

What is Generative AI?

Understanding the Technology Behind the Hype

UBUS 670 | AI for Business Leaders

Day 1 • Week 1 • Spring 2026



Today's Learning Objectives

By the end of today, you will be able to:

1. Explain how LLMs work at a conceptual level
2. Define tokens, context windows, and temperature
3. Identify hallucination risks and mitigation strategies
4. Navigate Google Gemini Chat for business tasks

Today's Tool:

Google Gemini

gemini.google.com

Your Mission: Beacon Retail Group

You've joined the AI Strategy Task Force advising CEO Pat Holloway on AI adoption opportunities.

- Regional retail chain, 25 stores
- 1,200 employees, \$312M revenue
- Founded 1987 in Rockford, IL
- Competing with Amazon, Walmart, Target

Your role: Evaluate where AI can create real business value—and where it can't.



BEACON
RETAIL GROUP

"Lighting the way since 1987"

Beacon's Three Challenges



HR: Seasonal Hiring

4,200

applications/year

6 weeks to screen. 42% turnover rate.
\$2,500 cost per bad hire.



Marketing: Customer Service

850

emails/week

4-hour response time. 60% are routine inquiries.



Finance: Expenses

1,200

reports/month

8-day processing. 4% manual entry errors.

Can AI help? Let's find out what it actually is first.

Part 1

What is Generative AI?

Three Types of Software

Traditional Software

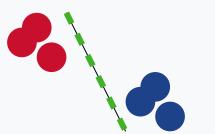
Rule-based

```
IF balance < 0  
THEN show_warning()
```

Does exactly what programmers coded.
Predictable but inflexible.

Machine Learning

Pattern-finding



10,000 examples → Learns patterns

Finds patterns in historical data. Great for classification.

Generative AI

Content-creating



Creates NEW content

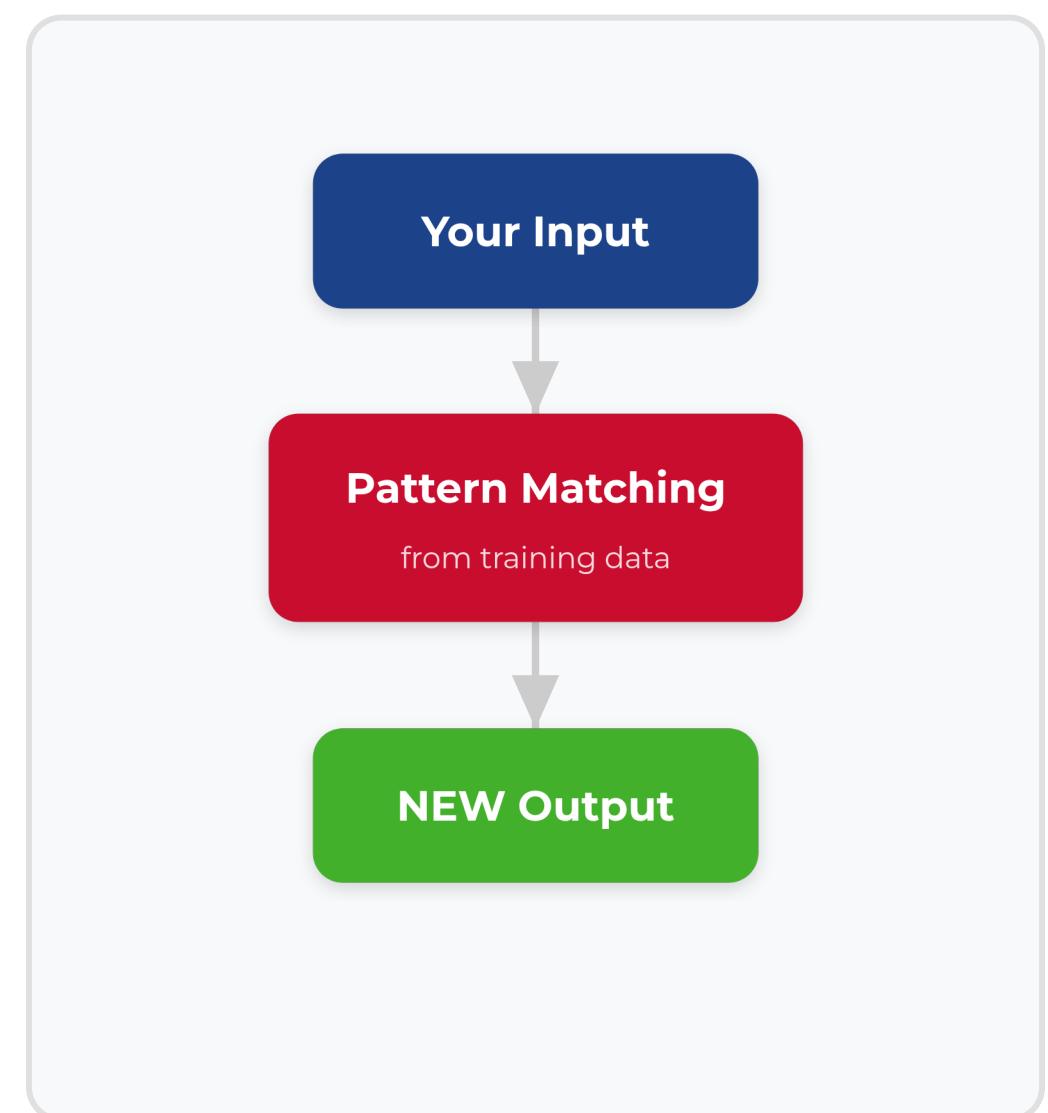
Produces novel outputs: text, images, code, ideas.

