

# SAS Club 2023

Der Business Analytics Club für SAS User

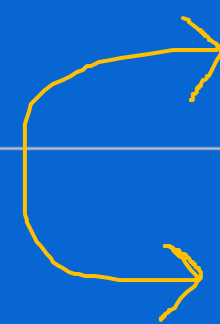
Wien, SAS Office Trabrennstraße  
19. Oktober 2023

Gerhard Svolba, Phillip Manschek, Jens-Ole Harden,  
Michael Weberberger (Premedia), Florian Stammer



# Agenda

14:15 - 14:20 Uhr	Begrüßung / Intro / News Gerhard Svolba, SAS
14:20 - 14:50 Uhr	Es geht auch anders! - Erstellung analytischer Modelle mit SAS Viya Gerhard Svolba, SAS
14:50 - 15:20 Uhr	SAS und Generative AI - Überblick, Entwicklungen und Anwendungsbeispiele aus dem Marketing Michael Weberberger, Premedia // Florian Stammer & Gerhard Svolba, SAS
15:20 - 15:35 Uhr	Die SAS Explore Konferenz in Las Vegas - Ein Vor-Ort Bericht Gerhard Svolba, SAS
15:35 - 15:55 Uhr	PAUSE
15:55 - 16:25 Uhr	Fuzzy Matching von Steuernummern in externen Datenquellen mit SAS Mihai Paunescu, Bundesministerium für Finanzen
16:25 - 16:50 Uhr	SAS Studio Analyst und die Erweiterungsmöglichkeiten mit Custom Steps Phillip Manschek, SAS
16:50 - 17:15 Uhr	SAS Tipps und Tricks Session Jens Ole Harden, SAS
ab 17:15 Uhr	Gemütliches Get-Together mit Buffet



# SAS und Generative AI - Überblick, Entwicklungen und Anwendungsbeispiele aus dem Marketing

Michael Weberberger, Premedia // Florian Stammer & Gerhard Svolba, SAS

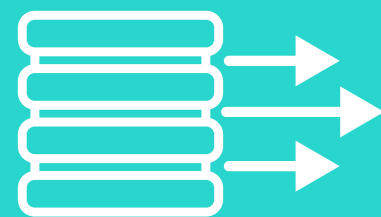


# Künstliche Intelligenz

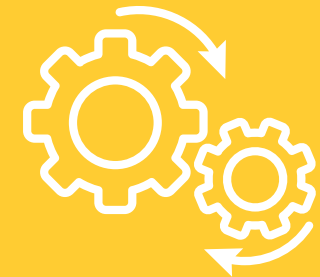
ist die Disziplin, Systeme zu trainieren, um Abläufe der menschlichen Tätigkeit durch LERNEN und AUTOMATISIERUNG zu emulieren.



Lernen aus  
Erfahrungen/  
Daten



Anpassen an  
neue Fakten



Automatisierung  
des Prozesses

**Statistik, Machine &  
Deep Learning**

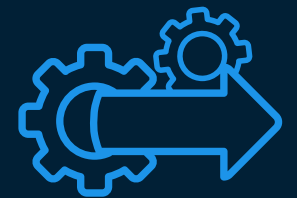
**Natural Language  
Processing**



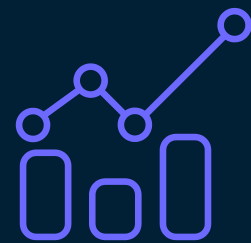
**Daten  
Management**



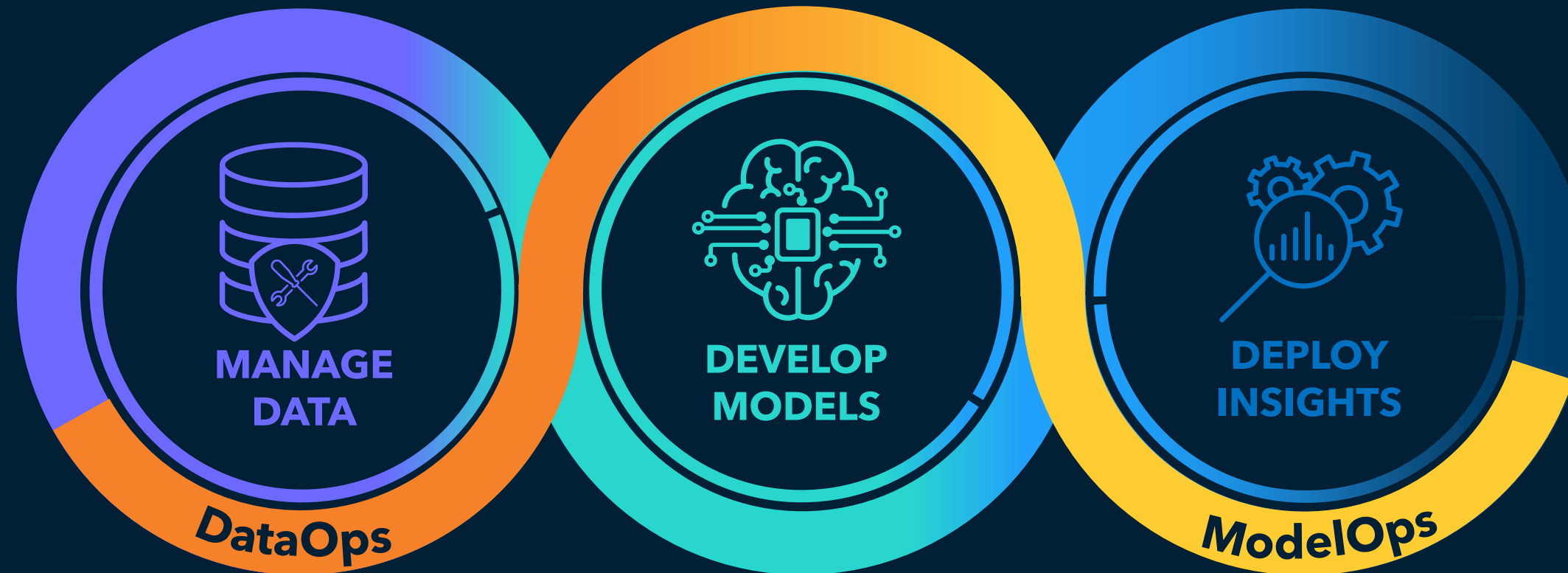
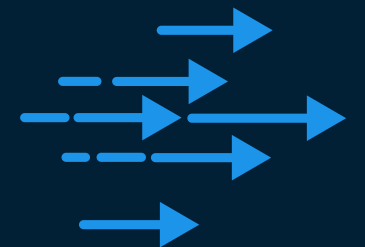
**Produktiv-  
stellung**



