

CM05: Unlocking the potential of Generative AI for the pharmaceutical industry – SAS AI-enabled assistant

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Agenda

- What is Generative Artificial Intelligence?
 - Different flavors of Al
 - How does it work?
 - Gen(eral) Al Versus Gen(erative) Al
- Healthcare & Life Sciences possible use cases
 - Clinical Ambiant intelligence with Nuance
 - Drug Discovery and Generative Al
 - Copilots
- Introducing new SAS development of a AI-enabled assistant
- Food for thoughts & Questions





Introduction to Generative Al



The world is changing

Industries are transforming rapidly

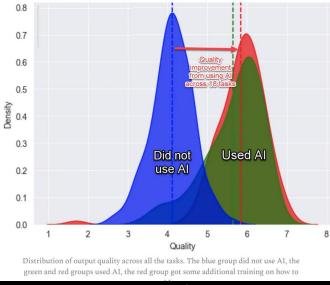
The drivers for economic growth are evolving

Technology is at an inflection point

The AI technology is here



impact is real



Source: Centaurs & Cyborgs

Generative Artificial Intelligence

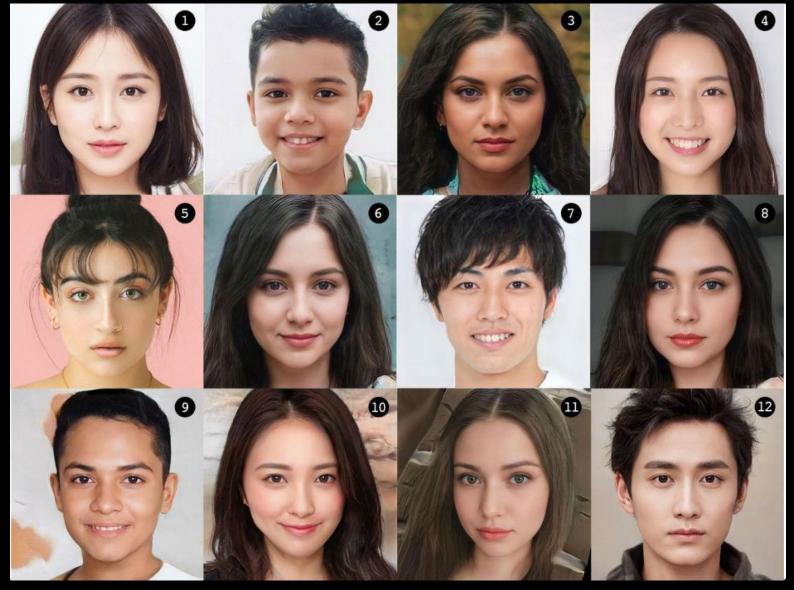


Generative AI algorithms learn the underlying patterns in a very large training dataset and can generate new instances that are statistically similar to the original data.

The most powerful generative AI algorithms are built on top of *foundation models* that are trained on a vast quantity of unlabeled data in a self-supervised way to identify underlying patterns for a wide range of tasks (*text, code, image, audio, video...*)



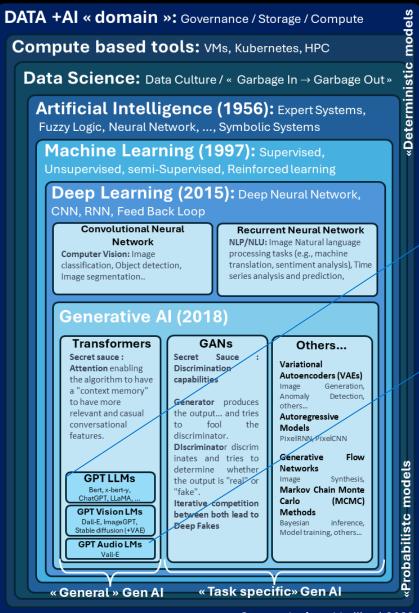
Can you beat the IA?
Identify the only one real face



https://thispersondoesnotexist.com



Data + AI: Historical review of the landscape



Within Azure, OpenAl services are "just" a powerful family of Al-services offered





Transformers & GANs: two pillars of Generative Al

TRANSFORMERS



GENERATIVE ADVERSARIAL NETWORKS (GANs)



Natural Language Understanding Natural Language Processing

Media Generation (Image/Video/Audio)

Discrimination



Generative AI: Potential Legal risks / Trustworthy AI

Ssas

Generative AI (GAI) has taken the world stage, and it is here to stay. Advances in GAI are accelerating faster than either governing bodies or society have had time to reasonably address. In some categories of GAI, business value, risk, and ethics have yet to be reconciled. Resolving these three factors is critical.