The Key Mental Model

Generative AI is a Synthesis Engine

It doesn't search for pre-existing answers.
It doesn't retrieve from a database.
It generates new content by predicting what comes next.

Think of it as an extremely well-read assistant who writes new content based on patterns learned from billions of documents.



Part 2

How Do LLMs Actually Work?

What is a Large Language Model?

Definition

A Large Language Model (LLM) is a deep learning model trained on massive amounts of text data that can understand and generate human-like language.

Key characteristics:

- Large — Billions of parameters (GPT-4: ~1.7 trillion)
- Pre-trained — Learned patterns from internet-scale text
- General purpose — Can handle many tasks without task-specific training

Training Data Scale

300B+

words in training data

1T+

parameters (model weights)

For comparison: A human reads ~1B words in a lifetime

Four Key Concepts to Understand

1 Tokens

How text is broken into chunks the model can process

2 Embeddings

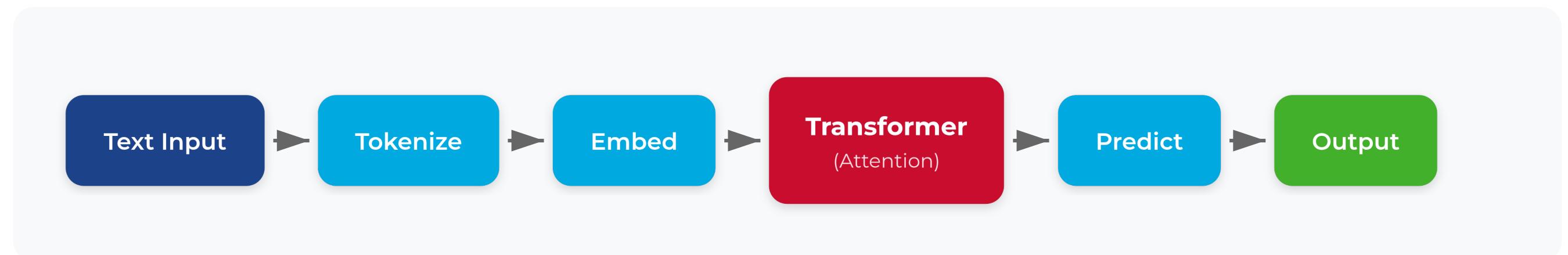
How words become numbers with meaning

3 Attention

How the model connects related words

4 Prediction

How responses are generated token-by-token



Step 1: Breaking Text into Tokens

LLMs don't read words—they read tokens.
A token is a chunk of text, roughly:

