



NEHIA/HFMA 2025 Compliance & Internal Audit Conference

Wednesday, December 3 - Friday, December 5, 2025 Mystic Marriott Hotel, Groton, CT

Generative AI – Deep Dive



Unlocking the Power of Language: Understanding Generative Al

WHAT is Generative AI | artificial intelligence that creates **original content across various modalities** (e.g., text, images, audio, code, voice, video) that would have previously taken human skill and expertise to create

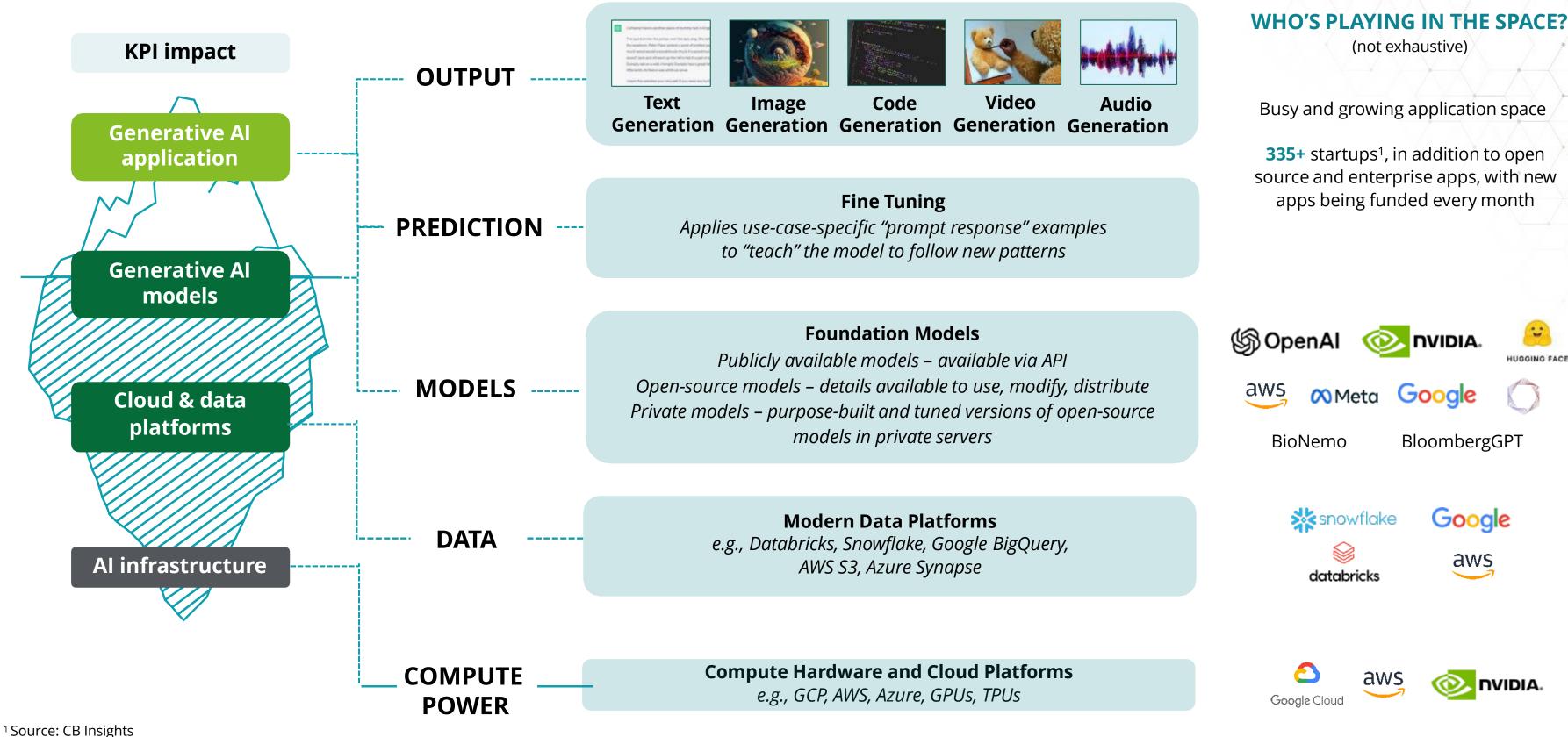
HOW does it work | Generative AI is powered by **large language models (LLMs)** such as OpenAI's GPT-4, NVIDIA's Megatron, and Google's PaLM, which are trained on **vast amounts of data and computation** to perform a broad range of downstream tasks