SAS has long been a pioneer in, and a champion of, Al innovation. At SAS, we know that advances in technology don't exist in a vacuum and can have far-reaching effects and implications.

SAS considers trustworthy and ethical AI to set the strategy and guardrails for all AI, and generative AI, innovation. Core to SAS DNA is to place human-centricity, customers, and "doing the right thing" first.

SAS is committed to innovating responsibly while inspiring and empowering our customers, partners, and communities to do the same.



Microsoft Responsible AI Principles:

1. Fairness

All Al systems should treat all people fairly.

2. Reliability & Safety

Al systems should perform reliably and safely.

3. Privacy & Security

Al systems should be secure and respect privacy.

4. Inclusiveness

Al systems should empower everyone and engage people.

5. Transparency

Al systems should be understandable.

6. Accountability

People should be accountable for AI systems.

Microsoft Pledges Legal Protection for Customers Against AI-Generated Copyright Claims

Microsoft will provide legal protection to its commercial customers against third-party copyright infringement lawsuits, provided they adhere to Microsoft's established guardrails and content filters.



Don't mistake Gen(eral) Al & Gen(erative) Al



Just Another Rather Very Intelligent System (JARVIS)

JARVIS is a General AI, multimodal, multipotent, augmented with robotic peripherals... It can handle a complex conversation and derive action plans for execution (it can calculate and build IronMan armors... and many other things).

JARVIS is an ecosystem of specialized systems with powerful orchestration

Speech to text + LLM+ Text to Speech would only be the NLP/NLU layer of JARVIS, the polite voice able to interact with Tony Stark (Conversational Language Understanding – CLU-)

BUT.... IT IS POSSIBLE TO FUNCTIONALY ENRICH LLM, STEPS TO MULTIMODAL AI

EXTEND LLM CAPABILITIES VIA PLUGINS

It is possible to extend the NLU/NLP core capabilities of LLM with « plugins ». Plugins provide new capabilities to LLM, leveraging their NLU/NLP capabilities to feed other specialized API (web search, PDF reader, computer vision...) which provide API-processed data back to the LLM.

PREPOPULATE THE PROMPT TO HAVE FOCUSED EMBEDDINGS

You can « ground » the LLM with a « preprompt » telling the LLM how it should behave (role, context, expectations, tone guidance...).

Providing pairs of Question/Answer is also another way to « ground » the LLM and limit hallucinations.

RAG & FINE TUNING

Retrieval Augmented Generation combines a retriever to the LLM. It will search documents, internet to enrich and generate more contextual embeddings along with the prompt itself. Fine-Tuning is a re-training of the LLM with a more specialized dataset, but also modifying other parameters of the training. Fine-Training can be used to augment the context memory (the attention) of the LLM.



SAS and Large Language SAS Viya to Models

Three distinct approaches

help build GAI apps.

SAS Viya provides capabilities that you can use in your LLM app stack and toolchain.

Al Assistants to help you use SAS tools.

SAS will incorporate GAL to improve productivity with SAS Viya, further democratize analytics, and reduce cost of ownership.

Al Assistants to help with industryfocused tasks.

SAS will deliver industryand domain-specific solutions that help organizations increase process efficiencies, optimize yield, or mitigate risks.



Use Cases in Life Sciences



Generative AI: most prominent HLS scenarios we are observing across the value chain

Healthcare Automation: Companies like **Anthropic**, *DataGen*, *data.world* are automating clinical data extraction, integration and analytics using generative models to standardize records and power evidence-based decisions. *Microsoft Nuance* <u>ambient clinical intelligence</u> is revolutionizing the way HCP and patient interpersonal relationship to produce gold clinical data.

Healthcare Chatbots: All assistants use generative capabilities for question-answering, decision support and virtual consultation in areas like imaging, genomics, clinical trials. During the COVID crisis, health organizations have created 1,230 COVID-19 self-assessment bots based on the Microsoft Healthcare bot service, reaching 18 million individuals and serving more than 160 million messages. Now these bots are powered by Generative AI (Azure Healthbot).

Drug Discovery: Al startups like *Exscientia* and *DeepMind* are employing GANs and VAEs to generate novel molecular structures with desired properties, speeding up drug R&D. DeepMind's AlphaFold has also transformed protein structure prediction. <u>Azure Quantum Elements</u> speeds up certain chemistry simulations by 500,000 times, which is like compressing a year into one minute.