**Visualisierung**



**Entscheidungs  
Management**



**Computer &  
Machine Vision**



**Forecasting, Optimierung**





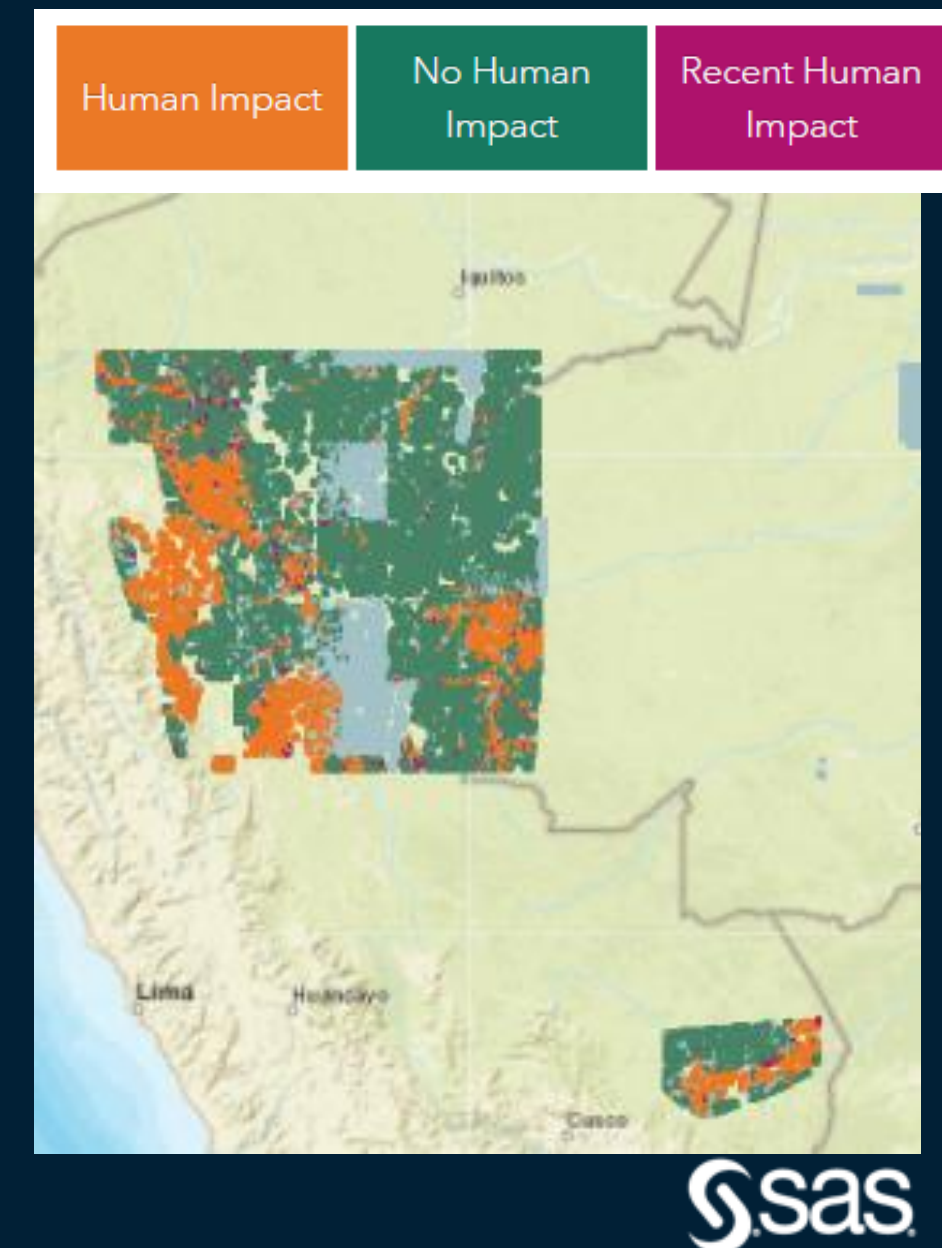