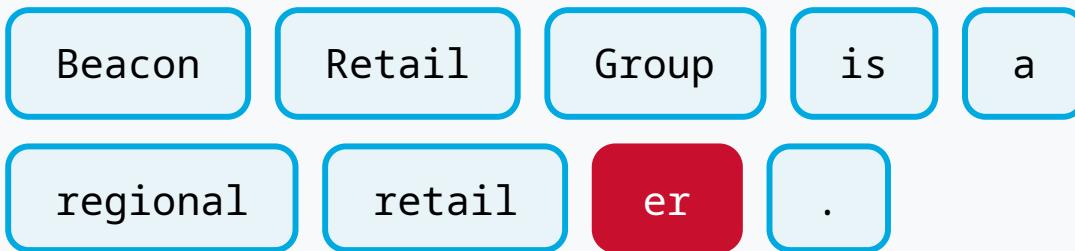
- ~4 characters, or
- ~0.75 words

Why does this matter?

You pay per token and context limits are measured in tokens.

Example:

"Beacon Retail Group is a regional retailer."



= 9 tokens (note: "retailer" splits into "retail" + "er")

Real Cost Example: Beacon's Emails

Let's calculate what it would cost Beacon to use AI for customer email responses:

- 850 emails per week
- Average email: 200 words (~270 tokens)
- Average response: 150 words (~200 tokens)

Bottom Line

AI email processing could cost Beacon \$5-170/month depending on the model—far less than one employee's time.

```
// Weekly email volume  
emails = 850  
tokens_per_email = 270 + 200  
weekly_tokens = 399,500  
// Monthly cost (4 weeks)  
monthly_tokens = 1,598,000  
// At Gemini Pro pricing ($0.00125/1K)  
gemini_cost = $2.00/month  
// At GPT-4 pricing ($0.03/1K input)  
gpt4_cost = $47.94/month
```

Step 2: Words Become Numbers (Embeddings)

Computers can't understand words directly. Embeddings convert tokens into lists of numbers (vectors) that capture meaning.

The Magic of Embeddings

Similar words have similar numbers. "King" and "Queen" are closer together than "King" and "Pizza."

Why this matters for business:

- AI understands "complaint" and "unhappy customer" are related
- Search can find similar products even with different words
- Enables semantic understanding, not just keyword matching

Simplified Example:

"King" → [0.8, 0.2, 0.9, ...]

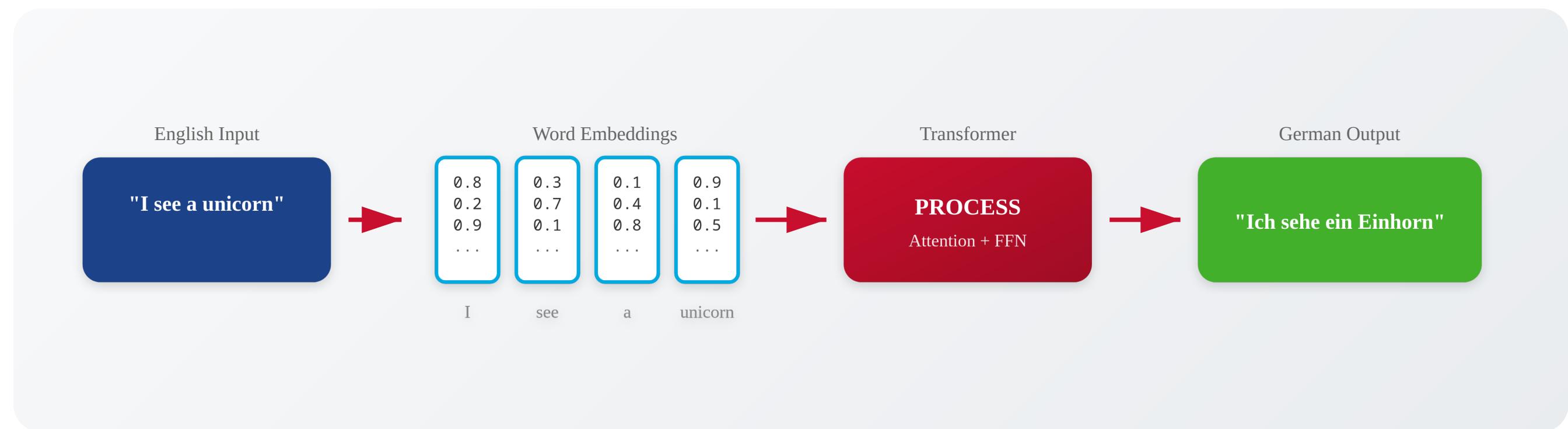
"Queen" → [0.7, 0.3, 0.9, ...]