Genomic Analysis: Companies like Anthropic, Insitro and TwoXAR are leveraging generative language models to surface novel gene relationships and functional impacts from integrated omics datasets, advancing precision medicine. Microsoft Genomics can support your most demanding sequencing needs in a secured and scalable environment.

Clinical Trial Optimization: Al startups are developing techniques to synthesize prototype patient data (data augmentation & digital twins) and dynamically adapt trials based on outcomes, making them more efficient and personalized. SAS Viya platform concentrates all latest digital innovations needed to speed-up your clinical analytics.

Generative AI: ambient intelligence to optimize the HCP/patient interaction (NUANCE)



- Ambient intelligence provided by NUANCE DAX capture with its NLP/NLU capabilities all the meaningful information during the anamnesis.
- It can complete the patient history in the EMR
- Empowering the HCP to be really focused on her/his interpersonal relationship with the patient.



Captures a multi-party conversation ambiently

Clinicians engage in natural conversation with patients and other family members while the mobile app securely captures the conversation at the point of care, allowing clinicians to connect with patients rather than screens.



Creates clinical documentation automatically

Converts encounter conversations into comprehensive specialty-specific clinical documentation summaries, and delivers the summary into the application and the Dragon Medical One desktop in seconds for clinician review and editing.



Produces high-quality documentation

Uses the latest AI technologies and is trained on over 10M encounters to produce accurate documentation efficiently and consistently.



Works seamlessly with electronic health records

Used in conjunction with Dragon Medical One which integrates with over 200 EHRs, streamlining implementation.



Generative AI: Generation is not the big issue when dealing with Drug Discovery

High Throughput Clinical Trial Clinical Trial Clinical Trial Combinatory Hit to lead PK/PD Formulation Pre clinical Phase Hit-to-lead **Combinatory Chemistry High-Throughput Screening** Combinatory chemistry to Hit-to-lead optimization involves **Process** High-Throughput Screening (HTS) generate large libraries of to rapidly test a large number of modifying the initial hit molecule description to improve its potency, selectivity, molecules (drug candidates) compounds against specific and pharmacokinetic properties. biological targets **Currently... Guided chemical synthesis Automated (robotics) micro-well** Computerized assays, In-vivo, Inpatterns (Lengthy & complex) assays against compounds Vitro assays, PK/ADME...(either in solid or liquid phase librairies (Very complex, lengthy, (Very complex, very lengthy, very expensive) expensive)

High Throughput Parallel Synthesis (Robotic Diversity Oriented synthesis)

De Novo Molecule Design:

networks (GANs) to generate

or generative adversarial

structurally diverse and

synthetically feasible

molecules.

variational autoencoders (VAEs),

The GAN discriminator's ability to evaluate the quality of generated compounds can aid in selecting and prioritizing compounds for experimental validation. Discrimination, not generation is

Assays against compound-librairies

in microwells with robotics and

computerized analytics.

the key point

Structure-Activity Relationship (SAR) Analysis, ADME (absorption, distribution, metabolism, and excretion), ...

Generative AI models can propose new scaffolds or analogs based on the initial hit compound(s).

Generative AI models could learn from large datasets of compounds with known absorption, distribution, metabolism, excretion (ADME), and toxicity properties



With Gen Al...



Introducing a IUSE SAS Copilot-like functionality



Generative AI: SAS and Microsoft partner to add top notched Gen AI to the SAS Platform





■ Could Github Copilot help us write SAS code?

iii Posted 07-31-2023 03:38 AM (551 views)

Has anyone done some research on github copilot? This Al plug-in (in VS Code) can support many mainstream programming languages, so can it also help us write SAS code?

There is a demand for this kind of features !!!