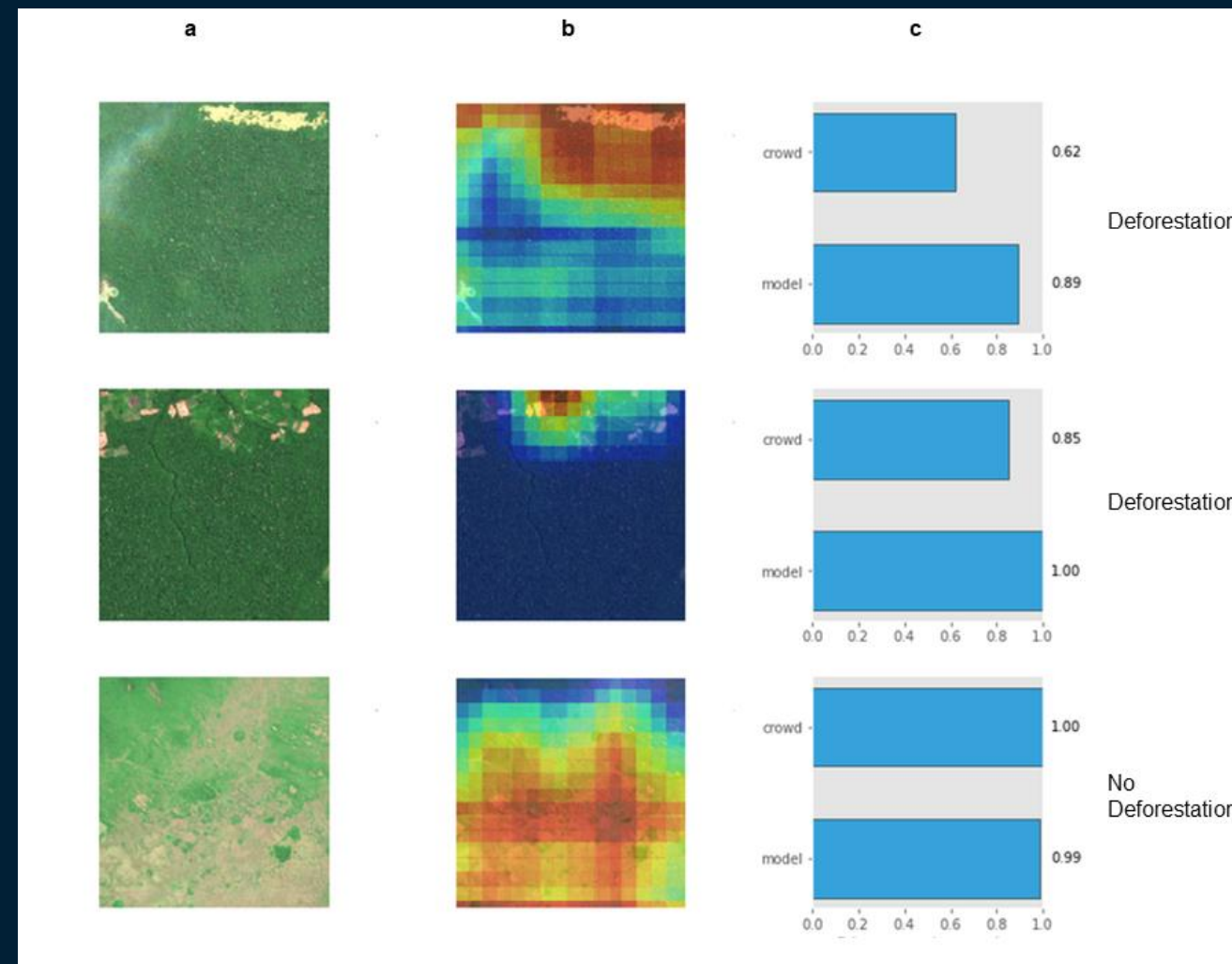
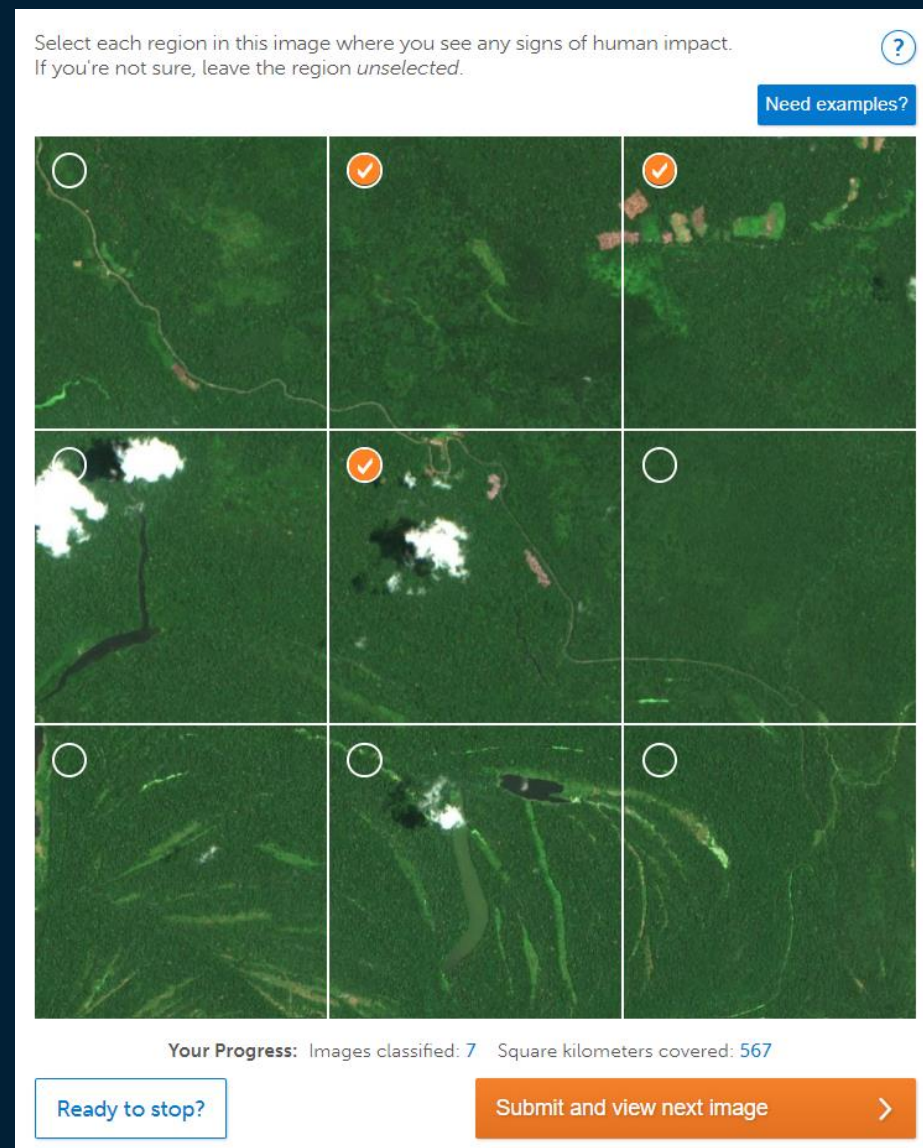
# Analyse der Abholzung des Amazonas Regenwalds auf Basis der automatischen Klassifikation von Satellitenbildern

