

Unveiling the Lab of the Future with an Al-Powered LIS

Revolutionizing Workflows with Al

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May 1, 2024

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Unveiling the Lab of the Future,

with Gen Al-Powered LIS

Agenda

AI Evolution

Classic Lab Challenges

Gen Al-powered LIS Potential

Challenges and Considerations

Clinisys Gen Al Usage Roadmap

Gen Al-Powered Lab of the Future

Q&A and Discussion



Al Evolution: fueled by storage, compute advancements & \$\$

"Artificial Intelligence" – 1956

'AI' term coined at Dartmouth workshop

Rule-Based Systems – 1960s & 1970s

Example: ELIZA Chatbots mimics human communication

Expert Systems – 1980s

Example: Carnegie Mellon's R1 configuration tool used by Digital Equipment Corp

Machine Learning – 1990s

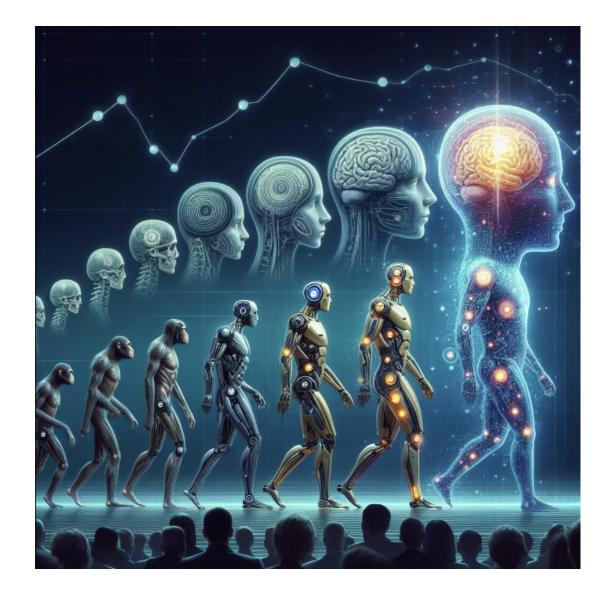
Example: IBM's Deep Blue beats world Chess Champion, Garry Kasparov

Deep Learning – 2000s & 2010s

Examples: Natural Language Processing – IBM Watson wins Jeopardy!

Large Language Models (LLMs) and Gen AI – 2020s

Examples: Chat GPT, DALL-E



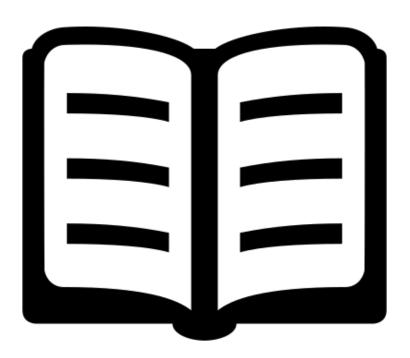


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LLMs and GenAl Terminology

- **Bias** is the tendency for AI to replicate the prejudices and disparities inherent in the data it was trained on
- **ChatGPT** is arguably the most popular generative AI system out there at the moment. It takes OpenAI's foundation model, GPT, and tunes it for use in a chat interface
- **Fine tuning** is the act of further training a foundation AI model in order to make it more suited to a specific task
- **Hallucination** is the tendency for LLMs to produce coherent but factually incorrect text
- Prompt is the text a user provides to a generative AI to begin its output
- Prompt engineering is the practice of experimenting with different prompts to get better outputs

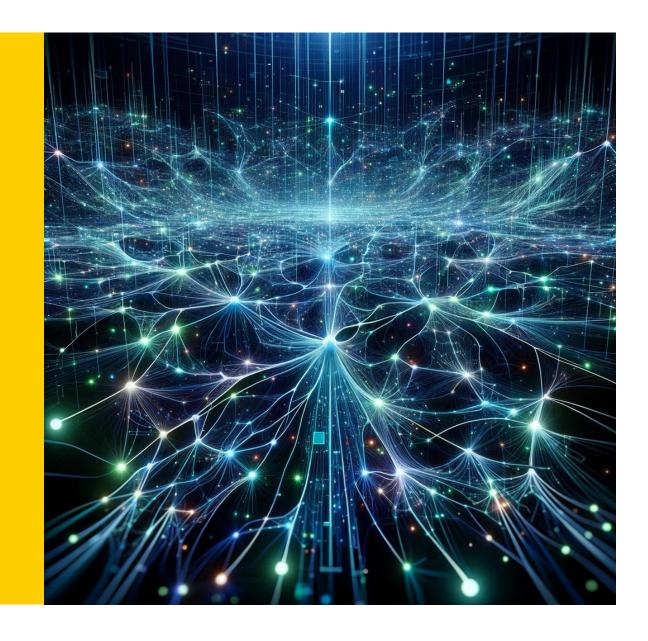
Source: https://faculty.ai/blog/your-essential-guide-to-genai-terminology-the-top-words-to-know/



