increase** (25–40 USD/month).
- Offers will become more **segmented** (Lite, Pro, Enterprise).
- Companies will rely on ancillary revenues (API, integrations, SaaS products).

## **Conclusion & outlook**

- LLMs are currently operated at a loss, even when considering only GPU + electricity.
- Long-term economic viability will depend on:
  - Higher subscription prices.
  - Energy optimization (PUE, specialized chips).
  - Scale effects and throughput improvements.

• Our projections indicate a **necessary increase in subscription fees within 5–10 years**, otherwise losses will become unsustainable.

## **Next steps:**

- Extend projections to 2030–2035 with optimistic/pessimistic scenarios.
- Broaden the analysis to other players (Mistral, Meta, Google DeepMind).
- Simulate the impact of a carbon tax on the final cost of LLMs.