Al in Drug Safety Operations - The Future is Here!

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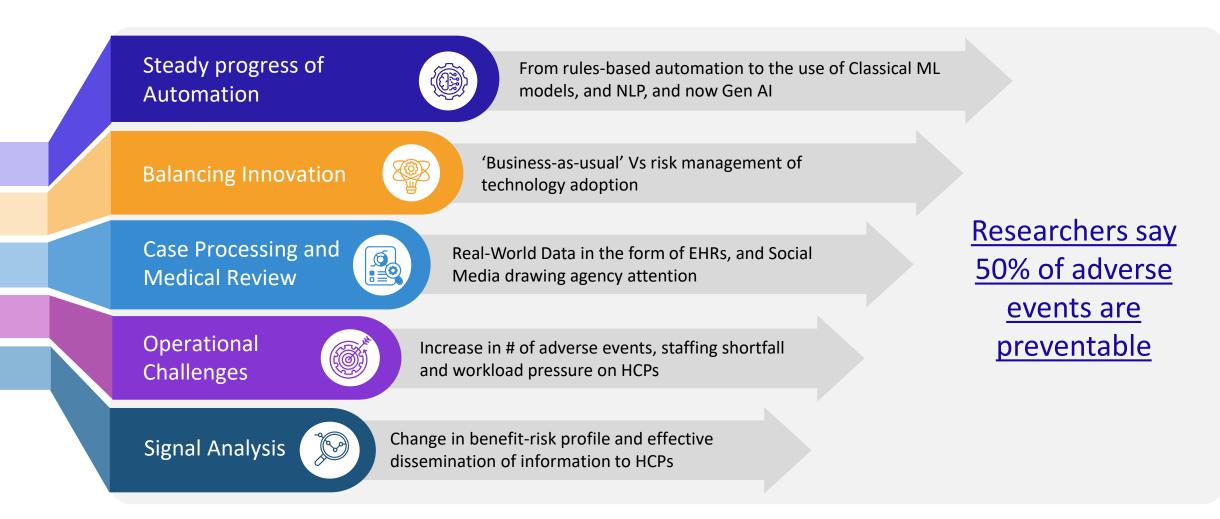
Chief Product Officer at Datafoundry AI



An Al-first company leveraging technology for better health outcomes

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Leveraging Technology for Patient Safety



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Accelerate technology adoption to improve health outcomes

Al Adoption – Navigating through the Hype

Compliance, Security and Privacy:

- Using 'Data' as evidence
- Build stakeholder trust
- AI/ML Validation strategy

Accuracy & Explainability:

- Interpretability
- Mitigate 'Bias' and 'Hallucination'
 - Change Management



Data Quality:

• Automation of data standardization

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- 4Vs Veracity, Variety, Velocity and Volume
- Data quality monitoring

Cost-Benefit Analysis:

- Choosing the right automation technique – PoC to Production
- Operationalizing and Scaling

Domain-focused Data Science is the Game Changer

Gen AI Example – The Future is Here!

- Gen AI will transform regular operations such as case intake, processing, narrative and report generation
- Human-in-the-loop will be required to confirm/validate medical review and analysis.

