

Artificial Intelligence and Technical Services

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RPS BOARD MEMBER

QP & SPECIALIST PROJECTS NOVARTIS



Agenda

Introduction and Overview

Understanding AI

Machine Learning

Large Language Models

Regulatory Landscape

Al in Practice

Critical Thinking and Al

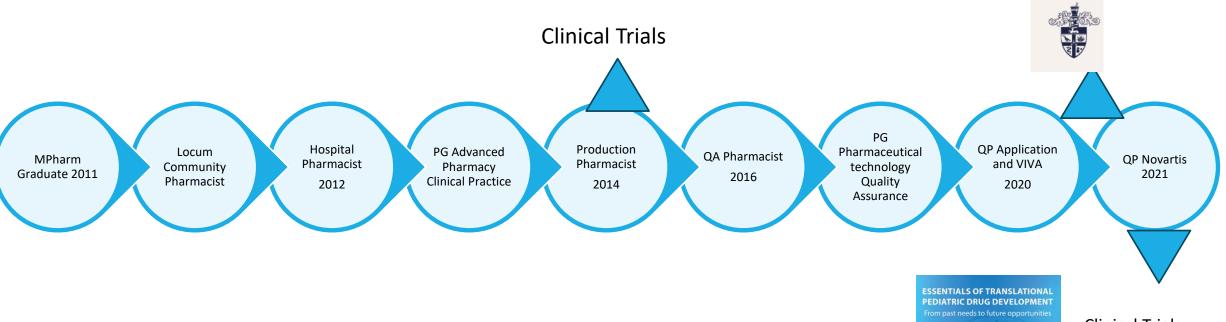
Conclusion



Looks can be deceviving: Always verify Al

Career Path





Investigational medicinal products considerations in pediatric clinical drug trials

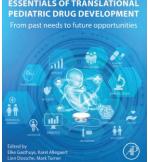
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Clinical Trials

+

Commercial Products

+

Specialist Projects

My Al Journey

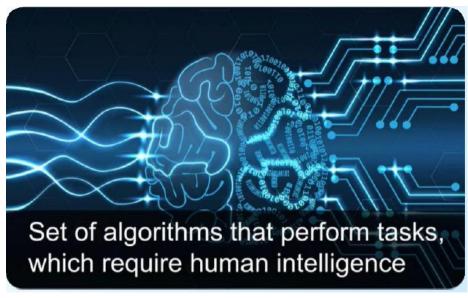


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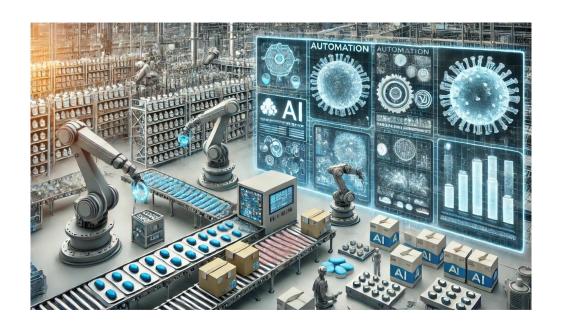
What is Artificial Intelligence (AI)?

Al, or Artificial Intelligence, refers to algorithms designed to mimic or simulate human cognitive abilities. This includes information processing, establishing relationships between multiple elements, learning and adapting to new concepts, making decisions, and adjusting to new situations.





Differences Between Al and Automation



AI VS AUTOMATION	AI - ARTIFICIAL INTELLIGENCE	AUTOMATION
Definition	A sector of computer science focused on creating intelligent machines that mimic human intelligence	A technique used to make a system operate automatically to accomplish repetitive tasks based on preset rules and commands
Key Capabilities	Problem-solving, speech recognition, learning, planning	Handles repetitive tasks without human intervention
Learning Ability	Has the ability to adapt and learn from new information	Lacks the ability to learn from processed data
Complexity of Tasks	Can cope with complex tasks involving decision-making or problem-solving	Best suited for straightforward, repetitive tasks
Interaction	Capable of interacting with humans and predicting user needs	Lacks human-like interaction
Data Analysis	Can analyze and interpret large volumes of data	Can process large data but lacks interpretation capability