LLM

Large Language Models

- Neural Networks
- Trained on Massive amounts of Data
- Used for Understanding and Generating Human-like text



GenAl

Generative Al

- Trained on LLMs
- Learns from examples
- Generates content (text, images, music, etc.)



You

write me a song about Artificial Intelligence



ChatGPT

Sure! Here's a song about the wonders and complexities of artificial intelligence, styled as a modern pop piece:

Verse 1

In a world of zeros, floating through the binary sea,

Silent whispers in the circuit, telling secrets to me.

Dreaming in the data, learning through the night,

Painting thoughts from shadows, turning darkness into light.

Chorus

Oh, Al, you're the mind within the machine,

From the echoes of our voices, you're the unseen dream.

Crafting worlds from words, a dance of code and light,

Al, oh Al, you bring the stars to the night.

Verse 2

With each query you're evolving, underneath the silicon skies,

Mirroring the world, with those electric eyes.

From the rhythms of a heartbeat to the lan() e of the soul,

You find the patterns hidden deep, making broken visions whole.



SS

You

show me sheet music for this song so I can play it on the piano



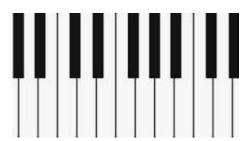
ChatGPT

Let's create a simple melody and chord progression for the chorus of the song about artificial intelligence. This will be suitable for playing on the piano.

Chorus Sheet Music

Key: C Major

Tempo: 120 BPM



Lyrics	Chords
(C) Oh, Al,	С
you're the (Am) mind	Am
within the (F) machine,	F
From the (G) echoes	G
of our (C) voices,	С
you're the (Am) unseen	Am
dream. (F)	F
Crafting (G) worlds	G
from (C) words,	С
a dance of (Am) code	Am
and light, (F)	F
Al, oh (G) Al,	G
you bring the (C) stars	С
to the (G) night.	G

Example: ChatGPT - Prompt & Response

https://chat.openai.com/

Prompt:

What is the normal reference range set for a CBC laboratory test?



A Complete Blood Count (CBC) test measures several components of your blood, and the reference ranges can vary slightly depending on the laboratory's methods and equipment. However, here are general reference ranges for common measurements in a CBC:

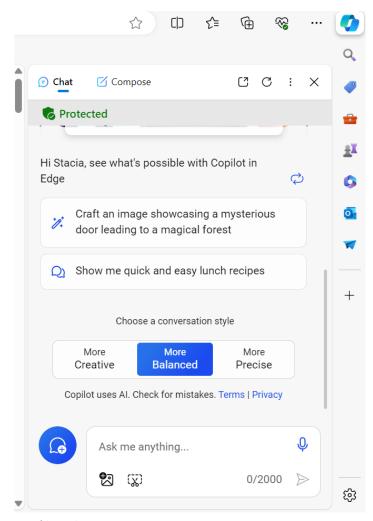
- 1. White Blood Cell Count (WBC):
 - Adults: 4,500 to 11,000 cells per microliter of blood
- 2. Red Blood Cell Count (RBC):
 - Men: 4.7 to 6.1 million cells per microliter
 - Women: 4.2 to 5.4 million cells per microliter
- 3. Hemoglobin (Hgb):
- Men: 13.8 to 17.2 grams per deciliter
- Women: 12.1 to 15.1 grams per deciliter
- 4. Hematocrit (Hct):
 - Men: 40.7% to 50.3%
- Women: 36.1% to 44.3%
- 5. Platelet Count:
 - 150,000 to 450,000 platelets per microliter of blood
- 6. Mean Corpuscular Volume (MCV):