But poor outcome « out of the box »!

```
FreelanceReinh JADE | LEVEL 19

Re: Extracting year and month from date9. date

| Posted 01-31-2023 09:04 AM (284 views) | In reply to PaigeMiller

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| Posted 01-31-2023 09:04 AM (284 views) | In reply to PaigeMiller

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Hopefully, SAS & Microsoft partner together to bring SAS "Copilot" to you !!!

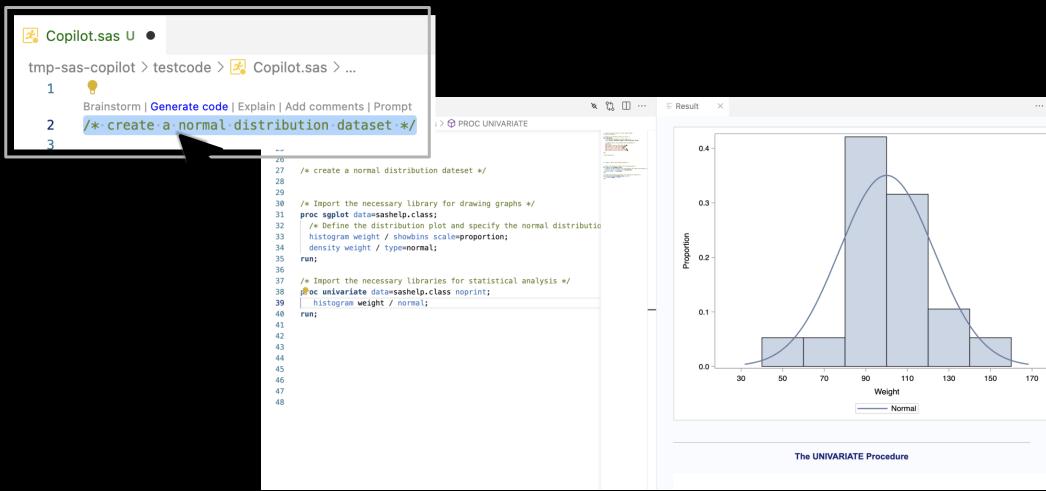


HOME ARTICLES ▼ INSIGHTS ▼ INFORMATION

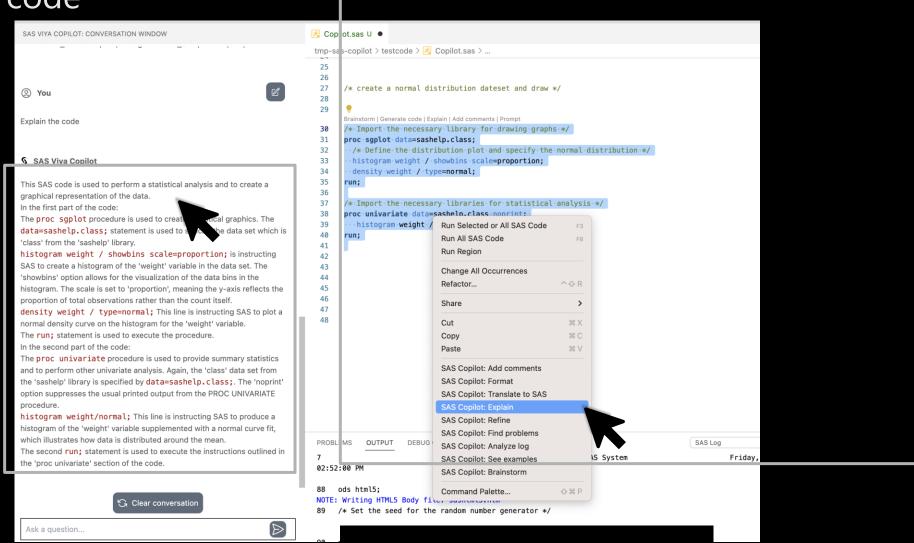
SAS partners with Microsoft to bring generative AI to the enterprise



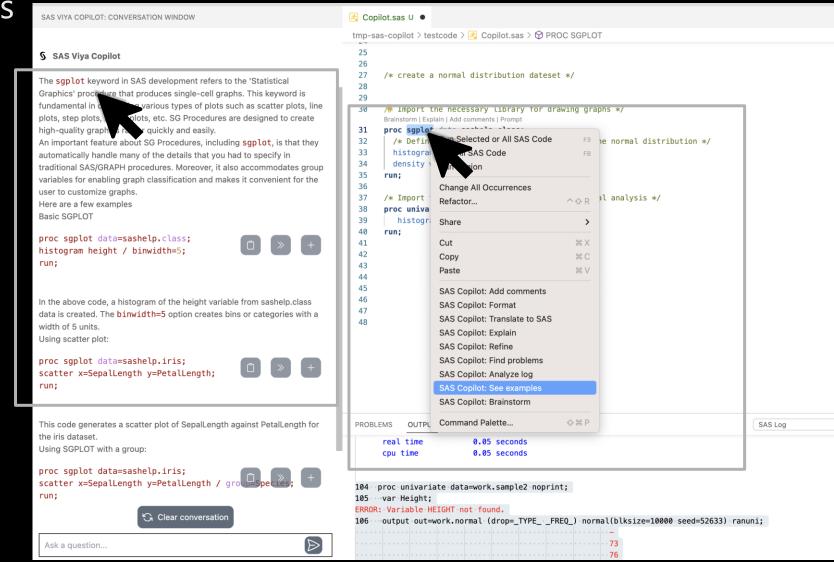
Type a comment to generate code



Explain the code



See Proc examples



SAS VIYA COPILOT: CONVERSATION WINDOW

Refine the code

SAS Viya Copilot