"Pizza" → [0.1, 0.9, 0.2, ...]

Real embeddings have thousands of dimensions, not just 3!

How AI Understands Language

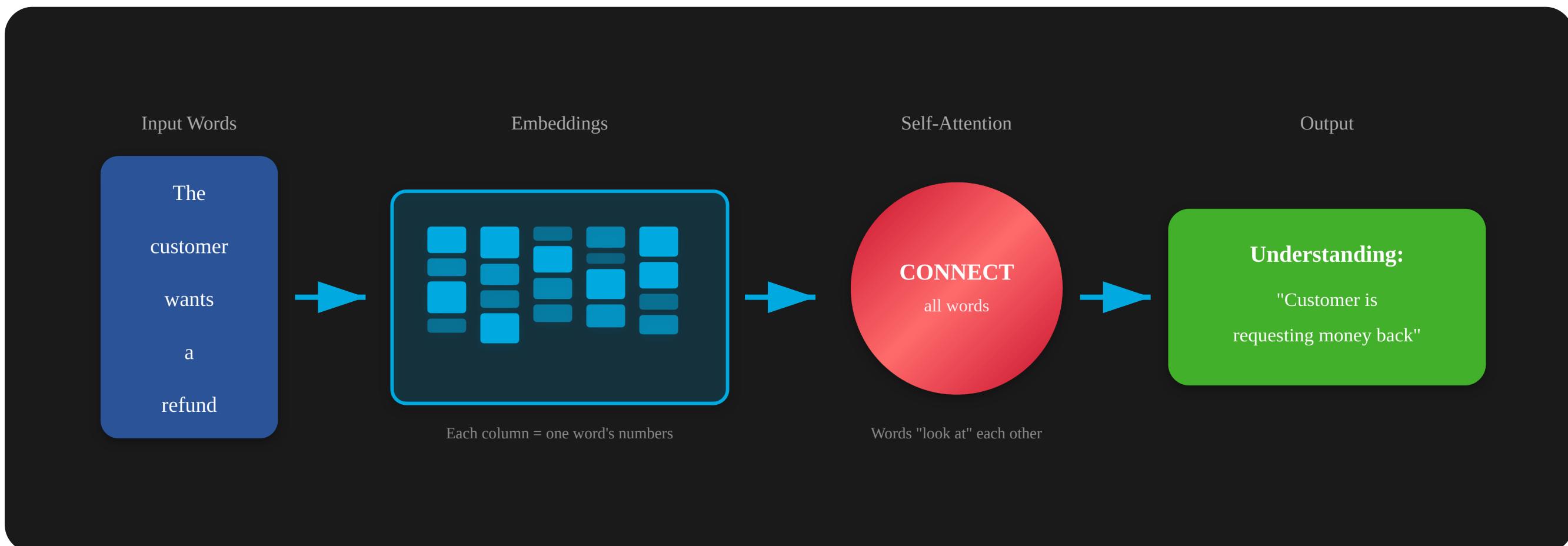
Let's see how a transformer processes a simple sentence:



Key insight: Words become numbers (embeddings), get processed through the transformer, and new words are generated. The model learned these patterns from billions of translated sentences.