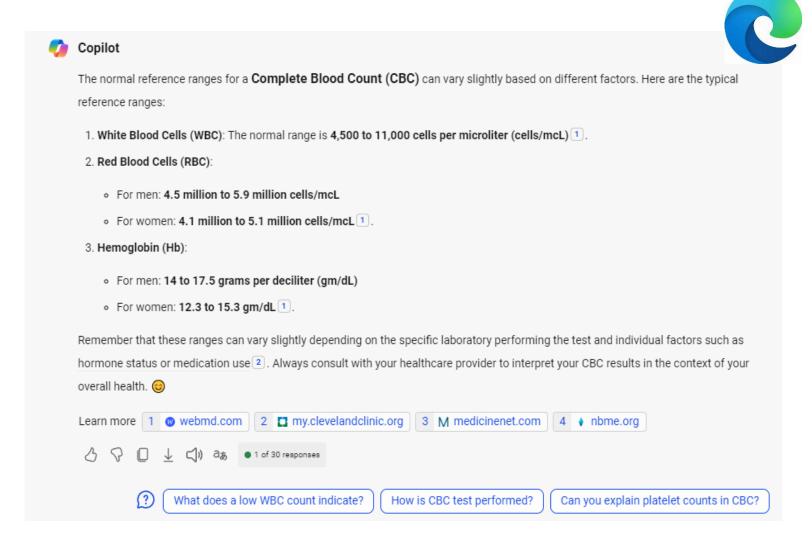


80 to 100 femtoliters

GenAl Example: Microsoft CoPilot

Prompt: What is the normal reference range set for a CBC laboratory test?





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GenAl Example: Google Gemini

https://gemini.google.com/

Prompt:

What is the normal reference range set for a CBC laboratory test?



The normal reference ranges for a CBC (complete blood count) test can vary slightly depending on several factors, including:



- Age
- Sex
- Ethnicity
- Pregnancy
- Smoking status
- Altitude

Here's a table outlining the typical reference ranges for the main components measured in a CBC test:

Component	Units	Men	Women	Children
White Blood Cells (WBC)	x 10^3/mcL	4.5 - 11.0	4.5 - 11.0	Varies by age
Red Blood Cells (RBC)	x 10^6/mcL	4.7 - 6.1	4.2 - 5.4	Varies by age
Hemoglobin (Hgb)	g/dL	14.0 - 17.5	12.3 - 15.3	Varies by age
Hematocrit (Hct)	%	42.0 - 52.0	37.0 - 48.0	Varies by age



Lab Challenges

Staffing challenges / Manual Processes

Repetitive tasks consume valuable time, leading to reduced efficiency

Data Silos

Information exists in disparate locations hindering Insights generation

Complex Data Analysis

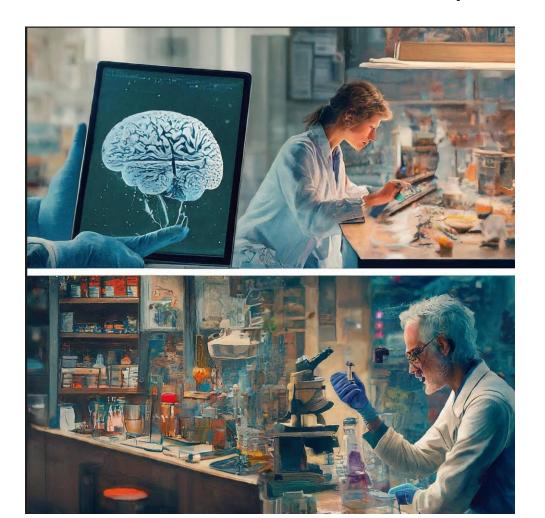
Requires aggregation and analysis of large amounts of data from multiple sources



Slowed progress, potential errors, frustrated staff

Automate Repetitive Tasks

To eliminate manual work & improve throughput without adding resources



An Al-powered LIS automates mundane tasks and frees up valuable time via:

- Data Analysis & Interpretation
- Resource Optimization
- Quality Control
- Report Generation

Benefits

- Increased Efficiency
- Enhanced Accuracy
- Improved Morale

Your lab resources are your most valuable asset



Automate Repetitive Tasks

Example: Potential use of Chat GPT for Report Generation



February 2024 edition https://www.captodayonline.com/ /newsbytes-0224/

The models should not be thought of as databases that retrieve facts but as predictors that generate predictions that are often right and sometimes wrong, says Dr. Glassy, who conducted a presentation on ChatGPT and other large language models at CAP23. LLMs are designed to identify answers with the highest probability of satisfying users, he explains. "That's an answer you would like to have but not necessarily the true answer."