```
Here is an optimized version of your SAS code:

/* Import the library only once */
libname mylib 'C:\sasdemo';

/* Perform statistical analysis before plotting */
proc univariate data=mylib.class noprint;
    var weight;
    histogram / normal;
run;

/* Then plot the graph with necessary labels and titles for proc sgplot data=mylib.class;
    title "Histogram with Normal Curve";
    histogram weight / showbins scale=proportion fill;
    density weight / type=normal lineattrs=(color=red);
    xaxis label = 'Weight';
    yaxis label = 'Frequency';
run;
```

Here are the reasons why I made these changes:

Ask a question..

Importing Libraries: Instead of importing libraries multiple times, it's a good practice to do it once at the beginning for efficiency.

Order of Operations: Perform the statistical analysis first before plotting for better workflow order.

Variables: In the **proc univariate**, I specified the **var** statement to clearly indicate which variable is being analyzed.

Labels and Titles: I added titles and labels on x-axis and y-axis to make the graph more understandable.

Plotting: I added fill option to the histogram for better visual aspect and change the line color for the density plot to red for contrast.

Note: Please replace 'C:\sasdemo' with your library current location.

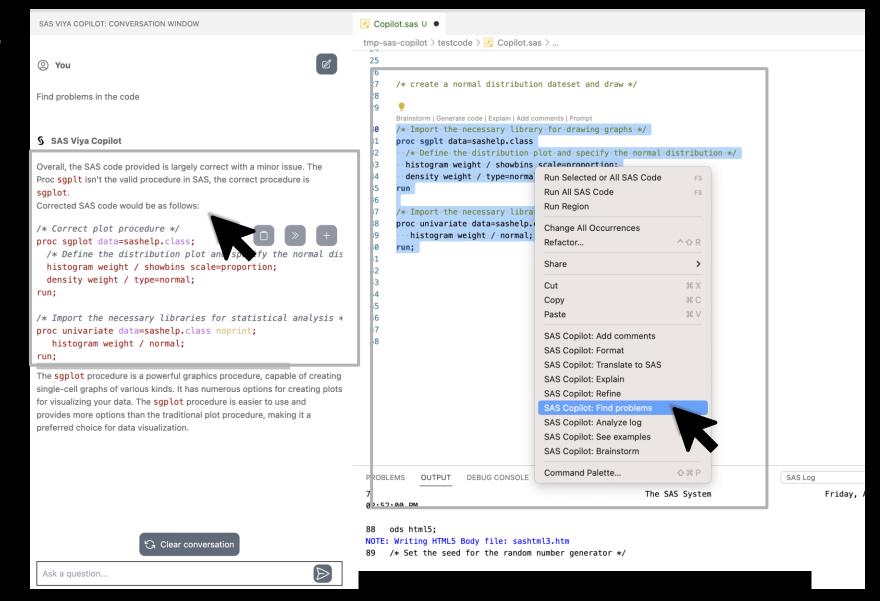


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Copilot.sas U 

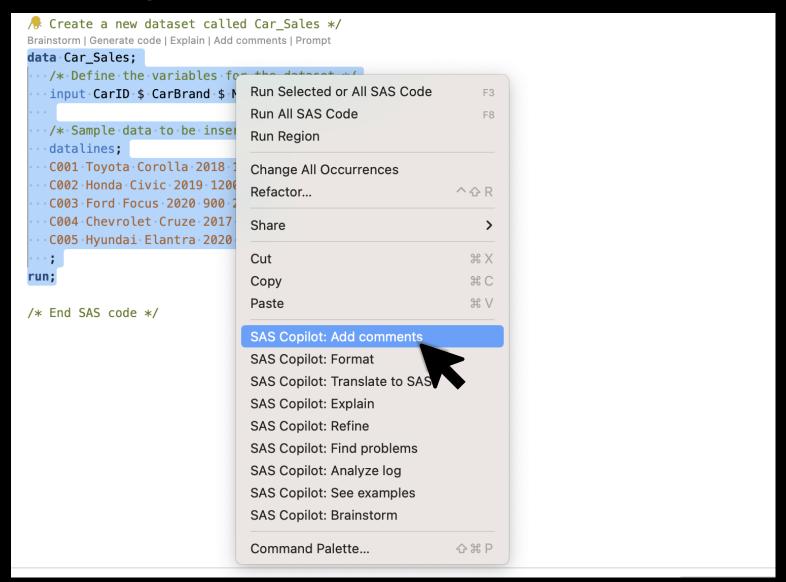
tmp-sas-copilot > testcode > 🛂 Copilot.sas > ...
25
26
      /* create a normal distribution dateset and draw */
28
29
       Brainstorm | Generate code | Explain | Add comments | Prompt
       /* Import the necessary library for drawing graphs */
       proc sqplot data=sashelp.class:
        /* Define the distribution plot and specify the normal distribution */
        histogram weight / showbins scale=proportion;
34
        density weight / type=normal;
35
       run;
36
37
       /* Import the necessary libraries for statistical analysis */
       proc univariate data=sashe
                                   Run Selected or All SAS Code
         ·histogram·weight·/·norm
                                   Run All SAS Code
       run;
                                    Run Region
41
42
                                    Change All Occurrences
43
                                    Refactor...
                                                                  ^ む R
45
                                    Share
47
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                                    SAS Copilot: Add comments
                                    SAS Copilot: Format
                                    SAS Copilot: Translate to SAS
                                    SAS Copilot: Explain
                                    SAS Copilot: Refine
                                    SAS Copilot: Find problems
                                    SAS Copilot: Analyze log
                                                                                             SAS Log
PROBLEMS
           OUTPUT
                      DEBUG CONSC
                                    SAS Copilot: See examples
                                                                                                      Friday,
                                    SAS Copilot: Brainstorm
02:52:00 PM
                                    Command Palette...
                                                                  企業₽
    ods html5:
NOTE: Writing HTML5 Body file: sashtml3.htm
89 /* Set the seed for the random number generator */
```