(Kooperation mit der IIASA, International Institute for Applied System Analysis)

Crowd Sourcing  
("Labelled Data")

Modellierung mit  
SAS DLPY

Ziel: laufendes Scoring  
und Monitoring





A man with a beard and short brown hair is looking down at a smartphone in his hands. He is wearing a light blue button-down shirt and a brown backpack strap is visible over his shoulder. The background is a blurred city street with a large, curved architectural structure, possibly a train or tram, and other people in the distance.

# Text-Analytik (Natural Language Processing)

Einsatz von **Technologie**, um die menschlichen Handlungen des **Lesens**, **Organisierens** und **Quantifizierens** von Freiform-Texten auf sinnvolle Weise zu skalieren.

## Defining Generative AI (GAI)

Broadly speaking, whereas traditional AI/ML systems recognize patterns and make predictions, GAI systems learn real-world data to generate data – *like text, images, audio, tabular data, simulated data, code* – with similar probabilistic distributions and characteristics of the real-world data.

Given the transformative nature of GAI, business value, risk, and ethical considerations coexist.

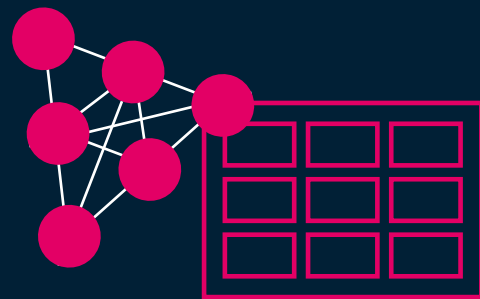
These three considerations have yet to be reconciled.

GAI regulatory controls and governance are fluid. Controls are critical.



While there is no official or generally agreed upon definition of Generative AI (GAI) in the market, SAS considers these three AI/ML categories to be generative in nature

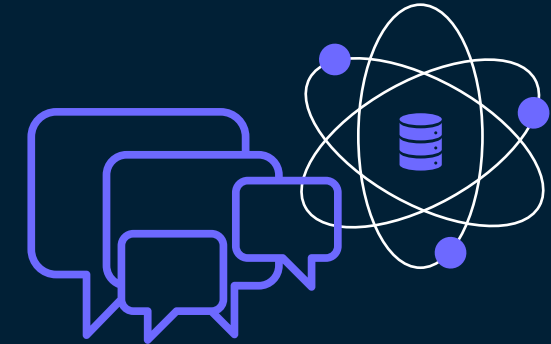
### Synthetic Data Generation



### Digital Twin



### Large Language Models



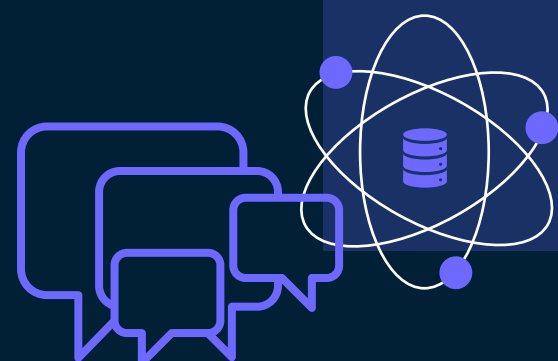
# What are Large Language Models (LLMs)?

A large language model (LLM) is a type of natural language processing (NLP) artificial intelligence (AI) model that is designed to process and generate natural language text.

These models are typically trained on massive amounts of text data, using techniques like deep learning and neural networks to identify complex relationships in language.

# SAS considers Large Language Models to be generative in nature given...

Algorithms like GPT, BERT, and recurrent neural networks (RNN) are used **to generate** text that is similar in style and content to the data the algorithms are trained on.



- + **Generate** embeddings or representations of text that can be used as inputs to downstream models
- + **Generate/scale** new insights by analyzing large amounts of text data and uncovering patterns and relationships that may not be immediately apparent
- + **Generate/combine** with other NLP techniques like clustering and topic modeling to help balance and improve unstructured data diversity

- + **Generate** text-based simulations of various scenarios like customer interactions or medical consultations for training purposes
- + **Generate/augment** data labels or annotations for real-world data in areas such as sentiment, topics, or entity resolution
- + **Generate/augment** text data to clean and normalize through spelling correction, format standardization, or the removal of redundant or irrelevant information



# Which methods are associated with Large Language Models?

## Sequence models

- Recurrent neural network (RNN) models like Long Short-Term Memory (LSTM) and Gated Recurrent Unit (GRU) were the gold standard in language modeling before transformers.
- They are called sequence models because sentences must be processed word by word.
- They're still used in speech recognition, music generation, sentiment analysis, machine translation, and DNA sequence analysis.

## Transformer models

- Transformer models are the current gold standard machine learning models in natural language processing. The first of these models, simply called Transformer was released in a paper called Attention is All You Need by Vaswani, et al., in 2017.
- They have the benefit of being **non-sequential** which means sentences are processed as a whole rather than word by word.
- **Self-attention** enables the model to look at the whole context of a sequence while encoding each of the input elements. It means the models will not "forget" parts of the input sequence because the window of retaining information is exactly as large as needed.
- Transformers feature positional embeddings, another innovation introduced to help replace recurrence. The idea is to use fixed or learned weights which encode information related to a specific position of a token in a sentence.