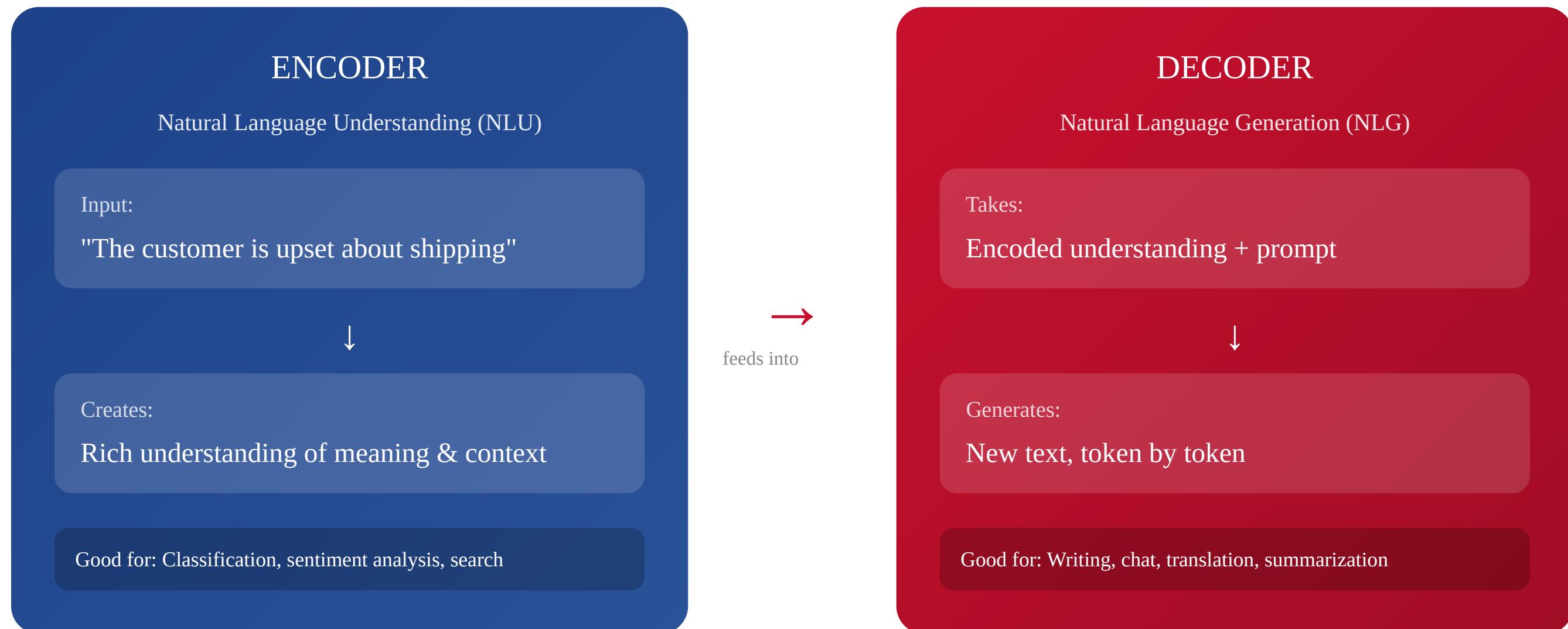
How a Transformer Works

The transformer architecture processes all words simultaneously:



Transformer Architecture: Encoder & Decoder

Modern LLMs use different combinations of encoders and decoders:



ChatGPT, Gemini, Claude: Primarily use decoders (GPT = "Generative Pre-trained Transformer")

BERT (Google Search): Uses encoder only for understanding queries

Step 3: Attention — Understanding Context

The Transformer architecture uses "attention" to understand how words relate to each other.

Example

"The bank was crowded, so I used the ATM."