Find problems



SAS Copilot-like functionality for Code Gen

Add comments to code





User Interface

- NOW via Visual Source Code Extension
- NEXT Any code editor integrated throughout Viya



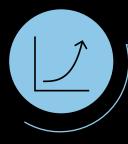
Users

• Novice → Advanced SAS Programmers



Use

- NOW Generate SAS code for a variety of data and analytical procedures
- **NEXT** Enable 9 → Viya migration
- **NEXT** Customize to industries



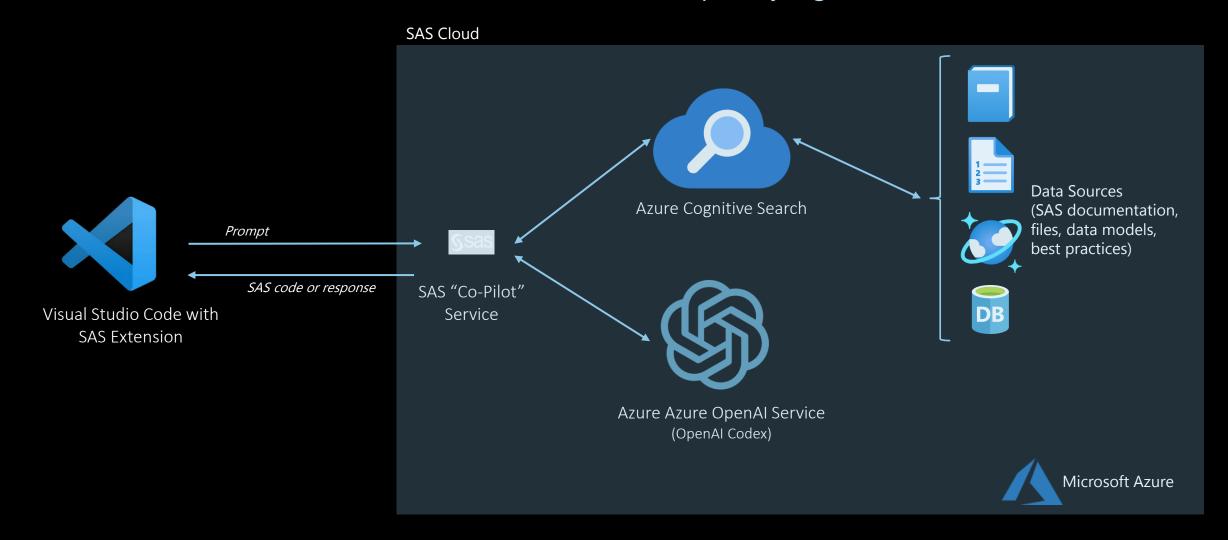
Value

- Increased productivity and speed to market
- Decreased developer time spent on reviewing SAS documentation & syntax
- Increased discoverability of SAS capabilities
- Available via subscription service

SAS Al-enabled assistant leveraging Azure & RAG

Available via connection to SAS Cloud on Azure.