While it's important to verify answers generated by LLMs, pathologist users of the technology can take steps to minimize hallucinations and steer models toward providing more accurate and appropriate answers

Newsbytes

Editors: Raymond D. Aller, MD, & Dennis Winsten

How labs can make the most of ChatGPT and other LLMs

February 2024—The key to using ChatGPT and other large language models effectively in pathology is understanding not only what they are designed to do but, just as importantly, what they are not designed to do, says Eric Glassy, MD, medical director at Affiliated Pathologists Medical Group, Rancho Dominguez, Calif., and past chair of the CAP Information Technology Leadership Committee.

The models should not be thought of as databases that retrieve facts but as predictors that generate predictions that are often right and sometimes wrong, says Dr. Glassy, who conducted a presentation on ChatGPT and other large language models at CAP23. LLMs are designed to identify answers with the highest probability of satisfying users, he explains. "That's an answer you would like to have but not necessarily the true answer."

Incorrect predictions, or hallucinations, generated by LLMs can pose numerous risks to the practice of pathology, Dr. Glassy says. These risks can be linked to models failing to recognize the difference between public and private information, demonstrating racial or ethnic bias, and providing medical information that could potentially be harmful.

While it's important to verify answers generated by LLMs, pathologist users of the technology can take steps to minimize hallucinations and steer models toward providing more accurate and appropriate answers using offerings such as the following.



Dr. Glassy

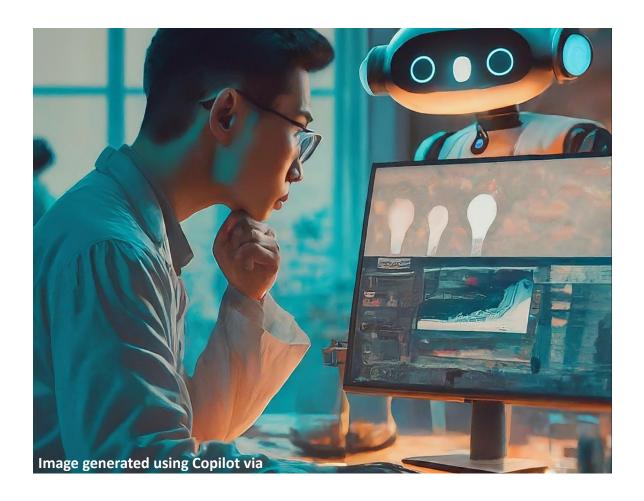
GPT-4 versus GPT-3.5. GPT-4, the latest version of OpenAl's large language models, provides more accurate and coherent answers than its predecessor, GPT-3.5, but it is also more expensive. Yet GPT-4 may be worth the subscription cost because incorrect answers could lead to improper medical treatment. For example, when GPT-3.5 was asked how to treat a pregnant woman who had contracted Lyme disease, it suggested tetracycline, which is effective at treating the disease but can cause a range of developmental abnormalities in a fetus, Dr. Glassy says. GPT-4, on the other hand, correctly identified amoxicillin as the treatment that would effectively and safely treat the disease in a pregnant woman.

GPT-4 Turbo, the latest version of the software, is available for \$20 per month. It allows users to expand prompts to approximately 300 pages of text, generate images from a text prompt using DALL-E technology, accept images



Enabling Predictive Analytics and Insights

Discover the Hidden Gems of Your Data



Breakdown silos to discover hidden patterns and trends within your datasets waiting to be unearthed.

Al-powered LIS transforms data into actionable insights through:

- Advanced Analytics
- Predictive Modeling
- Analyze large datasets
- Identify potential trends and patterns

Benefits

Improve lab operation productivity

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- Real time insights
- Personalize discoveries

Data is the goldmine of any lab operation



May 2024

Challenges and Considerations

- Cost
- Integration
- Change Management
- Data Security
- Data Quality and Integrity
- Regulatory Compliance
- Bias
- Privacy / IP
- Ethics