WHY now | the availability of LLMs and advanced computing power, coupled with the viral popularity of publicly released applications have propelled Generative AI into the zeitgeist

WHO is involved | **Big Tech** is building—and enabling access to—foundation models; **start-ups** are developing user applications on these underlying models; **companies** are beginning to adopt and the **average person** have begun to use apps

POTENTIAL BUSINESS IMPACT | the marginal cost of producing initial versions of knowledge-intensive content—such as IT code, marketing copy, and creative design—can decrease significantly

The Generative AI technology stack is made up of many key players



Prompt operations

There are 3 fundamental operations which form a key framework for understanding and utilizing prompts







Reductive Operations

These operations aim to produce a smaller output from a larger input. Examples include:

- **Summarization**: Condensing information into a shorter form.
- **Distillation**: Extracting the core underlying principle or fact from a text.
- Extraction: Pulling out specific information, such as answering questions or extracting dates and numbers from a text.
- **Characterizing**: Defining the nature of the text or the main topic within the text.

Transformational Operations

In these operations, the input and output are roughly the same size and meaning, but the form or structure is changed. Examples include:

- **Reformatting**: Changing the presentation of content, such as translating data between formats like XML and JSON.
- Refactoring: Rewriting code or text in a better or different way while preserving the original meaning.
- Language Change: Translating content between natural languages (e.g., English to Portuguese) or programming languages (e.g., C++ to Python).
- **Restructuring**: Altering the order of content, removing or adding sections, and optimizing for logical flow.

Generative Operations

These operations expand a small input into a much larger output. Examples include:

- Drafting: Providing instructions for a language model to create a document, such as a code file, legal document, or story.
- Planning: Giving a set of parameters to develop plans, such as action plans or project plans.
- Brainstorming: Using imagination to list possibilities, generate ideas, explore options, and solve problems.
- Hypothesizing: Generating hypotheses based on given information or a prompt.
- **Amplification**: Expanding on a topic to fully explore and explain it.

Common Risks with Generative Al

Bias in; bias out. If the training data is biased (e.g., over/under-representation of a population cohort, sexism, racism), then outputs generated could exhibit biases as well. Bias reductions in the training data and/or human supervision during model training is needed

Foundation Models generally offer a pay-as-you-go billing mechanism, and the cost per use of sophisticated models is materially significant. Fine tuning the biggest model and running large documents through several times could easily run up a bill of tens of US \$1000s

Is the AI being used in a manner consistent with the purpose of the overall exercise? Is a human being brought into the loop to decide whether the AI's suggestion needs adjustment before actual use? Submitting an AI-generated essay for a high-school assignment may not be ethical

Models might output facts that are factually false. Sources and citations are unavailable for most models. Users should be conscious that outputs could be inaccurate and should perform due diligence to validate generated content.

SaaS-AI companies may save some or all of prompt payloads for future training. Therefore, confidential data will be used to train future versions of the base model – how will this affect your organization's competitiveness in the market?

Bias



Cost





Hallucination



IP Protection



It is critical to proactively minimize risk from malicious behavior on the network to maintain operations and customer trust. For example, a customer service bot revealing confidential information to a hacker either by prompt or unintentionally

Foundation Models are comprised of billions of parameters (model size) and trained on petabytes of data. In theory, the larger the model, the better the output. Foundation Models take time to produce outputs, which may limit real-time use cases

SaaS-Al companies require to submit text as a payload to users' API call. The data could be crossing borders. Is this in accordance with data privacy laws and with your company's policies? Many cloud service providers offer market-leading controls to manage data privacy of Foundation Models

Models are good at understanding text but struggle when the data are in irregular formats, or when the position of the text on the page (e.g., infographic, PPT presentation slide) is relevant to the context and understanding. Other emphasis generators such as bolded text, font color, etc., don't play a role yet

Most models have a 2k token size limit. Some larger ones can process 4k tokens in a single call. 2k tokens are approximately 2-2.5 pages. This limit makes it difficult to process larger documents

Malicious behavior



Model Performance



Privacy



Text Formatting



Token Size Limits



6

Meet your panel



Director, Compliance Coding, Billing & Audit

Mass General Brigham



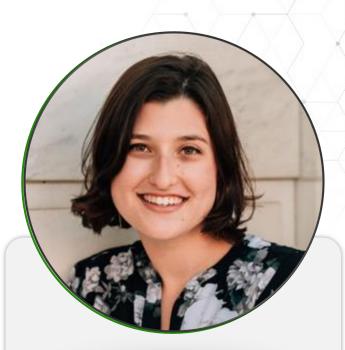
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QUESTIONS