Attention connects "bank" to "ATM" to understand it's a financial institution, not a riverbank.

This is revolutionary because:

- Previous AI processed words in order (slow, forgetful)
- Attention looks at ALL words simultaneously
- Enables understanding of long documents

Attention Visualization

The diagram shows a sequence of words: The, customer, was, frustrated, with, the, service. The words 'customer', 'frustrated', and 'service' are highlighted with darker shades of pink, indicating they are the most attended to in the context of predicting the word 'complaint'.

Darker = stronger attention when predicting "complaint"

Step 4: Predicting the Next Token

An LLM is Autocomplete on Steroids

Given some text, it predicts the most likely next token.

It does this one token at a time, thousands of times, to generate a complete response.

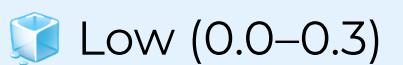
This is why AI can sometimes produce plausible-sounding nonsense—it's optimizing for "likely" not "true."

Input: "The quarterly sales report shows that revenue"

increased	34%
decreased	28%
grew	15%
fell	12%
...thousands more	11%

The Temperature Dial

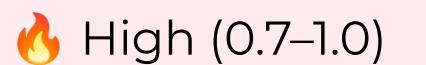
Temperature controls how the model samples from those probabilities.



Low (0.0–0.3)
Predictable
Always picks highest probability
Use for: Facts, data extraction, classification



Medium (0.4–0.6)
Balanced
Some variety, mostly consistent
Use for: Email drafts, summaries



High (0.7–1.0)
Creative
Willing to pick less likely tokens
Use for: Brainstorming, marketing copy

Warning: Higher temperature = higher hallucination risk. For business tasks with factual requirements, keep temperature low.

The Context Window

Think of it as "Working Memory"

The context window is everything the AI can "see" at once.

What's in the context:

- Your instructions (the prompt)
- Any documents you provide
- The conversation history
- The response being generated

Critical insight: The AI has NO memory beyond the context window. Each new conversation starts completely fresh.

System Prompt

Your instructions...

Documents

Uploaded files...

Conversation

Back and forth...

Current Response

Being generated...

Limited capacity (128K – 2M tokens)

Context Window Sizes (2026)

Model	Context Window	Roughly Equivalent To	Cost Impact
Gemini 1.5 Flash	1 million tokens	~10 novels	\$ (cheapest)
Gemini 1.5 Pro	2 million tokens	~20 novels	\$\$
GPT-4 Turbo	128K tokens	~1 novel	\$\$\$
Claude 3 Opus	200K tokens	~1.5 novels	\$\$\$

Tip: Larger context ≠ always better. Larger contexts cost more, and models can "lose focus" in very long contexts. Use only what you need.



Checkpoint: Test Your Understanding

Before we continue, let's make sure these concepts are clear. Click an answer to check.

Q1: If a business email is 200 words, approximately how many tokens is that?

A) 50 tokens

B) 200 tokens

C) 270 tokens

D) 800 tokens

Q2: You asked Gemini about a document yesterday. Today you start a new chat and ask a follow-up question. Gemini doesn't remember the document. Why?

A) New conversations start with an empty context window

B) The document was too long

C) There's a bug in Gemini

Part 3

The Hallucination Problem

What is a Hallucination?

Definition

A hallucination is when AI generates content that sounds confident and plausible but is factually incorrect, made up, or inconsistent with provided information.

Why does this happen?

- LLMs predict likely text, not true text
- No built-in "I don't know" instinct
- Confidence doesn't correlate with accuracy
- Training data may contain errors or outdated info

Invented Citation

"According to Smith & Jones (2024)..."

Paper doesn't exist.

Confident Wrongness

"Founded in 1982 by Michael Beacon..."

Wrong year, founder, and city.

Three Types of Hallucinations

Factual Fabrication

Making up facts that sound true

"The CEO of Walmart, James Robertson, announced..."

There is no James Robertson at Walmart. The AI invented a plausible-sounding name.