# NLP & LLM capabilities at SAS

## Current Capability

### BERT-based Text Classifier action set

Capture the context and meaning of words in a text to improve accuracy compared with traditional models. In addition to general classification, the BERT-based classification can be used to do sentiment analysis.

### RNN-based speech-to-text capabilities

Build n-gram language models and apply them to generate transcripts for speech-to-text applications. Evaluate the application's performance by calculating error rates of the transcription results.

## Future Development

As SAS continues its large language model research and development efforts, teams are maintaining a key focus on innovating responsibly and safely in this emergent technology area.

SAS currently *isn't* prescribing LLM solutions to customers given GAI regulation is fluid. However, SAS is collecting customer feedback on potential LLM use cases.

## Programming Guide

- [Introduction](#)
- [Examples](#)
- [LDA Topic Modeling Action Set](#)
- [Recurrent Neural Network Action Set](#)
- [Search Action Set](#)
- [Search Analytics Action Set](#)
- [Sentiment Analysis Action Set](#)
- [Smart Data Set Analysis and Processing Action Set](#)
- [Text Analytics Conditional Random Fields Action Set](#)
- [Text Analytics Rule Development Action Set](#)
- [Text Analytics Rule Score Action Set](#)
- **Text Classifier Action Set**
  - [Syntax](#)
  - [Examples](#)
  - Details**
- [Text Management Action Set](#)
- [Text Mining Action Set](#)
- [Text Parse Action Set](#)
- [Text Rule Discovery Action Set](#)
- [Text Summarization Action Set](#)

## SAS Visual Text Analytics Programming Guide

# Text Classifier Action Set: Details

Provides actions for classifying textual data

[Syntax ▾](#)
[Examples ▾](#)
[Details](#)

## About the Text Classifier Actions

### trainTextClassifier Action

[Overview of the trainTextClassifier Action](#)

[BERT and the trainTextClassifier Action](#)

[Input for the trainTextClassifier Action](#)

[Output for the trainTextClassifier Action](#)

[Using the Classifier for Sentiment or an Arbitrary Classification Task](#)

[Training Considerations and Memory Management for the trainTextClassifier Action](#)

[Reproducibility for the trainTextClassifier Action](#)

[Bibliography for the trainTextClassifier Action](#)

### scoreTextClassifier Action

[Overview of the scoreTextClassifier Action](#)

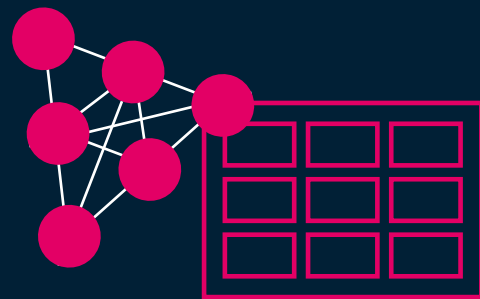
[Input for the scoreTextClassifier Action](#)

[Output for the scoreTextClassifier Action](#)



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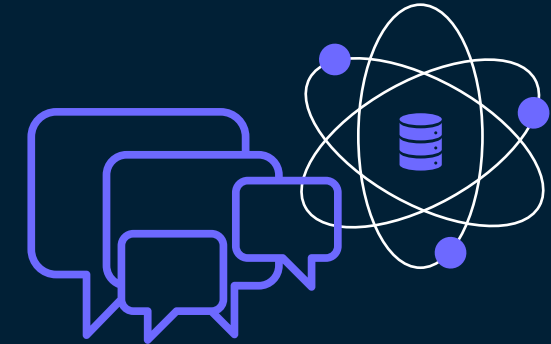
### Synthetic Data Generation



### Digital Twin



### Large Language Models



## Synthetic data is...

**Synthetic data** is on demand, self-service, or automated data generated by algorithms or rules, vs gathered in the real world, to meet conditions that may be lacking in real-world data.

Synthetic data reproduces the same statistical properties, probability, patterns, and characteristics of the real-world dataset from which the synthetic data is trained.

**Synthetic data** is considered to be a privacy preservation enabler.

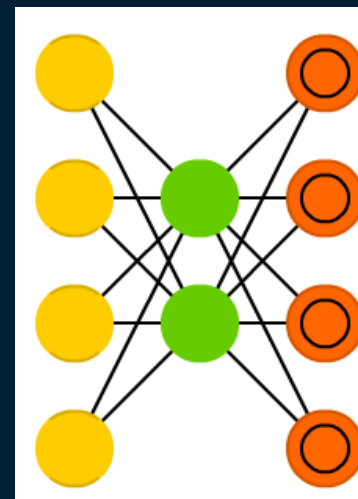
Many regulated industries are interested in using synthetic data.

# Gängige Methoden zur Datensynthetisierung



Multivariate Distributions

Auto  
Encoders



kNN k-nearest-neighbour

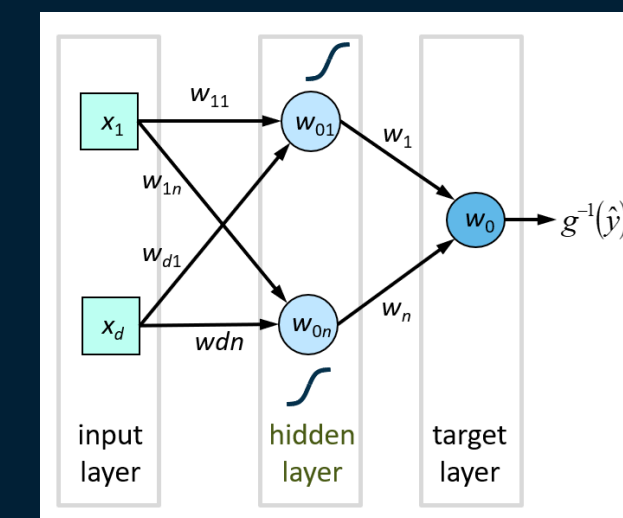
## Generative Adversarial Network Action Set Action Set

Provides actions for training generative adversarial network models.

