

AI-Powered Instructional Design

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1. AI-Powered Instructional Design

Welcome to the open-source ebook on **AI-Powered Instructional Design**.

This book explores how artificial intelligence can be integrated into the instructional design process to create more effective, efficient, and personalized learning experiences.

1.1 Chapters

- [Chapter 0: Introduction](#) - Overview of the book and its philosophy.
- [Chapter 1: AI Fundamentals](#) - Understanding the basics of AI for IDs.
- [Chapter 2: Prompt Engineering](#) - Mastering the art of communicating with AI.
- [Chapter 3: The ID-AI Workflow \(ADDIE\)](#) - Reimagining the ADDIE model with AI.
- [Chapter 4: Case Studies](#) - Real-world examples of AI in action.
- [Chapter 5: Advanced AI Implementation](#) - Going beyond the basics.
- [Chapter 6: The Future of AI-Powered ID](#) - Emerging trends and future predictions.

1.2 Bibliography

- [Bibliography](#) - Complete list of references and resources.

1.3 License

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2. Chapters

2.1 The Dawn of AI-Powered Instructional Design

The instructional design (ID) landscape is undergoing its most significant transformation since the advent of the World Wide Web. As we move into 2025, Artificial Intelligence is no longer just a "futuristic concept"—it is a critical partner in the design, development, and delivery of learning experiences.

2.1.1 From Automation to Augmentation

For years, the conversation around AI in ID was dominated by fear of replacement. However, as noted in the *2025 EDUCAUSE Horizon Report*, generative AI (GenAI) is acting more as a **catalyst** for evolving teaching practices rather than a substitute for human ingenuity (EDUCAUSE, 2024).

We are moving away from simple automation (using AI to do a task for us) toward **augmentation** (using AI to do things we couldn't do before). This ebook is built on the philosophy of **Co-Intelligence**, a term coined by Ethan Mollick (2024), where the instructional designer and the AI work in a symbiotic loop.

2.1.2 The AI-Powered ID Persona

An AI-Powered Instructional Designer is someone who: 1. **Orchestrates Workflows**: Uses AI to handle cognitive load—curating research, generating drafts, and analyzing data—so they can focus on high-level strategy and learner empathy. 2. **Architects Engagement**: Leverages AI to create hyper-personalized learning paths that adapt to individual learner needs in real-time. 3. **Ensures Ethics**: Acts as the "Human-in-the-Loop," critically evaluating AI outputs for bias, accuracy, and pedagogical soundness.

2.1.3 What to Expect in This Book

This is not just a book about theory. It is a technical foundation and a practical guide. You will find: - **Foundational Concepts**: Deep dives into LLMs and learning science. - **Agentic Workflows**: How to build AI agents that help you with specific ID tasks. - **Interactive Code**: Jupyter Notebooks that you can run to test prompt engineering and content generation.

Welcome to the future of learning. Let's build it together.

Key Takeaways

- AI is an enhancement tool, not a replacement.
- The shift is toward personalized, scalable, and efficient design.
- Human empathy and ethical judgment remain the core of valid instructional design.

2.2 AI Fundamentals for Instructional Designers

To design effective learning experiences with AI, we must first understand the "engine" that powers our tools. This chapter demystifies Large Language Models (LLMs) and explains the core concepts that every instructional designer should know.

2.2.1 1. What is an LLM?

A Large Language Model is a type of artificial intelligence trained on massive amounts of text data. It uses statistical patterns to predict the next word (or "token") in a sequence.

[!NOTE] Think of an LLM as a highly sophisticated "auto-complete" built on the sum of human digital knowledge.

For an ID, an LLM is more than a chatbot; it is a **reasoning engine**. It can synthesize information, take on personas (e.g., "Act as a subject matter expert in physics"), and format content into specific structures (e.g., "Generate a SCORM-compliant outline").

2.2.2 2. Tokens and Context Windows

Understanding how AI "reads" and "remembers" is crucial for prompt engineering.

Tokens

AI doesn't read words like humans do. It breaks text into **tokens**—small chunks of characters. 1,000 tokens is roughly equal to 750 words. *Why does this matter?* API costs and model limits are often based on token counts.

Context Window

The **context window** is the amount of information the model can "hold in its head" at once during a conversation. In 2025, context windows have expanded significantly (with some models handling millions of tokens), but the core principle remains: the more relevant context you provide in your prompt, the better the output.