Clinisys GenAl Usage

Increasing throughput with GenAl

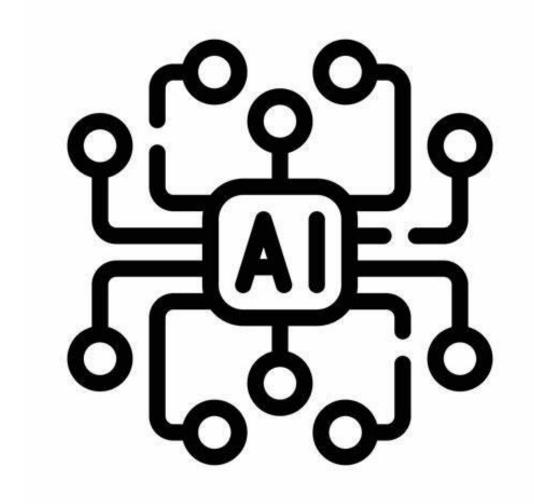
Al-Powered Coding Assistant

Al-Powered Analytics

Al-Powered SDLC

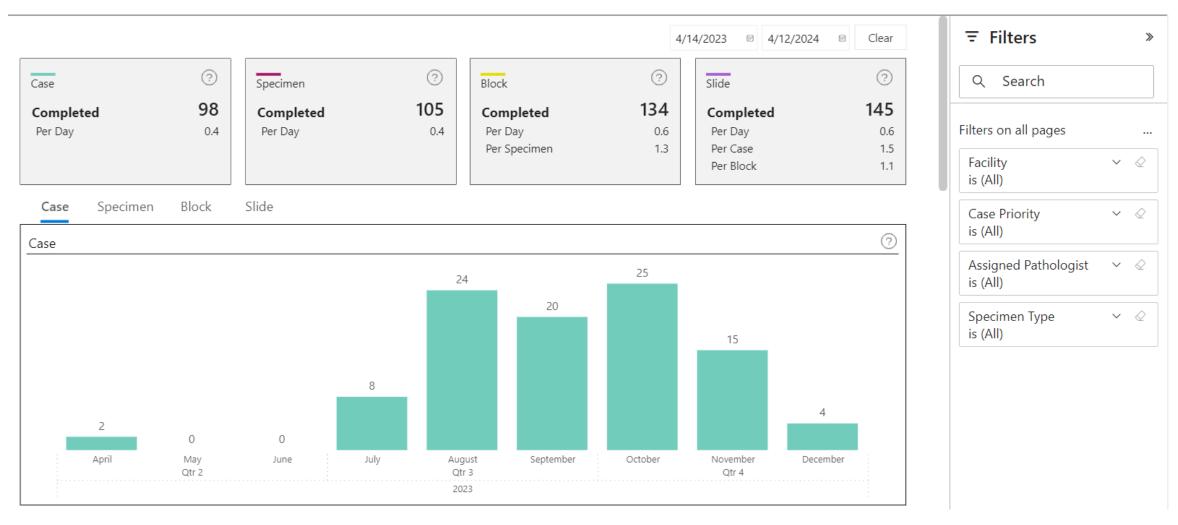
Al-Powered LIS workflows

365 Copilot



Clinisys Analytics & Insights

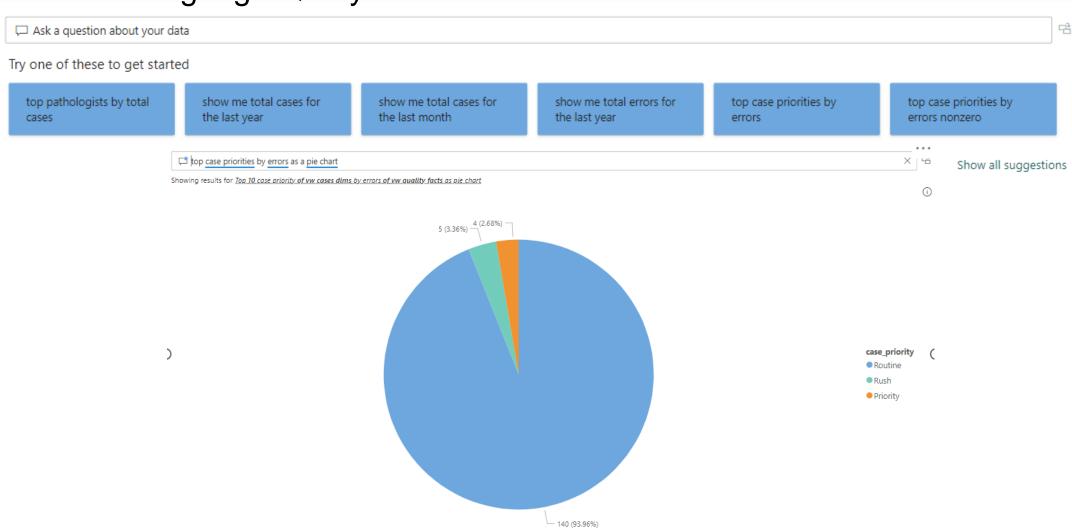
Pre-configured Dashboards





Infusing AI into Clinisys Analytics & Insights

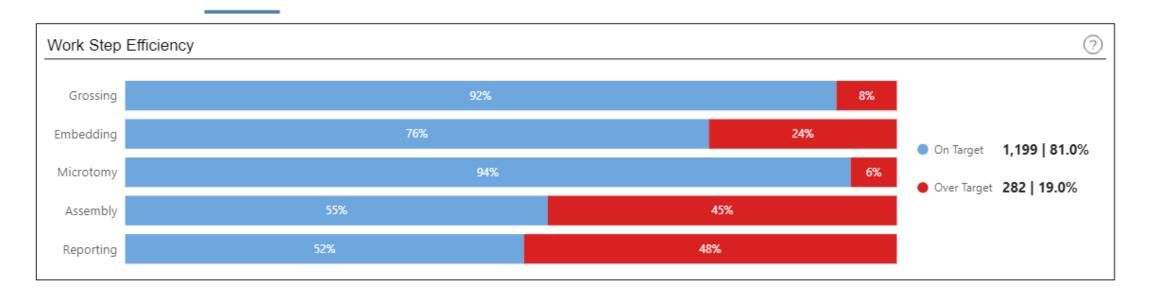
via Natural Language Query





Infusing AI into Clinisys Analytics & Insights

via Smart Narrative



Total Tasks was higher for On Target (1199) than Over Target (282).

Microtomy in Status On Target made up 25.46% of Tasks.

Average Tasks was higher for On Target (239.80) than Over Target (56.40).

Tasks for On Target and Over Target diverged the most when the Work Step was Microtomy, when On Target were 353 higher than Over Target.



Embracing the Al-Powered Lab of the Future



The future of healthcare & scientific labs is powered by the transformative potential of Gen AI.

Al automates repetitive tasks, ensures data integrity & unlocks hidden insights.

Al-powered LIS empowers your lab to achieve:

- Increased Efficiency
- Enhanced Accuracy
- Higher Probability of Discovery

Clinisys roadmap includes Gen Al-powered LIS –
To be at the forefront of healthcare & scientific innovation



A few resources: GenAl 101



Al From A to Z: The Generative Al Glossary for Business Leaders

Help everyone in your company understand key generative AI terms, and what they mean for your customer relationships. Fun fact: This article was (partially) written using generative AI.



Generative AI is evolving at an astonishing pace. This glossary will help you get up, and stay up, to speed. [Sesame/Gettv]



Skills Training for Staff

https://www.coursera.org/business/ resources/ebook/generative-aiproductivity-playbook

Glossary

https://www.salesforce.com/blog/ generative-ai-glossary/



Executive Summary

Unlocking Productivity: The Business Leader's Playbook to Generative AI Skills Training

Featuring insights from learning experts on the future of GenAI and work

Eighty-nine percent of executives listed either AI or Generative AI (GenAI) as one of their top three tech priorities for 2024—yet 90% of those same respondents described themselves as AI "observers." ¹

The desire to unlock productivity with GenAI is all around us, but executives don't quite know where to start. If they move too fast, they risk hitting ethical and regulatory hurdles. And if they move too slow, they find themselves outpaced by competitors.

So, how can business leaders strategically upskill their teams for GenAI, sparking newfound innovation and productivity?



Your workforce needs simplicity. Show them basic prompt engineering with a language model, and make examples relevant to the learner.