Syntax ▾ Details ▾ Examples ▾ References

### Table of Actions

Action Name	Description
<a href="#">styleGanTrain</a>	Trains a styleGAN model
<a href="#">tabularGanTrain</a>	Trains a tabular GAN model

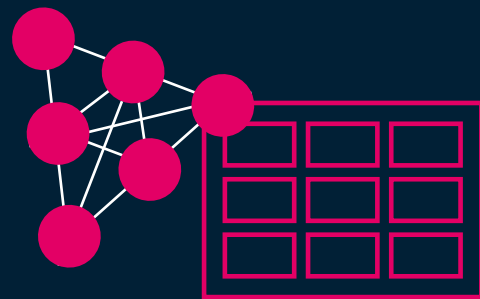


Generative Adversarial Networks 



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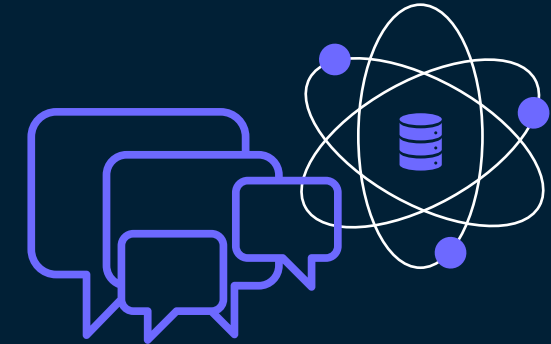
### Synthetic Data Generation



### Digital Twin



### Large Language Models



# A digital twin is...

Digital Twins

*A digital, animate, dynamic ecosystem – comprised of an interconnected network of software, generative & non-generative models, & (historical, real-time, & **synthetic**) data – that both mirrors & synchronizes with a physical system*

*Digital twins simulate “what-if” scenarios & stress test systems in the digital world to prescribe actions that optimize the physical world – to improve the lives of individuals, populations, cities, organizations, the environment, systems, products, & more*

# SAS success stories for digital twins

Digital Twins

MANUFACTURING

*Georgia-Pacific:*

*Optimizing the supply chain with analytics & IoT*

*“Because SAS makes it easy for people to do their own modeling, it’s translated into millions of dollars in value for us.” - GP*

150

manufacturing  
facilities

85K

vibration  
sensors

1TB

data generated  
/ day

15K+

models to calculate optimal production  
settings based on current business needs

10%

improvement in overall  
equipment efficiency

30%

reduction in unplanned  
downtime

[Click here to read more](#)