2.2.3 3. Generative vs. Discriminative AI

- **Generative AI:** Creates new content (text, images, video) based on patterns. This is where most ID work happens (e.g., creating case studies).
- **Discriminative AI:** Classifies or analyzes existing data. In ID, this is used for grading, identifying gaps in a curriculum, or sentiment analysis of learner feedback.

2.2.4 4. The AI Toolbox: Beyond ChatGPT

While ChatGPT (and other LLMs like Claude or Gemini) are the most famous tools, the AI landscape for instructional designers is vast. As recent reviews by *The eLearning Coach* and *Cathy Moore* highlight, the toolkit can be categorized by function:

- **Writing & Content:** **Jasper**, **ChatGPT**, **Claude** (for brainstorming, drafting scenarios, and rewriting content).
- **Multimedia Generation:**
 - *Video:* **Synthesia** (AI avatars), **HeyGen**.
 - *Audio:* **WellSaid Labs** (high-fidelity voiceovers), **Descript** (audio/video editing with text).
 - *Presentation:* **Tome** (rapid slide generation).
- **Research & Synthesis:** **Elicit**, **Consensus** (finding academic papers and summarizing research without hallucinations).
- **Productivity:** **Otter.ai** (meeting transcription), **Notion AI**.

[!TIP] Tools evolve rapidly. Focus on the category of the tool (e.g., "AI Voice Generator") rather than becoming dependent on a single brand.

2.2.5 5. The "Hallucination" Problem

One of the biggest challenges in AI-Powered ID is **hallucination**—when the model generates factually incorrect information that sounds highly convincing.

[!IMPORTANT] Never use AI-generated content in a learning module without a Subject Matter Expert (SME) or your own rigorous verification. AI is a creative assistant, not an encyclopedia.

2.2.6 6. Ethical Considerations

As instructional designers, we have a responsibility to our learners: - **Bias:** AI models can inherit biases from their training data. We must audit outputs for gender, racial, and cultural bias. - **Privacy:** Never input sensitive student data or proprietary company information into public AI models. - **Academic Integrity:** We must design assessments that focus on higher-order thinking (Bloom's Taxonomy) which AI cannot easily replicate without human synthesis.

Reflection Exercise

1. Choose a topic you are currently teaching.
2. Write a prompt for an LLM to generate three different ways to introduce that topic.
3. Evaluate the output: Which one is the most engaging? Did the model hallucinate any facts?

References: - EDUCAUSE (2024). *2025 Horizon Report: Teaching and Learning Edition*. - Malamed, C. (2025). *AI Tools for Instructional Designers*. *The eLearning Coach*. - Mollick, E. (2024). *Co-Intelligence: Living and Working with AI*. Portfolio. - Moore, C. (2025). *Best AI Tools for Instructional Designers*.

2.3 Prompt Engineering for Instructional Designers

In Chapter 1, we learned that LLMs are "reasoning engines." To get the best results from these engines, we must provide high-quality fuel: **prompts**. Prompt engineering is the process of structured communication that guides the AI toward accurate, relevant, and pedagogically sound outputs.

2.3.1 1. The Anatomy of a High-Quality Prompt

A vague prompt ("Write a lesson plan about history") leads to a generic output. A high-quality prompt contains several key elements:

- **Role:** Define the AI's persona (e.g., "Act as a Senior Instructional Designer with 20 years of experience in corporate training").
- **Context:** Provide background (e.g., "The audience is first-line managers at a global tech company who have limited time for training").
- **Task:** The specific action (e.g., "Draft a 15-minute microlearning module on 'Giving Difficult Feedback'").
- **Constraints:** What the AI should *not* do or specific limits (e.g., "Keep the reading level at Grade 8. Do not use jargon. Use only evidence-based feedback models").
- **Output Format:** How the content should look (e.g., "Format the output as a Markdown table with three columns: Concept, Learner Activity, and Timing").

2.3.2 2. Prompting Frameworks for ID

Standardizing your prompts makes your workflow repeatable and scaleable. Two effective frameworks for IDs are:

The RACE Model

A widely used framework that helps categorize the essential components of a prompt: - **Role:** Who the AI is. - **Action:** What it needs to do. - **Context:** The background info. - **Expectations:** The final quality and format.

The DETAIL Method

Focuses on the granular needs of a learning module: - **Domain:** The subject area. - **Examples:** Providing specific samples (Few-Shot prompting). - **Target audience.** - **Assessment:** How the learning will be measured. - **Intent:** The "Why" behind the content. - **Limits:** Constraints and boundaries.