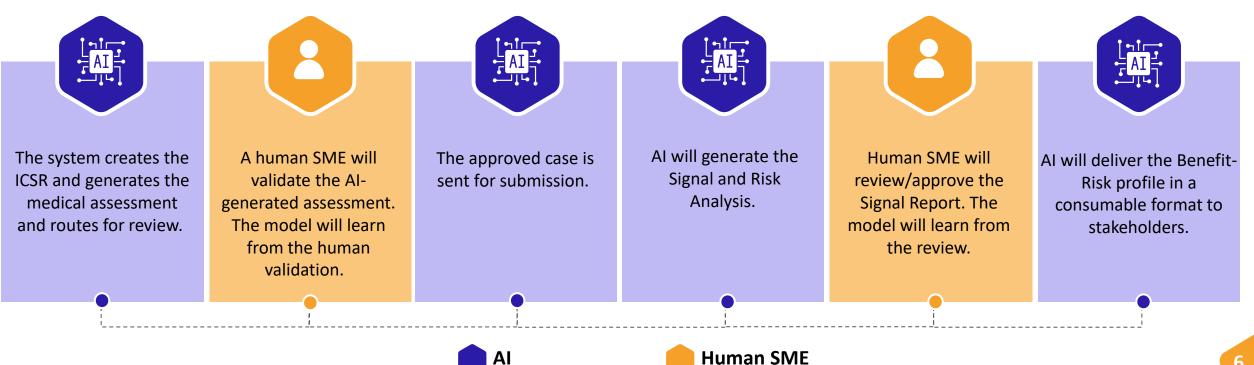
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DRUG NEFFECTIVE FOR UNAPPROVED INDICATION	1 10043 19501 49265077	NA	ня.	0.345	NA.	-1.423	0.345	1.493	NA	0.215	-1.663
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An example of LLM generated Signal Detection

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Safety Operations – Prepare to be transformed

The case data is fed into a multi-source and multi-modal system, processed by a **Gen Al-driven workflow**



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Al-assisted Use Cases – Disrupting Traditional Processes

EHRs could potentially become the single source of reporting globally Goal would be to ensure data quality when reporting for the first time Apps for safety reporting can facilitate: • AE reporting by HCPs and Patients • Voice to Text to E2B R3 • Export E2B XML into the Safety system • Product complaints tracking • Access label information in digital format

• Verify the drug source – Spuriousness check

Companies receive drug purchase and usage data

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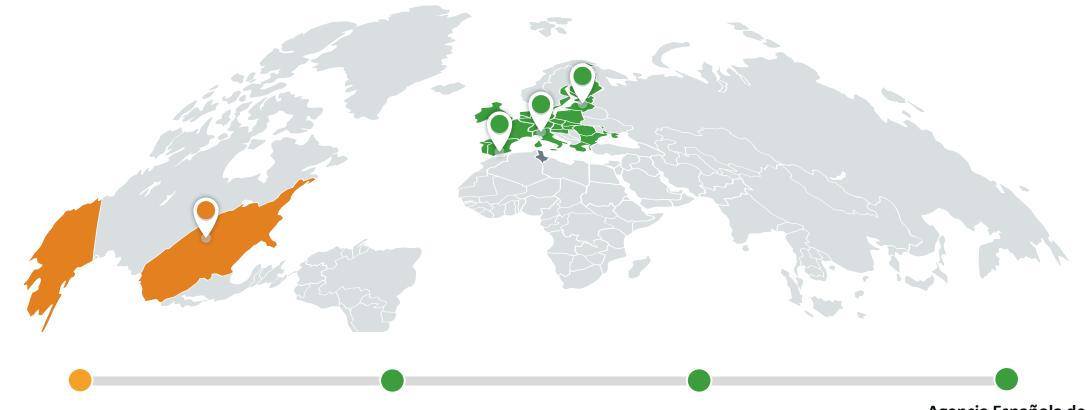
• Chat with AI agent

A combination of human + AI agents can do the follow-ups

Medical reviewers can focus on complex cases as the case processing time is reduced by 80%

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Regulatory Authority Initiatives



US FDA - CBER's Biologics Effectiveness and Safety (BEST) System

Develops innovative methods to utilize electronic health records (EHR) effectively and establish automated adverse events reporting, utilizing natural language processing and AI

WHO Vigibase

Uses AI algorithms to analyze large volumes of safety reports to detect emerging patterns

EMA Eudravigilance

Uses AI algorithms to identify potential signals from data reported by patients and HCPs

Agencia Española de Medicamentos y Productos Sanitarios (AEMPS)

The Spanish RA has started using AI algorithms to aid swift detection of adverse events.

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When Artificial General Intelligence(AGI) becomes a reality...



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Thank You

Datafoundry is at **Booth #120** Narasimha.k@datafoundry.ai

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