Differences Between Al and Automation

Automation

(Past & Present)



Rule-based image checks (pixels, contrast, thresholds

Example: Detecting if a vial is empty or misaligned "If X, then Y — no learning, just ules

Discriminative AI (Now)





Machine learning models trained on "good" vs "defective" vials

Example: CNN classifies particles, cracks, or label defecta
Learns from data to classify defects

Generative Al (Emerging)



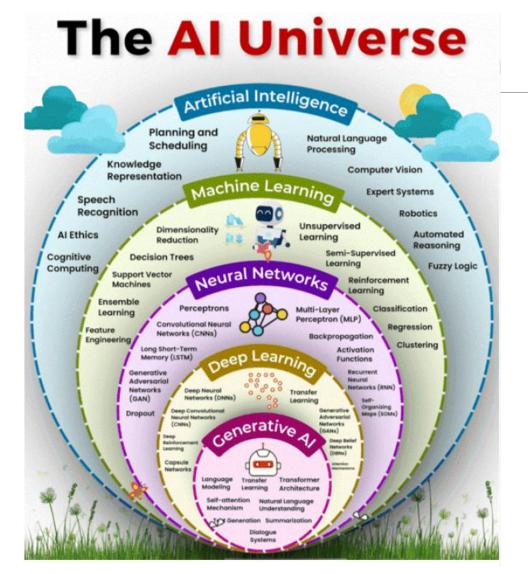


Generates synthetic defect images to improve training sets

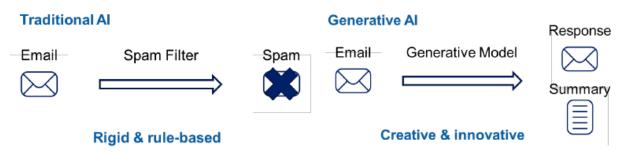
Example: Creating rare defect scenarios for validation

Creates training data, making inspection smarter

Al Terminology



Al Type	Discriminative (Narrow) Al	Generative Al
Overview	Implements predefined rules and applies machine learning for predictive analytics	Employs deep learning for tasks involving creativity and innovation, generating new, useful data from existing information
Example in Drug Manufacturing	Analyzes historical production data and optimizes assembly line operations, making real-time adjustments to enhance efficiency and reduce waste	Learns from a vast database of chemical reactions and designs and optimizes synthetic routes for drug manufacturing. It proposes the most efficient, cost-effective, and ecofriendly way to synthesize a drug
Key Applications II	Visual inspection, Monitoring, predictive maintenance	Content Generation Automation (Text, Tables), Draft Document generation,



Artificial Intelligence Models - Examples

Model Type	What It Does	Everyday Example
Transformer-based LLM	Learns patterns in language to understand, predict, or generate human-like text	ChatGPT writing emails, Copilot drafting documents
CNN (Convolutional Neural Network)	Recognises patterns in images by scanning for shapes, edges, textures	Google Photos tagging faces, Instagram filters
GNN (Graph Neural Network)	Learns from networks (like social or molecular graphs); considers relationships between items	LinkedIn friend suggestions, fraud detection
Knowledge Graph	Connects pieces of information as relationships in a web of knowledge	Google Search showing "related people" or facts
Reinforcement Learning	Learns by trial and error -rewards success, punishes failure	YouTube recommending videos based on watch behaviour
Collaborative Filtering	Looks at what similar users liked to predict what you'll like	Spotify suggesting songs, Netflix recommending shows
Recommendation System	Combines filters, behaviour, and rewards to make smart predictions	Amazon suggesting "People who bought this also bought"

Machine Learning

Machine learning is a subset of artificial intelligence (AI) that involves the development of algorithms and statistical models enabling computers to learn from and make predictions or decisions based on data.



Machine Learning

Reinforcement Learning

Learn to make sequential decisions to maximize rewards

Eg: Chess Al.

Unsupervised Learning

Discover Patterns, relationships or groupings with unlabeled data

Eg: Recommendation system used by websites like Amazon and Netflix.