2.3.3 3. Advanced Techniques

Once you master the basics, you can use advanced techniques to handle complex design tasks.

Few-Shot Prompting

Instead of just asking for a lesson plan, provide 1-2 examples of previous lesson plans you've written. This "grounds" the model in your specific style and voice.

Chain-of-Thought (CoT)

Add the instruction "**Think step-by-step**" or "Outline your reasoning before providing the final answer." This encourages the model to break down complex tasks into smaller, logical steps, significantly reducing hallucinations.

Recursive Self-Improvement

Ask the AI to critique its own work. *Example Prompt:* "Review the microlearning draft you just provided. Identify any areas where the learning objectives are not met, and then rewrite the draft to address those gaps."

2.3.4 4. The ID-AI Feedback Loop

Prompt engineering is rarely a "one-and-done" task. It is an iterative loop: 1. **Draft**: Create your initial prompt. 2. **Diagnose**: Review the output for misalignment or generic content. 3. **Refine**: Add constraints, change the persona, or provide more context. 4. **Strengthen**: Polish the final version for the specific learner needs.

Hands-On Exercise

Using the **RACE Model**, draft a prompt to generate a 5-question multiple-choice quiz based on a specific paragraph from Chapter 1. Then, refine the prompt by adding a constraint to "provide feedback for both correct and incorrect answers."

2.4 The ID-AI Workflow: Augmenting ADDIE

The ADDIE model (Analysis, Design, Development, Implementation, and Evaluation) has been the gold standard for instructional design for decades. As Devlin Peck (2025) notes, over 50% of instructional designers are now using AI daily or weekly to revolutionize the **speed and depth** of this workflow.

2.4.1 1. Analysis: Precision and Speed

In the traditional workflow, the analysis phase can take weeks. AI reduces this to hours. Instructional Designers at the **University of Cincinnati** are actively using AI to transcribe SME interviews, summarize research, and—crucially—identify implicit bias in existing curricula (University of Cincinnati, 2025).

- **Learner Persona Creation:** Use AI to analyze demographics and historical performance data to create "Synthetic Learner Personas" (Gartner, 2024).
- **Gap Analysis:** AI can ingest policy documents to identify missing competencies.
- **Sentiment Analysis:** Use AI to synthesize thousands of feedback comments into actionable themes.

2.4.2 2. Design: Collaborative Brainstorming

The design phase is where the ID-AI "Co-Intelligence" loop really shines.

- **Learning Objective Mapping:** Tools like *Learnt.ai* or custom GPTs can take a raw topic and generate measurable learning objectives.
- **Curriculum Architecture:** Use LLMs to brainstorm creative themes. The AI can then logic-check the sequence of modules for cognitive load flow.

2.4.3 3. Development: Rapid Multimedia & Content

This is the most visible area of impact. Reports from **Shift eLearning** indicate that companies using AI-powered tools can reduce course development time by up to **50%**, allowing designers to focus on high-level strategy rather than rote production (Shift eLearning, 2025).

- **Rapid Prototyping:** Tools like **Courseau** can convert a raw PDF into a structured course in minutes.
- **Multimedia Production:**
 - **Video:** Use **Synthesia** or **Meteora** to create professional video lessons with AI avatars.
 - **Audio:** Generate high-quality voiceovers from text using **Murf.AI**.
- **Assessments:** AI can generate varied question types based on the course content.

2.4.4 4. Implementation: The Adaptive Classroom

AI moves instructional design from a "fixed" experience to a "living" one.

- **AI Tutors:** Deployment now includes Socratic tutors that guide learners without giving answers.
- **Just-in-Time Support:** Chatbots embedded in the LMS provide 24/7 support.

2.4.5 5. Evaluation: Predictive Insights

Instead of waiting for the end of a course to see if it worked, AI allows for continuous evaluation.