Benefits from the most secured Cloud environment and data privacy highest standards

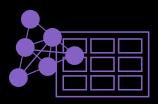




SAS and Generative Al

SAS includes these three AI/ML categories as part of its approach to generative AI:

Synthetic Data Generation



Digital Twin



Large Language Models



SAS and Generative Al

SAS includes these three AI/ML categories as part of its approach to generative AI:

In Clinical Trials...

Synthetic Data Generation



Useful when not enough data, such as with rare diseases

Digital Twin



For virtual patients to simulate "what if" scenarios

Large Language Models



Automate manual tasks, like building an FDA-regulated eCRF from a protocol



Combining SAS and Microsoft's Azure OpenAl Service

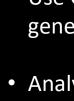
Strategic: How do we transform clinical trial analysis and submission?

AI-Enhanced SAS Life Sciences Analytics Framework

SAS strives to combine:

- OpenAl's GPT capabilities to process natural language queries and generate responses
- SAS's expertise in statistical computing environments for clinical research
- Dynamic embeddings built by SAS to tailor the results for clinical studies
- All on top of the enterprise-grade characteristics of Azure.

What are possible outcomes?



- Use GPT-3+ to turn a protocol into an autogenerated electronic case report form
- Analyze vast amounts of patient data to identify high-potential candidates during cohort and patient selection
- Review enrollment simulations for patient recruitment
- Use NLP to extract patient-reported outcomes.

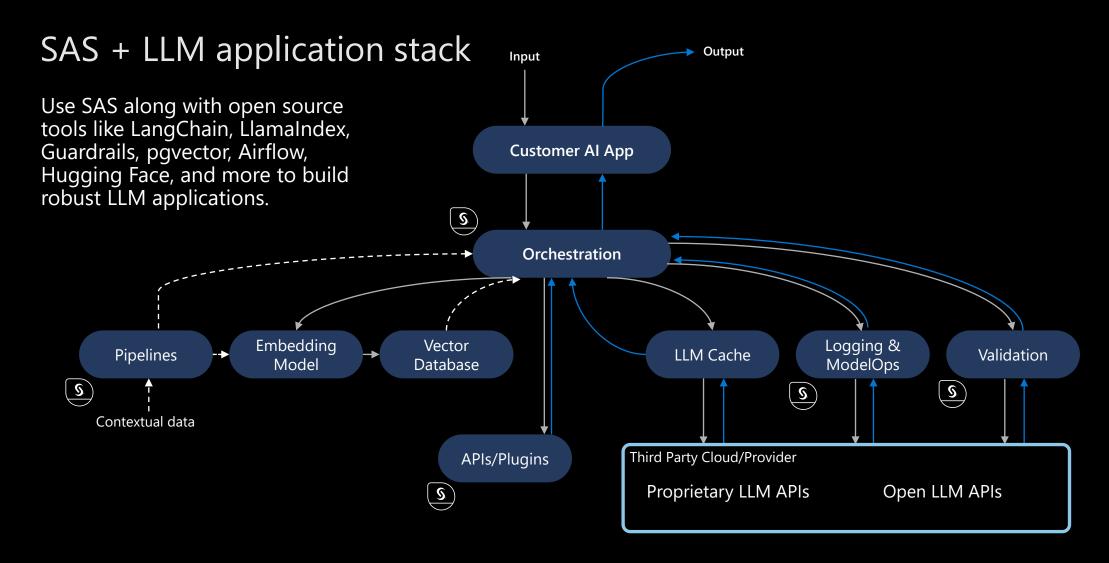




Al Assistants to help with industry-focused tasks

SAS will deliver industry- and domain-specific solutions that help organizations increase process efficiencies, optimize yield, or mitigate risks.

SAS Viya to help build gen Al apps





Food for thought...



Explainability of AI and LLMs: the lost fight?

Principle #3: Transparency
How Al works must be easily
explained by those who create it and
by those who use it.

Problem: To promote unbiased and inclusive AI, we must therefore be able to give the means to anyone (without being an expert) to be able to easily check its functioning and detect possible bias or manipulation.

How can you trace the dynamic behavior of 175+ billions of parameters when it comes to decision making explainability?

The chaos theory and emergence laws make the emergence of "eloquence" possible...

...but they also make explainability obsolete

(41) Artificial intelligence: Is it as "stupid" as a Galton board? | LinkedIn



Responsible AI and Sustainability?