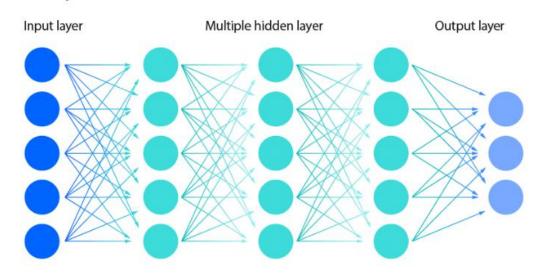
Supervised Learning

Predicts or Classify based on labeled data

Labeled data with input output pair Eg:Spam emails

Machine Learning: Neural Networks

Deep neural network



 Neural networks are brain-inspired models that power AI to spot patterns and make decisions from data.

Deep networks pack many layers and millions of parameters, making them powerful—but hard to explain from input to output.

Large Language Models

- LLM: A Large Language Model is a type of neural network model that has been trained on a vast amount of text data to understand and generate human language.
- LLMs can perform a variety of language-related tasks such as text completion, translation, summarization, and question answering.
- Generative Pre-trained Transformer GPT is a specific implementation of a Large Language Model developed by OpenAI Trained on large text corpora





Limitations and Risks of Large Language Models (LLMs)

Unpredictable Outcomes: Reliance on probability models can produce unexpected results.

Lack of Transparency: Hidden layers in deep neural networks are difficult to interpret.

Dependency on Prompts: Output accuracy is heavily influenced by the quality and clarity of given prompts.

Data Exposure: LLMs require relevant data to generate accurate results; insufficient exposure increases error risk.

Bias Amplification: LLMs can inherit and amplify biases from their training data, potentially causing unfair or discriminatory outcomes.

Overfitting: Excessive fitting to training data may result in poor p erformance on new,unseen data.



Regulatory Hierarchy

LAW

Reg Guidance eg: MHRA, Eudralex, ICH, PIC/S

Consensus standards eg: ISO, BSI

Evolving Interpretations eg: Inspection finding

Professional Guidance RPS, ISPE, PDA

Company Policy and Procedures

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EU Regulatory Landscape

EU AI Act – <u>Artificial Intelligence Act 2024/1689</u>

Global First

Risk-based approach

Phased implementation;

1 Aug 2024 \rightarrow Act entered into force.

Feb 2025 → Prohibited AI practices banned (e.g., social scoring, manipulative AI).

Aug 2025 \rightarrow General Purpose AI / foundation models rules apply (like GPT-style models).

Aug 2026 → High-risk AI rules apply (covers pharma, healthcare, medical devices, critical infrastructure).

Aug 2027 \rightarrow All provisions fully applicable.

Key obligations

Governance

- Risk management and quality management systems
- Transparency, logging, and record-keeping

Data & Technical Standards

- Data governance, traceability, bias controls
- Accuracy, robustness, and cybersecurity

Human Oversight

Human-in-the-loop where required

Model Scope

- GPAI providers: transparency and risk mitigation
- Systemic-risk models: enhanced obligations for very large LLMs

RoW



Al Regulation - Rest of the World



United States

- No single federal AI law
- Executive Orders + NIST AI Risk
 Management Framework (voluntary)
- Sector regulators (FDA, FTC) shaping Al oversight



Canada

 Artificial Intelligence & Data Act (AIDA) tied to Bill C-27



Singapore

Model Al Governance
 Framework + Al Verify
 –practical, non-binding
 but detailed tools



- 'Soft law' approach
- Interim government response (2024)

. (

China

- Generative Al rules'(Aug 2023) binding for providers (registration, transparency, content controls)
- Draft Al Law (2024–25) under discussion full framework likely coming

Japan

- "Soft law' approach
- Principles & guidelines focused on Innovation + international
- alignment (67 Hrrognapoe)



- No omnibus Al law
- MeitY advisories (2024–25)

Maria Australia

- Safe & Responsible
 Al programme
- Interim government response (2024)
- No singlu Act yet, but risk-based guidance coming