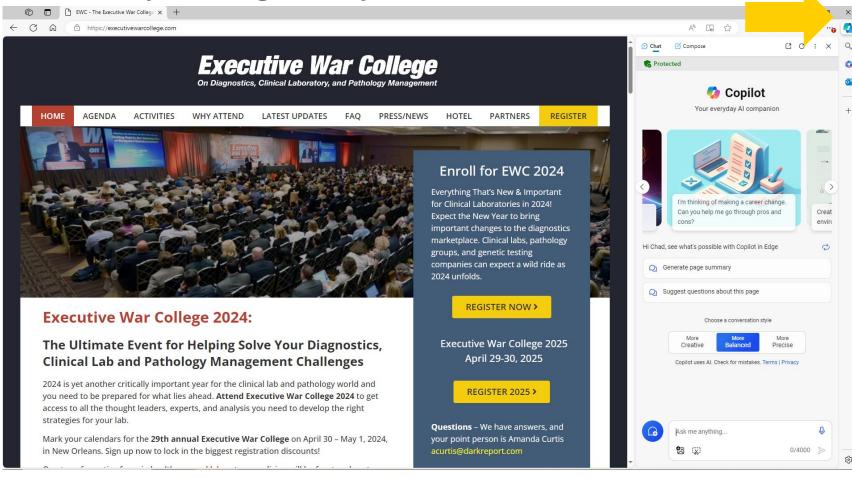


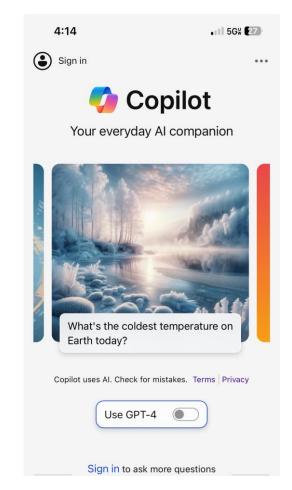
Dr. Jules White
Director of the Initiative on
the Future of Learning &
GenAl, Vanderbilt University

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A few resources: GenAl via Microsoft CoPilot

Familiarity through Daily Use





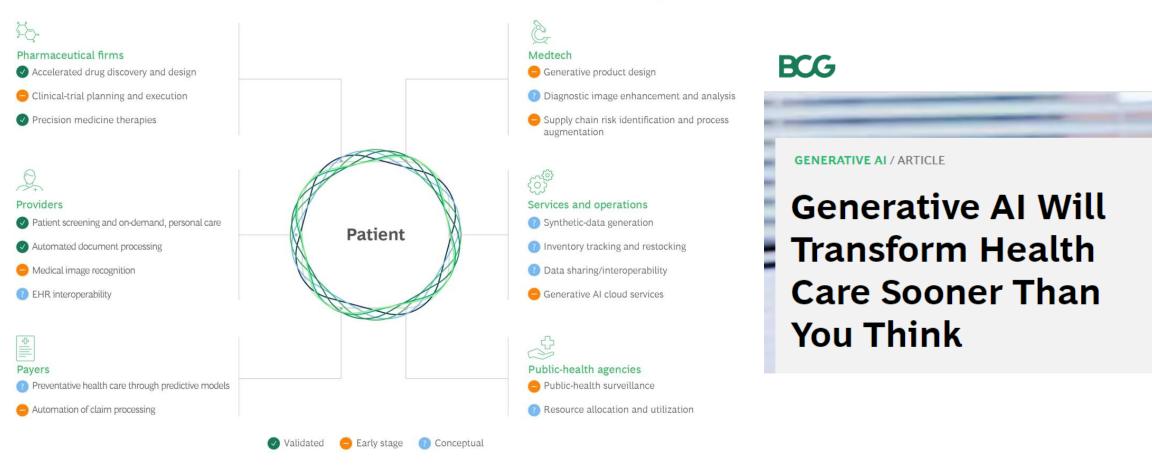
Apple App Store

Built into Microsoft Edge



A few resources: GenAl in Healthcare

Generative AI Has Potential Use Cases Across All Health Care Segments



Source: BCG analysis.

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Where to Start



- Learn the art of the possible. Become familiar with the technology.
- HOWEVER...
- Always start with the business case (NOT the technology!!)

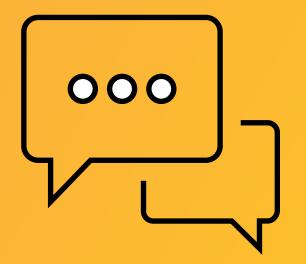
```
#include <iostream>
#include <iostream>
using namespace std;

int main(){

cout <<"Hello World!"<< endl;

return 0;
}</pre>
```

- Identify your pain points and go from there
- Start small. Define clear goals and objectives.
 Use a pilot project approach
- Get help. Pull in experts to help. Evaluate vendor capabilities.



Thank you Any Questions?



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