MANUFACTURING

*USG Corporation:*

*Optimizing time to value, production  
& modern manufacturing with analytics*

*Testing materials to ensure products meet quality standards:*

*Before SAS*

24+

hours to test

*With SAS*

Real-time

testing

*With SAS:*

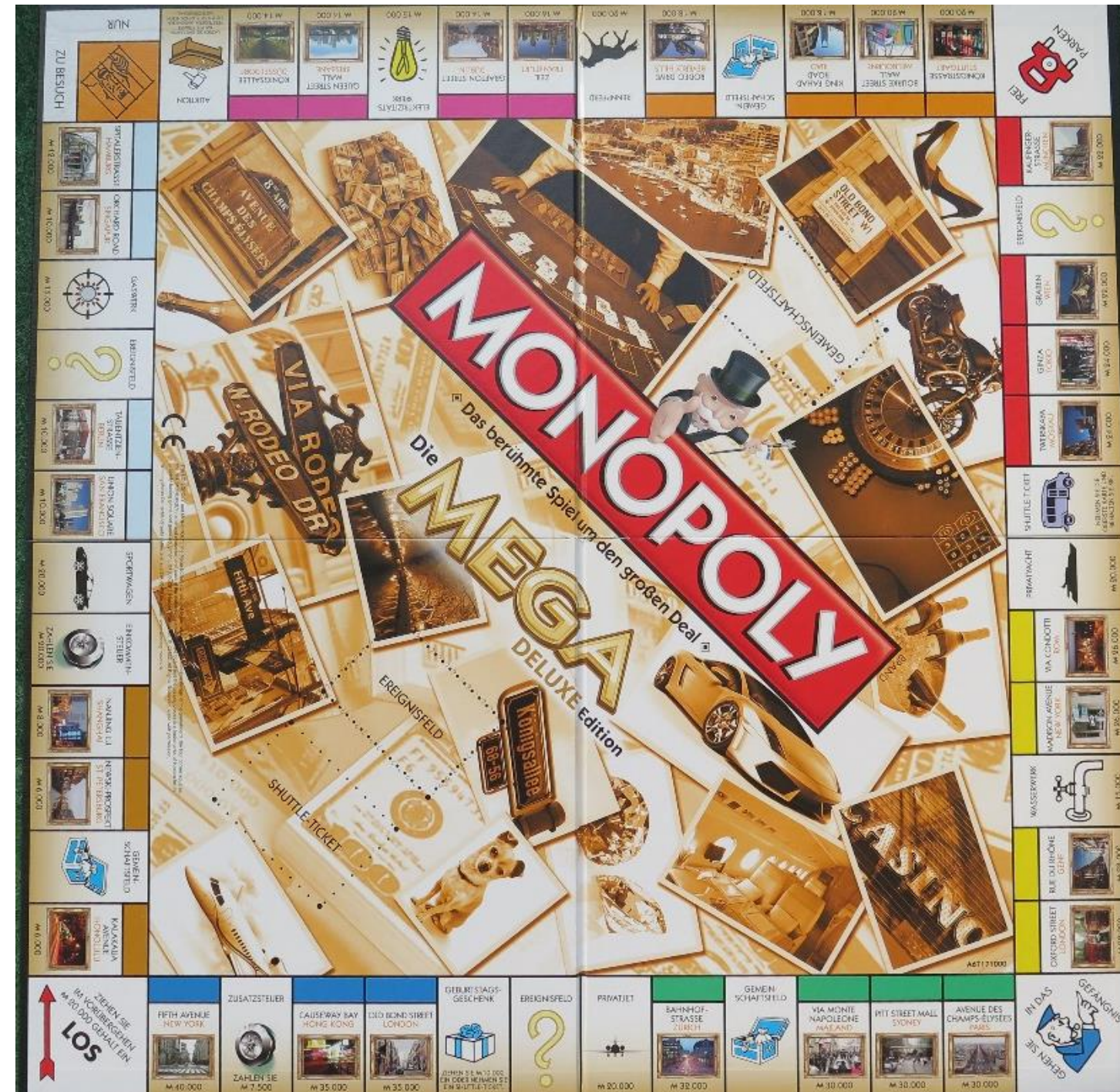
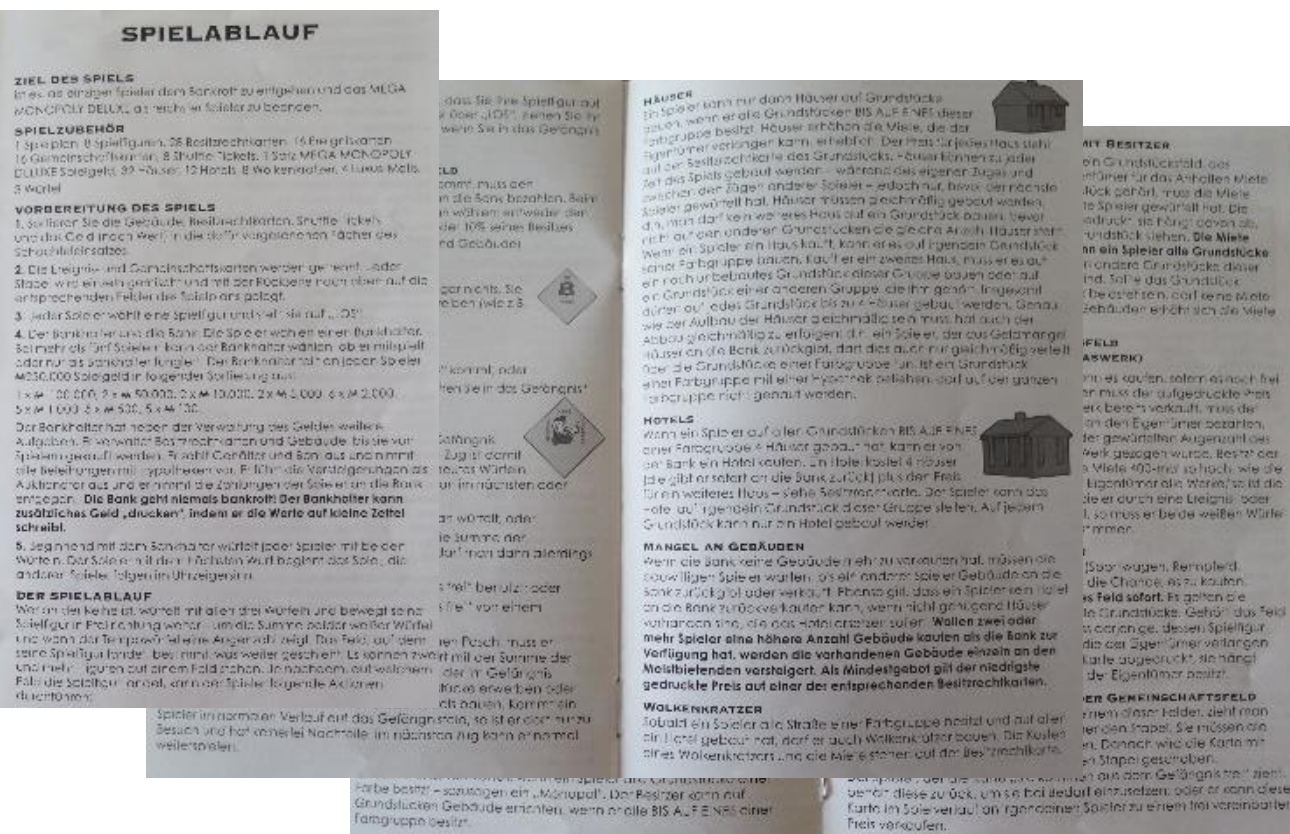
- + analyze plant inputs, such as flow rates & raw material additives, to predict quality outcome before production even starts
- + improve the quality, efficiency, safety, & cost of products

[Click here to read more](#)