Why does it matter? LLMs

US	Europe	China	Other
OpenAI → GPT-4, GPT-4o (ChatGPT, Copilot)	Mistral (France) → lightweight open LLMs	Baidu → ERNIE Bot	Cohere (Canada) → enterprise LLMs
Anthropic → Claude models	Aleph Alpha (Germany) → Luminous	Alibaba → Tongyi Qianwen	xAI (US) → Grok
Google DeepMind → Gemini		Huawei / iFlyTek / SenseTime → sector models	Stability AI (UK/US) → Stable Diffusion (images)
Meta → LLaMA family (open source)			MidJourney (US) → image generation





Contents

- 1 Recommendations
- 2 Introduction
- 3 Scope for AI in pharmacy practice
- 4 Considerations for AI in pharmacy practice
- Development and Collaboration

2023 Al White Paper

principles

2024 Gov Response 5

Regulators ICO GDPR, minimisation, explainability

NHS 2025 AI & IG Guidance

2024-25 AI Regulation Bill Draft in Lords

Regulators MHRA Als as devices

NIHR Innovation Observatory

Consultation outcome

иттедицион, и рто питотицоп ирргоист рокез ргорозил

Department for

& Technology

Science, Innovation



A Horizon Scan of Algorithms Used in the Development of Artificial Intelligence (AI)-Enabled **Healthcare Technologies**

A pro-innovation approach to Al

regulation: government response

Authors: Imogen Forsythe, Elizabeth Green, Olushola Ewedairo, Andrew Mkwashi Date: July 2024

> NICE AI & Digital **Regulations Service**

Caldicott Principles apply



Medicines & Healthcare products

of medical products

Impact of AI on the regulation

Implementing the Al White Paper principles

Policy paper

New approach to ensure regulators and regulation support growth (HTML)

GMP -EMA Reflection Paper on Al in Drug Lifecycle

Reflection paper issued by EMA on use of Artificial Intelligence (AI) in the medicinal product lifecycleSept 2024.

Manufacturing Considerations:

Model Development & Assessment:

Follow quality risk management principles

Consider ICH Q8, Q9, and Q10 principles.

Revisions to current GMPs are anticipated.

Responsibility:

Marketing authorization holders must ensure compliance with GxP standards.

Early regulatory interaction and scientific advice recommended if AI impacts drug benefit-risk balance.

Practical Applications & Regulatory Challenges:

EMA meeting:

Explored AI and digital technology applications in drug manufacturing.

Discussed regulatory challenges and solutions for AI and ML in production settings.

Emphasized aligning AI applications with current and evolving GMP standards.



EU GMP

Document / Annex	What's Changing	Implications (for Pharma / NHS / GMP Teams)
Annex 11 (Computerised Systems – revision)	Tighter requirements for lifecycle management, risk assessment, supplier qualification, audit trails, e-signatures, data integrity.	More rigorous supplier oversight; stronger validation of computerised systems; audit readiness will focus on data integrity and system lifecycle evidence.
Chapter 4 (Documentation – revision)	Documentation expanded beyond paper → must cover hybrid, digital, image, audio, video. Stronger links to risk management and electronic records.	Teams need to manage diverse formats (not just paper/PDF). Expect inspectors to check governance of digital records, metadata, and long-term accessibility.
Annex 22 (NEW – Artificial Intelligence)	Guidance for use of AI/ML in GMP systems. Covers validation, oversight, human review, and risk to product quality/patient safety.	Any AI tool that touches GMP decisions (e.g., batch release support, deviation triage) will need validation and human oversight. Use of "black box" AI will be questioned by regulators.