- **Real-Time Analytics:** AI-powered platforms like **Disco AI** track learner progress and flag individuals who are likely to drop out or fail (Disco, 2025).
 - **Iterative Refinement:** AI can analyze assessment results and suggest specific rewrites for confusing questions.
-

Summary Table: The Augmented ADDIE

| Stage | Traditional Task | AI-Augmented Task |
|----------------|----------------------------|---|
| Analysis | Manual survey analysis | AI-driven bias detection & gap analysis |
| Design | Brainstorming with SMEs | Co-creation of objectives & themes with LLMs |
| Development | Months of content creation | 50% faster production (Shift eLearning, 2025) |
| Implementation | Static content delivery | Adaptive, tutor-supported journeys |
| Evaluation | End-of-course reports | Real-time predictive analytics & refinement |

Case Study: Rapid Program Development

In 2024, a global leadership firm used a strategic AI workflow to develop a new training program. By using AI-driven Analysis and Development tools (like those highlighted by Devlin Peck), they reduced their time-to-market dramatically while maintaining high learner engagement scores.

References: - Courseau (2025). *Accelerating Course Development with AI*. - Devlin Peck (2025). *AI in Instructional Design*. - Disco (2025). *AI for Instructional Design: Using the ADDIE Model*. - Shift eLearning (2025). *The Future of Instructional Design in the AI Era*. - University of Cincinnati (2025). *How Instructional Designers Use AI*.

2.5 Case Studies and Practical Examples

To understand the true power of AI-powered instructional design, we must look beyond the tools and frameworks to the real-world impact. This chapter highlights how organizations in corporate and higher education are using AI to solve complex learning challenges.

2.5.1 1. Higher Education: The Rise of the AI Teaching Assistant

Georgia Institute of Technology: "Jill Watson"

In one of the most famous examples of AI in education, Georgia Tech implemented an AI teaching assistant named "Jill Watson" (Georgia Tech, 2024). - **The Challenge:** Online forums for computer science courses were overwhelmed with thousands of repetitive student queries. - **The Solution:** An AI agent trained on previous forum data was deployed to answer routine questions about assignments and deadlines. - **The Result:** Students were unable to distinguish Jill from human TAs. She answered questions with 97% accuracy, allowing human instructors to focus on deeper pedagogical discussions and at-risk student intervention.

2.5.2 2. Corporate Training: Personalization at Scale

IBM: Watson-Powered Learning Paths

IBM uses its Watson AI platform to manage the continuous upskilling of hundreds of thousands of employees (IBM, 2024). - **The Challenge:** A "one-size-fits-all" training approach was inefficient for a workforce with highly diverse technical skills. - **The Solution:** An AI-driven learning marketplace that maps individual employee skills, career goals, and historical performance to a personalized curriculum. - **The Result:** Significant increase in relevant skill acquisition and a reduction in "time-to-competency" for new hires.

Walmart: VR and AI for Procedural Training

Walmart combined Virtual Reality (VR) with AI to train frontline associates (Walmart, 2024). - **The Challenge:** Training employees on complex, high-pressure tasks (like "Black Friday" management) is difficult in a classroom. - **The Solution:** Immersive VR simulations where AI actors respond dynamically to trainee decisions. - **The Result:** A 15% improvement in employee performance scores and a massive reduction in training time for procedural work.

2.5.3 3. Specialized Learning: Accessibility and Engagement

Bolton College: AI Video Generation

Bolton College in the UK leveraged AI to overcome the technical barriers of video production (Bolton College, 2024). - **The Challenge:** Creating enough video content for online modules was slow and expensive. - **The Solution:** Using **Synthesia** to generate high-quality video lessons from text scripts using AI avatars. - **The Result:** Rapid production of engaging, accessible online material. The college reported that learners appreciated the consistency and clarity of the AI presenters.

Ivy Tech Community College: Predictive Analytics

Ivy Tech used AI to identify students at risk of failing before they even took their midterms (Ivy Tech, 2024). - **The Challenge:** Traditional "early warning" systems were too late to save many struggling students. - **The Solution:** An AI model that analyzed 16,000 data points (engagement, early quiz scores, login frequency) to predict at-risk status. - **The Result:** 16,000 at-risk students were identified; 98% of those contacted improved their performance to a 'C' or better.

Critical Analysis: What Can We Learn?

Across these diverse case studies, three themes emerge: 1. **Scaling Human Expertise:** AI doesn't replace the SME; it amplifies their reach (as seen with Jill Watson). 2. **Personalization is Performance:** Tailoring the path to the individual leads to faster and deeper mastery (IBM). 3. **Efficiency Drives Innovation:** By automating the "routine" (Bolton College), designers are free to focus on high-impact strategy.