On the evening of Wednesday, December 2, Timnit Gebru, the co-lead of Google's ethical AI team, announced <u>via Twitter</u> that the company had forced her out.

"On the Dangers of Stochastic Parrots: Can Language Models Be Too Big?" lays out the risks of large language models—AIs trained on staggering amounts of text data. These have grown increasingly popular—and increasingly large—in the last three years. They are now extraordinarily good, under the right conditions, at producing what looks like convincing, meaningful new text—and sometimes at estimating meaning from language. But, says the introduction to the paper, "we ask whether enough thought has been put into the potential risks associated with developing them and strategies to mitigate these risks."

ARTIFICIAL INTELLIGENCE

Training a single Al model can emit as much carbon as five cars in their lifetimes

Deep learning has a terrible carbon footprint.



Common carbon footprint benchmarks

in lbs of CO2 equivalent

Roundtrip flight b/w NY and SF (1 passenger)

Human life (avg. 1 year)

American life (avg. 1 year

US car including fuel (avg. 1 lifetime)

Transformer (213M parameters) w/ neural architecture search

1,984

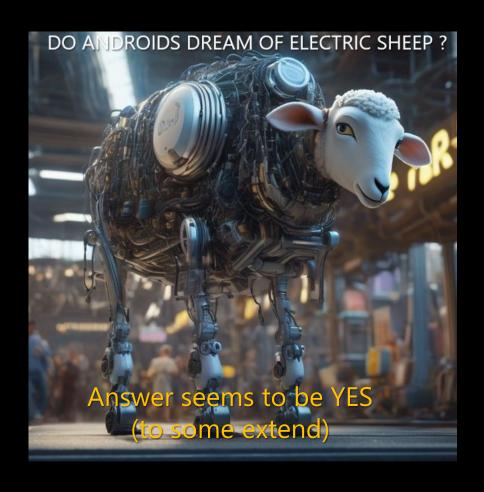
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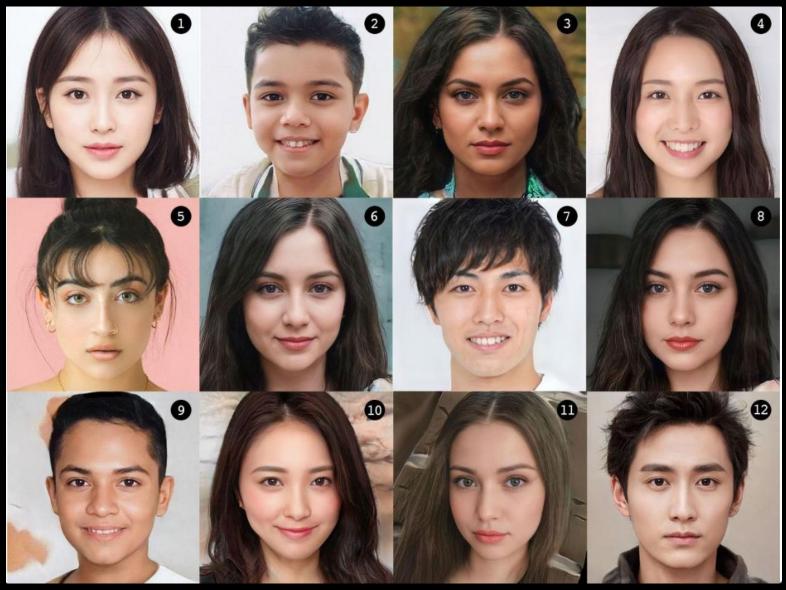
Chart: MIT Technology Review • Source: Strubell et al. • Created with Datawrapper

Multi modality: The next frontier



- Generative AI has advanced in various domains, including image/video generation, voice synthesis, text synthesis, and even molecule design.
- The next frontier of generative AI is multi-modality, where models can generate and combine multiple types of data simultaneously.
- Multi-modality refers to the ability of AI models to generate outputs that incorporate diverse modalities seamlessly.
- Multi-modal generative Al combines different modalities to create cohesive and expressive outputs.
- It enables the synthesis of complex and integrated outputs, going beyond individual components.
- Applications of multi-modal generative AI include multimedia content generation, interactive storytelling, virtual environments, and personalized recommendations.
- Challenges include designing effective architectures, utilizing multi-modal datasets, and developing specialized training techniques.
- Some other philosophical, social, legal challenges will arise: intellectual property, deep fakes and mass control. Overcoming these challenges opens up exciting opportunities for AI in understanding and generating multimodal data.

Did you make the right guess?



https://thispersondoesnotexist.com



Questions