Part I EU GMP, Annex 15, Annex 16, PIC/S P1011 ICH Guidelines - ICH Q8, ICH Q9, ICH Q10

Example responses to consultation

	Comment (only one topic per comment) (max 600 characters)	Rationale (must be included when proposing a change) (max 600 characters)	Proposed wording (must be included when proposing a change) (max 600 characters)
Annex 22	Include validation of data sources and acquisition context in the intended use description.	that data comes from verified and trusted sources to support GMP	This should include a comprehensive characterisation of the data the model is intended to use as input, including its source systems, acquisition methods and all common and rare variations within the sample space.
	Add requirement to assess potential model bias in test plans.	•	The test plan should also include a bias impact assessment where input data includes operator, equipment, process-drift or sitespecific attributes.
	Comment (only one topic per comment) (max 600 characters)	Rationale (must be included when proposing a change) (max 600 characters)	Proposed wording (must be included when proposing a change) (max 600 characters)
Annex	Align wording with Annex 22/EU AI Act principles regarding transparency and human oversight	To future-proof the guidance, references to model transparency, explainability and human-in-the-loop decision-making should be embedded, particularly during lifecycle updates.	'In systems incorporating AI, organisatiation uld ensure that transparency, human oversight and explainability are maintained throughout the lifecycle, and in accordance with Annex 22 Guidance.'
	Define responsibilities for reviewing Al system updates (e.g., model drift or retraining)	There is no mention of roles responsible for monitoring AI system changes, which is critical for sustained GMP compliance and ensuring proper oversight by qualified personnel.	Periodic review of system requirements should include inputs from qualified personnel including QP, QA, and AI system owners to assess impact of AI model updates or drifts on GMP compliance.'
Ch	Comment (only one topic per comment) (max 600 characters)	Rationale (must be included when proposing a change) (max 600 characters)	Proposed wording (must be included when proposing a change) (max 600 characters)
	Explicitly define how risk-based approaches apply to Al-enabled documentation systems	Risk controls will vary depending on whether Al is involved in decision- making, review or data summarisation.	Risk-based approaches should also account for whether AI systems are used in data interpretation, generation, or validation processes."
4	Lifecycle controls should consider Al model lifecycle, not just data.	Al/ML systems evolve over time, requiring traceability not only for data but also for model training, retraining and deployment steps.	Controls should ensure traceability and integrity across both data and AI model lifecycles, including training data, model versions and algorithm changes.

Artificial Intelligence for Drug Development

Artificial Intelligence for Drug Development

What is Artificial Intelligence (AI)?

Artificial Intelligence (AI) refers to a machine-based system that can, for a given set of human-defined objectives, make predictions, recommendations, or decisions influencing read or virtual environments. AI systems (1) use machine- and human-based inputs to perceive real and virtual environments, (2) abstract such perceptions into models through analysis in an automated manner, and (3) use model inference to formulate options for of: 02/20/2025

> Product(s) Biologics

FDA's Interaction with AI	Details
Historic Regulation	FDA has a historical track record of regulating AI largely in the context of medical devices, commonly featuring static algorithms.
Submissions	Over 300 submissions for drugs/biological products with AI components have been reviewed by FDA. (Source: Robert M. Califf, M.D., FDA Commissioner)
Approvals	The FDA has granted approval, clearance, or authorization for more than 880 Al/ML-enabled devices. (Source: FDA webpage)
Recognized Benefits	Al is acknowledged by FDA for its potential in improving product quality, reducing manufacturing defects, and mitigating drug shortages by addressing quality issues.
Current Challenges	Regulating generative AI, owing to its relatively recent emergence, poses a modern regulatory challenge for the FDA.

Regulatory Landscape-FDA

FDA Discussion Papers

Using Artificial Intelligence and Machine Learning in the Development of Drug and Biological Products

Background and Scope: The FDA's Center for Drug Evaluation and Research (CDER) discusses the use of AI/ML in drug development to enhance safety and innovation.

Current Uses of AI/ML: AI/ML is applied in drug discovery, nonclinical and clinical research, postmarket safety surveillance, and advanced pharmaceutical manufacturing.