Source Fabrication

Inventing citations and references

"Research by MIT (Chen et al., 2023) found that 78% of companies..."

Paper doesn't exist. Statistics are fabricated.

Context Contradiction

Contradicting information you provided

*You: "Our budget is \$50K"
AI: "With your \$75K budget..."*

AI ignores or misremembers context you provided.

Real-World Hallucination Disasters



Lawyers Cited Fake Cases (2023)

A New York law firm used ChatGPT to write a legal brief. The AI invented 6 fake court cases with made-up citations.

Result: Lawyers sanctioned, fined \$5,000, case dismissed.



AI News Errors (Ongoing)

CNET published AI-written articles containing basic math errors and factual mistakes about financial products.

Result: Public apology, articles corrected or removed.

The Lesson

AI can produce content that passes a quick glance but fails under scrutiny. Always verify outputs before using them professionally.

Mitigating Hallucinations

Provide Source Documents

Ground the AI in YOUR specific facts. Instead of asking about Beacon in general, paste the actual company data.

Use Lower Temperature

For factual tasks, use temperature 0.0-0.3. Higher creativity = higher hallucination risk.

Ask for Uncertainty

Add: "If you're not sure about something, say so rather than guessing."

Always Verify

Never trust AI output for critical decisions without human verification.

Rule of thumb: The more critical the decision, the more verification you need. AI can draft, but humans must validate.

Part 4

What AI Cannot Do (Yet)

Current Limitations of Generative AI

🚫 No Real-Time Information

Models have a training cutoff date. They don't know about events after that date unless you provide the information.

🚫 No True Reasoning

LLMs simulate reasoning through patterns. Complex logic, math, and multi-step planning can fail unpredictably.

🚫 No Persistent Memory

Each conversation is independent. The AI doesn't remember you, your company, or previous conversations.

🚫 No Guaranteed Accuracy

AI can be confidently wrong. There's no internal "fact checker." Verification is YOUR responsibility.

The upside: Many of these limitations are being actively addressed. But for now, design your AI workflows with these constraints in mind.

Part 5

Meet Google Gemini

Why We're Using Gemini

Perfect for Learning

- Free tier — No cost for coursework
- No credit card required — Just a Google account
- Large context window — Handle long documents
- Multimodal — Text, images, PDFs, and more

Our Tool Progression

- Week 1: Gemini Chat (prompt basics)
- Week 2: Google Opal + AI Studio
- Week 3: Multi-Agent Systems

Already used ChatGPT? Great! The skills transfer. Gemini's interface is similar, but we'll explore features specifically useful for business tasks.

Gemini vs. ChatGPT: What's Different?

If you've used ChatGPT before, here's what to expect with Gemini:

✓ Similar

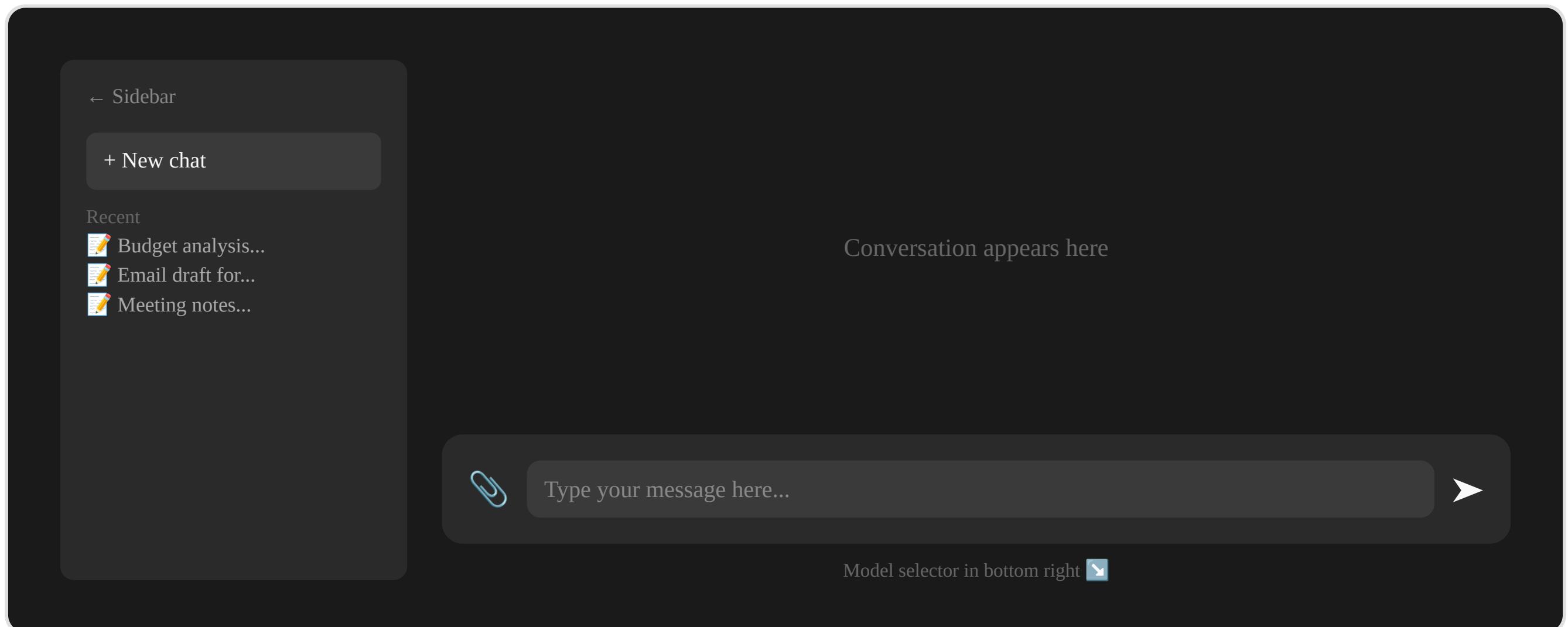
- Chat-based interface
- Follow-up questions in same conversation
- Can upload files and images
- Conversation history saved

NEW Different

- Google account login (not OpenAI)
- Integrates with Google Drive, Docs, etc.
- Different model versions available
- May give different answers than ChatGPT

Key insight: Different AI tools have different "personalities" and strengths. Learning multiple tools makes you more versatile.

Gemini Interface Tour



Key Gemini Features for Business



Upload Documents

PDFs, spreadsheets, images—
Gemini can read and analyze them



Iterate & Refine

Ask follow-up questions to improve
outputs



Copy & Export

Easy to copy responses or export to
Google Docs



Model Selection

Choose between speed (Flash) vs.
capability (Pro)

Pro tip: Start with the default model. If responses seem shallow or miss nuance, try switching to a more capable model.

Getting Started with Gemini

1. Go to gemini.google.com

Works in Chrome, Firefox, Safari, Edge

2. Sign in with your Google account

Personal account is fine—no upgrade needed

3. Start typing in the message box

No special commands required

4. Press Enter or click Send

Watch Gemini generate a response

Ready to try it?

[Open Gemini →](#)

You'll set this up properly in today's lab

Key Takeaways

1. Generative AI creates new content by predicting the next token based on patterns in training data.
2. Tokens = cost + limits. Understanding tokens helps you manage expenses and work within constraints.
3. Context window = working memory. The AI only knows what you put in the current conversation.
4. Temperature controls creativity vs. predictability. Use low for facts, high for brainstorming.
5. Hallucinations are inevitable. Always verify important information—never blindly trust AI output.

Up Next: Hands-On Lab

In today's lab, you will:

1. Set up your Google Gemini account
2. Complete three Beacon business tasks:
 - o Summarization
 - o Q&A with context
 - o Email drafting
3. Trigger and document a hallucination
4. Reflect on what you learned

Estimated Time

90-120

minutes

Start Lab →

Questions?

Before we move to the lab...