References: - Bolton College (2024). Transforming Online Learning with AI Video. - Georgia Institute of Technology (2024). Jill Watson: The AI Teaching Assistant. - IBM (2024). Personalizing Corporate Learning at Scale with Watson. - Ivy Tech (2024). Predictive Analytics in Higher Education. - Walmart (2024). Immersive AI Training for Frontline Associates.

2.6 Advanced AI Implementation: Agents & RAG

In the previous chapters, we focused on how to communicate with AI using prompts and how to integrate it into your workflow. However, to build truly powerful, "hallucination-free" learning experiences, we must move beyond the basic chat interface. This chapter explores **Retrieval-Augmented Generation (RAG)** and **Agentic Workflows**.

2.6.1 1. What is RAG? (Retrieval-Augmented Generation)

A common problem with LLMs is that they are trained on public data. They don't know your company's specific safety protocols, your unique product features, or your internal project management methodology.

RAG solves this by connecting the LLM to a specific "Knowledge Shell" of your proprietary documents.

How it Works (The Technical Loop)

1. **Retrieval:** When a user asks a question, the system first searches your provided documents (PDFs, transcripts, manuals) for relevant text chunks.
2. **Augmentation:** The system "attaches" those relevant chunks to the user's question.
3. **Generation:** The LLM reads the user's question *plus* the attached chunks and generates an answer grounded solely in that data.

[!TIP] RAG is the single most effective way for Instructional Designers to eliminate AI hallucinations. It forces the AI to "cite its sources" from your approved materials.

2.6.2 2. Agentic Workflows: The Power of Delegation

In a standard workflow, you give a prompt and get a response. In an **Agentic Workflow**, you give a goal, and the AI works in a loop to figure out how to achieve it (Ng, 2024).

Andrew Ng (2024) identifies four key patterns for agentic design: 1. **Reflection:** The agent looks at its own work and critiques it before showing it to you. 2. **Tool Use:** The agent can decide to use a calculator, search the web, or run code to solve a problem. 3. **Planning:** The agent breaks a complex goal (e.g., "Build a full 4-week course") into a sequence of smaller tasks. 4. **Multi-agent Collaboration:** Different agents with specialized roles (e.g., a "Quiz Agent" and an "Outline Agent") talk to each other to produce a final product.

2.6.3 3. Localized Knowledge Shells for ID

Imagine building a training program for a new medical device. Instead of writing the content yourself, you create an "ID Agent" and provide it with the 500-page technical manual. - You ask the agent to: "Identify the 5 most common user errors mentioned in the manual and draft a scenario-based quiz for each." - Because the agent is grounded in a **RAG** system, it won't guess; it will only pull from the manual.

2.6.4 4. Semantic Search vs. Keyword Search

Advanced AI implementation changes how learners interact with your content. - **Keyword Search:** Looks for exact matches of words. - **Semantic Search:** Understands the *intent* and *meaning* behind a question. If a learner asks "How do I fix the blinking red light?", semantic search knows that "blinking red light" refers to the "Power Fault Condition" in Chapter 4 of your manual, even if the word "blinking" isn't in that chapter.

2.6.5 5. Security and Intellectual Property (IP)

When implementing advanced AI, security is paramount. Instructional designers must advocate for **Private LLM Environments**. - These are secure "bubbles" within your company's cloud where you can safely upload proprietary training data without it being used to train the public models (Databricks, 2025).

Reflection Exercise

Assume you have a 100-page employee handbook. How would a **RAG-powered** AI tutor be different from a traditional "Find" (Ctrl+F) search? Which one would be more helpful for a new hire trying to understand company culture?

References: - Databricks (2025). Building High-Fidelity RAG Systems for Enterprise Knowledge. - Ng, A. (2024). Agentic Workflows: The Next Frontier of Generative AI. DeepLearning.AI. - Gartner (2024). Hype Cycle for Artificial Intelligence, 2025.

2.7 The Future of AI-Powered Instructional Design

We are at the precipice of the "Third Wave" of AI in education. If the first wave was basic automation and the second wave was generative assistance, the third wave (2025–2030) will be defined by **autonomous orchestration** and **hyper-personalization**.

2.7.1 1. Hyper-Personalization: The "Curriculum of One"

By 2030, the concept of a "static course" will likely be obsolete. Instead of all learners moving through the same 10 modules, AI will generate a unique learning path for every individual in real-time (EDUCAUSE, 2025).