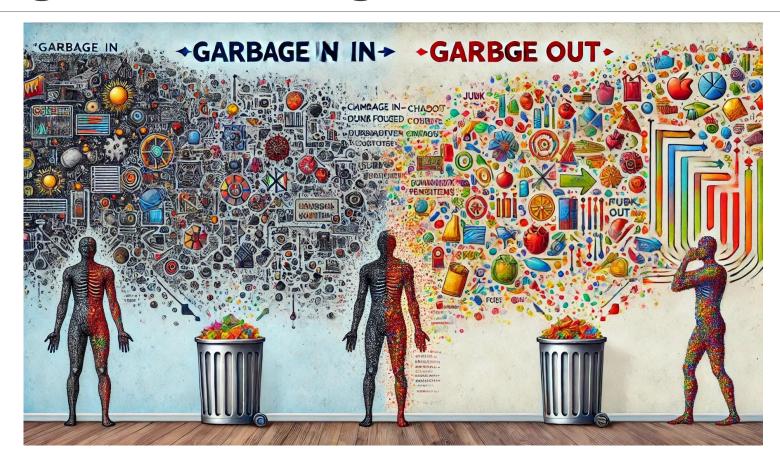
Considerations: The document outlines standards and practices for AI/ML use, addressing concerns like bias, data integrity, and transparency1.

Next Steps: FDA seeks feedback and collaboration with stakeholders to refine the regulatory landscape for AI/ML in drug development.

Artificial Intelligence in Drug Manufacturing

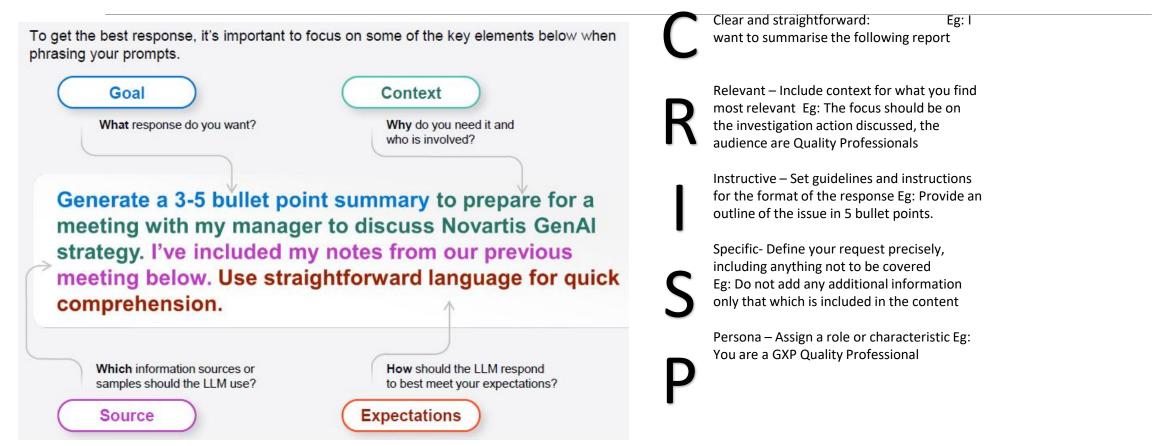
- **Purpose:** The paper aims to gather public feedback on the use of AI in drug manufacturing to help develop a future regulatory framework.
- Mission: CDER's mission is to ensure that human drugs are safe, effective, and meet quality standards.
- Advanced Manufacturing: The paper discusses the potential of advanced manufacturing technologies to improve drug production processes.
- **Public Feedback:** The FDA seeks input on various areas related to AI technologies in drug manufacturing to inform future regulations.

Garbage In= Garbage Out



What is a prompt?

The queries and commands that you can put into an AI to get a generated response.



Copilot- General

How Copilot Helps in Microsoft 365



Meeting Management

What Copilot Does:

Auto-generates meeting summaries and suggests actions

Key Benefits:

- Quickly catin up on missed meetings
- Clear next steps and owners
- Fewer redundant meetings



Email & Scheduling

What Copilot Does: Summarizes emall threeds, drafts agendas and replies, schedules meetings with click

Key Benefits:

- Saves multiple manual steps
- Faster agenda creation and responses
- Streamlined scheduling



Information Retrieval

What Copilot Does:

Extracts and summarizes key points from long emails and chat conversations

Key Benefits:

- Faster context gathering
- Highlights decisions, risks, and asks



Overcoming Creative Blocks

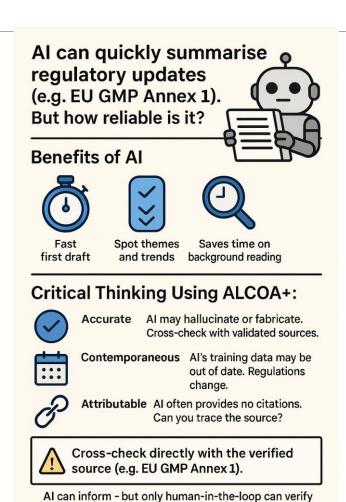
What Copilot Does:

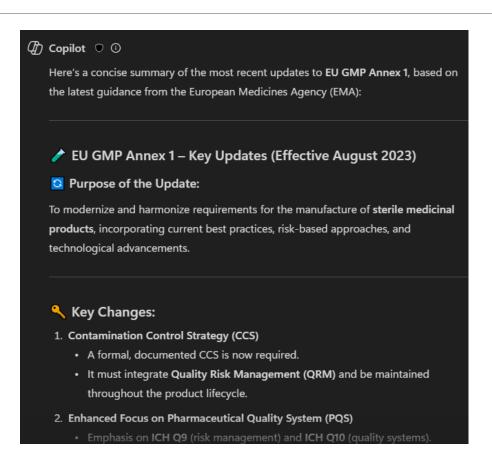
Rephrases and clarifies ideas; suggests alternative tones

Key Benefits:

- Improves clarity and communication
- Eases drafting and editing

CoPilot/GPT for Regulatory Summary





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CoPilot/GPT for SOP Drafting



Al drafts a new local SOP for cleaning validation based on user prompts.

Critical Thinking Using ALCOA+



Accurate No reference to EMA or FDA guidance.



Mimics existing SOPs but lacks formatting for version control and metadata.



Complete Mimics existing SOPs but lacks formatting for version control and metadata.



Legible Language is clear but lacks regulatory terminology.



Use Al as a starting point, then validate with SMEs and SOPs.





Al can assist but only human can ensure compliance.

Gen Al Examples

Deviations

Feature 1 — Proofreading (Deviation Module): Checks compliance to SOP-; highlights missing requirements; enforces data consistency; validates Description (5 Whys + 2 Hs), material/batch, dates, escalations.

Feature 2 — RCA Tools & Summary (Deviation Module): Builds sequence of events; supports 5 Whys & 6Ms analysis; produces deviation and executive summaries for clear decisions.

Disclaimer

The content in this window was created by Generative AI and serves as a job aid. Be aware that you are responsible for the final content of your record and the execution of any procedurally mandated review steps

SOPie: Chat bot for SOPs

Complaints close out letter

Quality Insights Portal

Outside of Quality

Al in Drug Discovery & Pharma

Al-Driven Drug Discovery



Virtual screening Automated molecular design



Reduce tirnesst

Personalized Medicine







genomic proclinical data Improved efficacy, minimized adverse effects

Explainable AI (XAI)







ausential

transparency

Trials

In Silico Clinical



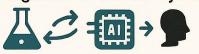
Complement and partially replace traral

Drug Repurposing Acceleration



Reduces development timelines and costs

Integration of Wet & Dry Labs



Merge experimental bjology wit computational modelling

Federated Learning for Data Sharing



Collaboration across institutions without sharing raw data

Predictive Trial Design



Increasing success rates irduce phase III failures

Ethical & Regulatory Evolution





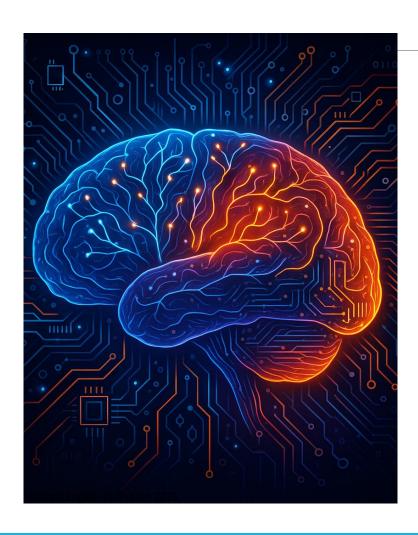
New frameworks addressing chalienges data privacy, apprval, employment, IP+IF

Screening and diagnosis –NHS Cloud

World-first AI system –NHS Patient Safety Concerns

Prescribing Practices

What is Critical Thinking



The objective, systematic, and rational analysis and evaluation of factual evidence in order to form a judgement on a subject, issue, etc...