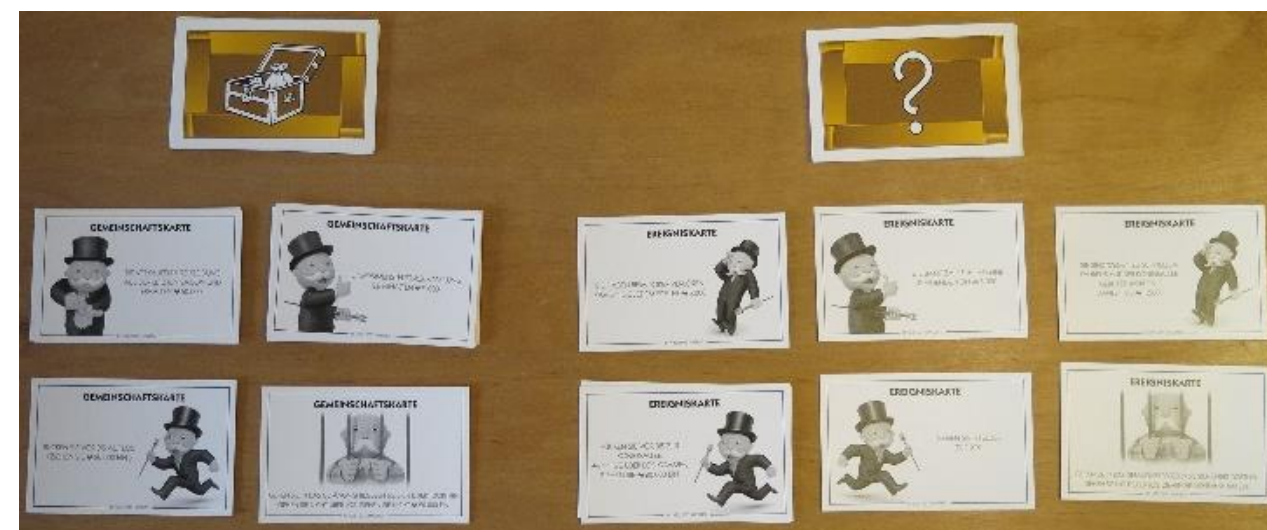


# Das Monopoly Spiel ist ein komplexes System



## Komplexe Regeln

## Monetäre Dimension



## Dynamische Komponenten

## Zufällige Komponenten

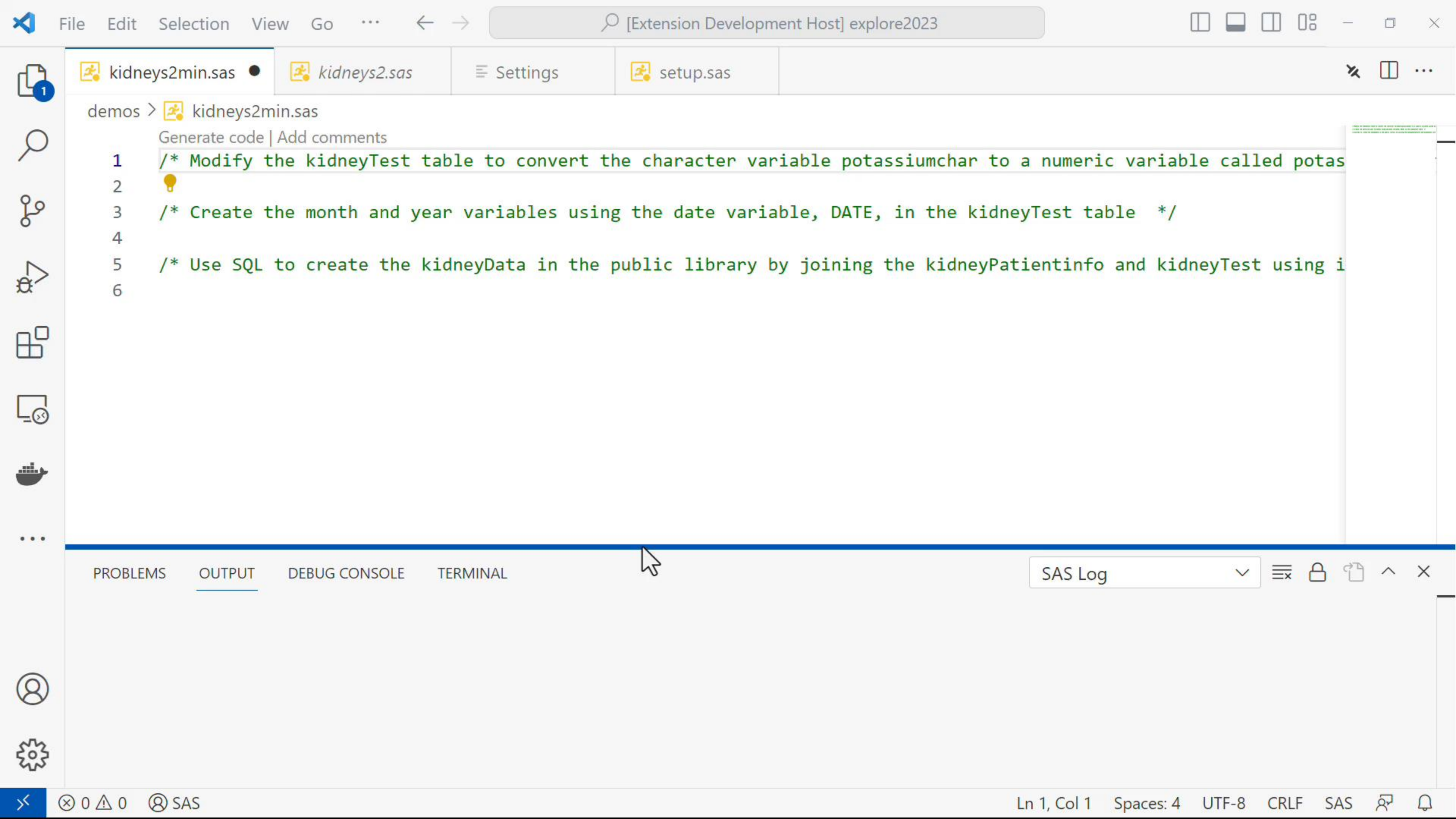


## Rahmenwerk von Möglichkeiten und Ereignissen

## Zusätzliche Anweisungen



# Applying Generative AI in SAS Software



kidneys2min.sas

kidneys2.sas

Settings

setup.sas



demos &gt; kidneys2min.sas

Generate code | Add comments

```
1 /* Modify the kidneyTest table to convert the character variable potassiumchar to a numeric variable called potas
2  *
3  /* Create the month and year variables using the date variable, DATE, in the kidneyTest table */
4
5  /* Use SQL to create the kidneyData in the public library by joining the kidneyPatientinfo and kidneyTest using i
6
```

PROBLEMS

OUTPUT

DEBUG CONSOLE

TERMINAL

SAS Log