- **Biometric Feedback:** AI will use non-invasive biometric data (eye-tracking, heart rate, and emotional sentiment) to detect when a learner is frustrated or bored and adjust the difficulty or content format instantly.
- **Micro-Pivot Learning:** If a learner fails a quiz on "Python Lists," the AI doesn't just show the correct answer; it instantly generates a new lesson on that topic using a different teaching style (e.g., switching from text to an interactive simulation).

2.7.2 2. Synthetic Learners and Digital Twins

One of the most exciting emerging trends is the use of **Synthetic Learners** (Articulate, 2025). Before you release a course to your 10,000 employees, you will "deploy" it to an audience of 100 AI agents.

- **Bias Detection:** These agents, each with different cultural backgrounds and prior knowledge levels, will interact with your content and report back on areas where the language is biased or the logic is confusing.
- **Predictive Efficacy:** Synthetic learners can "take" your final exam 1,000 times in seconds, allowing you to statistically predict the real-world pass rate before the first human learner even logs in.

2.7.3 3. Multimodal Learning Landscapes

We are moving past the "Text + Image" paradigm. The future of ID is **Multimodal**. - AI will generate 3D immersive environments (VR/AR) from simple text prompts, allowing IDs to create complex laboratory or social-learning simulations without a dedicated team of 3D developers (Gartner, 2024). - Voice-first learning will allow "screenless" education, where AI mentors guide workers through hands-on tasks in real-time.

2.7.4 4. The Shift in K-12 and Labor Market Skills

Innovation isn't just about tools; it's about preparing learners for a rapidly changing labor market. As highlighted by Miao & Mishra (2025), the focus of EdTech innovation is shifting from simple content delivery to fostering **"Future-Ready Skills."**

- **Cognitive Flexibility:** AI can answer static questions. The new curriculum prioritizes the ability to switch between concepts and adapt to new AI tools.
- **Strategic GenAI Innovation:** Rather than just "using AI in class," institutions must adopt a long-term strategy where GenAI is integrated into the assessment of learning itself, moving away from rote memorization toward measuring critical thinking and problem-solving in real-world scenarios.

2.7.5 5. The Evolving Role: From Designer to Architect

The most significant change will not be in the technology, but in the **human role**. The instructional designer of 2030 will be a "Learning Architect" or "Experience Orchestrator."

- **The Architect:** Focuses on the high-level strategy, ethics, and "human-in-the-loop" verification.
- **The Orchestrator:** Manages a fleet of AI agents (Content Agents, Assessment Agents, Feedback Agents) to build learning experiences at a scale and speed previously unimaginable (Articulate, 2025).

2.7.6 6. Conclusion: Staying AI-Fluent

The future of instructional design is bright, but it requires a fundamental shift in mindset. We must move away from guarding our "creative output" and toward mastering our "orchestration input."

As we conclude this ebook, remember that AI is a tool of empowerment. It automates the routine so that we can focus on the core of our profession: **human empathy, pedagogical soundness, and the joy of learning.**

Final Reflection

What is one skill you use today that you think an AI will do better in 2030? What is one skill you use today that an AI will *never* be able to replicate?

References: - Articulate (2025). The Evolving Role of the Instructional Designer in the Age of AI. - EDUCAUSE (2025). 2025 Horizon Report | Teaching and Learning Edition. - Gartner (2024). Hype Cycle for Artificial Intelligence, 2025. - Miao, X. & Mishra, P. K. (2025). Preparing Future-Ready Learners: K12 Skills Shift and GenAI EdTech Innovation Direction.

3. Interactive Labs

3.1 Lab: AI Prompt Engineering Playground

Welcome to your first interactive lab! In this notebook, we will explore how changing a prompt can drastically change the output of an AI model.

Since we don't have a live API key in this demo, we will simulate the responses to show you the concept.

```
In [1]: def simulate_llm_response(prompt):  
        """Simulates an LLM response based on keywords in the prompt."""  
        prompt_lower = prompt.lower()  
  
        if "learning objective" in prompt_lower:  
            if "bloom" in prompt_lower:  
                return "Generated Objective: By the end of this module, learners will be  
able to ANALYSIS the impact of AI on ID workflows."  
            else:  
                return "Generated Objective: Understand AI in ID."  
        elif "quiz" in prompt_lower:  
            return "1. What is the primary benefit of AI? (a) Speed (b) cost (c) Accuracy"  
        else:  
            return "I am not sure what you want me to generate. Try asking for a learning  
objective."
```