Critical thinking is our human edge in an AI world

- ☐ Understanding & Evaluating Arguments
- ☐ Awareness of Cognitive Biases

Constant Noise & Speed

Urgent emails. Real-time data. Instant messages

→ Distraction is the norm

Decisions Under Pressure

Risk being shaped by bias & emotion

→ Reflection is often skipped

Complex Challenges

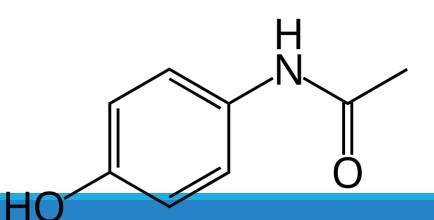
Unprecedented times go beyond experience or intuition

→ They demand deeper thinking

AI: A Double-Edged Tool

Powerful but can be flawed

→ Unchecked flawed AI data leads to flawed decisions



The Two Minds of Critical Thinking



Daniel Kahneman's "Thinking, Fast and Slow" Fast Thinking (System 1)

Automatic, unconscious, effortless. Rooted in basal ganglia

Examples: Speaking your native language, recognizing faces, reacting instinctively

- Strengths: Speed, efficiency, pattern recognition
- Weaknesses: Prone to biases, errors, and emotional reasoning



Automatic thinking is not just simple repetition

It's powerful and capable of learning nuanced, complex patterns-like our spoken language, without our conscious awareness.

The Two Minds of Critical Thinking

Daniel Kahneman's "Thinking, Fast and Slow"

Slow Thinking (System 2)

Deliberate, conscious, effortful. Tied to the hippocampus/prefrontal cortex

Examples: Used for reasoning, problem-solving, planning, and self-control

- Strengths: Analytical and logical thought
- Weaknesses: Slower, mentally taxing, limited capacity

What thinking habits do you rely on most?"

"Have you ever regretted a fast decision?"

"When was the last time AI helped/hindered your thinking?"

Power lies in observation





But beliefs are hard to change

- Beliefs resist change due to dopamine rewards from confirming info.
- ☐ Confirmation bias: we seek trust in supportive evidence; contradicting facts feel unrewarding.
- ☐ Changing beliefs requires effort; the brain prefers cognitive ease and consistency.
- ☐ Use AI to challenge assumptions: Devil's Advocate, reflective partner, "rubber duck."

Think Critically

Building Quality Through Mindset + Tools

Mindset

Stay Curious



Practise intellectual humility. Keep an open mind and question assumptions.

Question Assumptions

- → Seek Challenge
- → Be Open to Change

Collaboration



Leverage Team Diversity

Invite different perspectives to strengthen decisions.

Methods & Tools



6W+H

(Who, What, When, Where, Why, How)



ALCOA+

(data integrity check)



Ask thought questions



Evaluate evidence critically

C Challenge assumptions





E Evaluate critically



A Ask questions



Reflect before deciding



Looks can be deceiving: Always verify Al

Utilising AI: Strategies for Success

Evaluate Proficiencies and Constraints

 Analyse the strengths and weaknesses of Al and human capabilities.

Strategic Task Assignment

 Delegate tasks based on whether humans or AI are better suited, optimising efficiency.

Human-Al Collaboration

 Implement AI as a tool to augment human efforts, fostering a synergistic approach.



Take responsibility for your own education

Why this matters



Empathy & compassion

- • Judgment & wisdom
- • Creativity & imagination
- Ethics & values
- Human connection & dignity

Al can't replace this — but it can help us protect it.