



Towards AI-Enabled Training Needs Analysis Using Dual AI- Agent Collaboration – A Position Paper

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What is Training Needs Analysis (TNA)?

Training → Human Resource Development

Needs → Organisational or Industrial Context of why is training required

Analysis → Scientific and Data driven understanding

- 350 Billion USD global corporate training market in 2023
- TNA is
 - Tool for **organisational growth**
 - Deciding factor for **addressing “know-do” gap**, knowledge -> Action

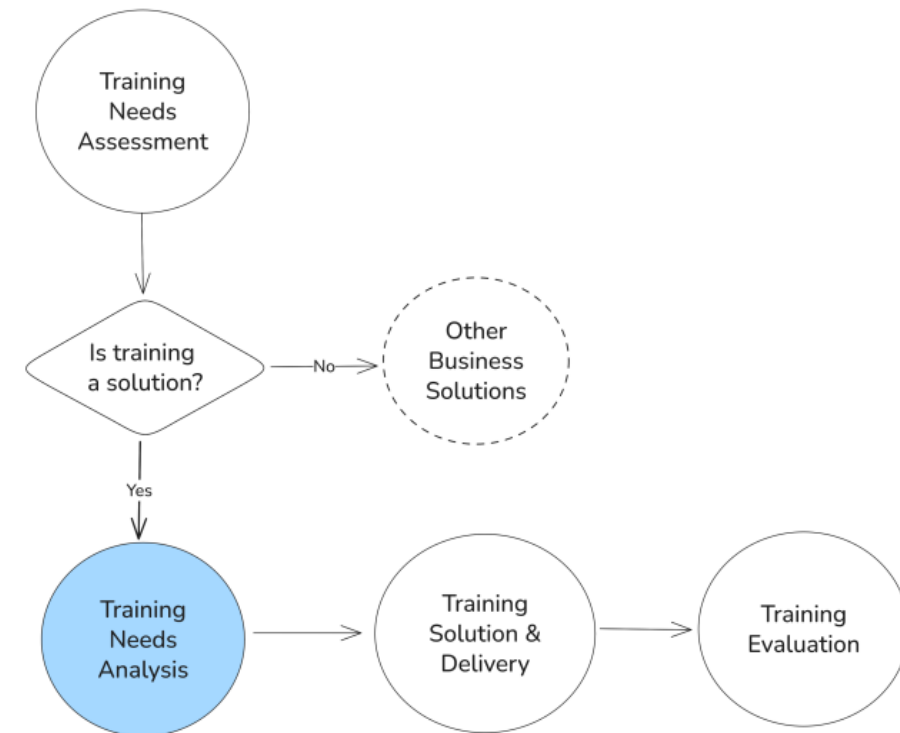
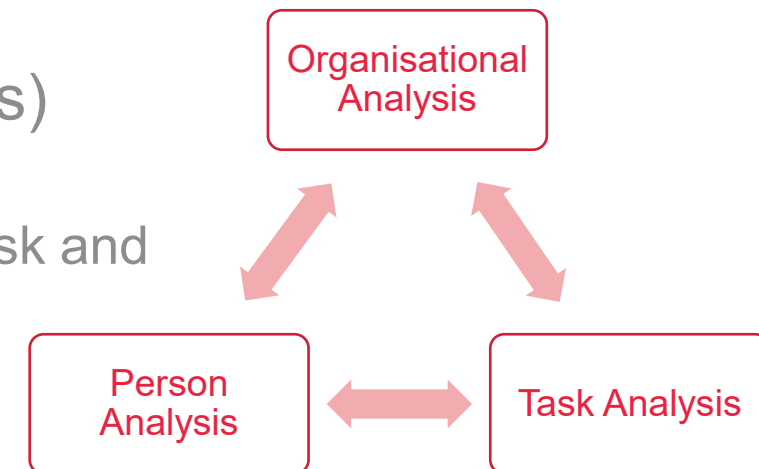


Figure 1: A high-level overview describing the position of TNA within the organisational training programme.

Training and TNA in Industry

- Training in organisations is mostly **reactive than proactive**, especially in SMEs
- More agile organisations = less resources
- To **automate and scale TNA process** using AI (mainly LLMs) for **industrial implementation**
- To **empower human professionals** with data-driven insights and scientific methods that **complement their tacit knowledge and practices – Human centric AI**
- OTP - Widely used TNA Model (and its variants)
 - McGehee, W., & Thayer, P. W. (1961)
 - Relational understanding between Organisation, Task and Person
 - Multistep analysis



Related Works

Very similar - Effective Needs Analysis in Knowledge Management using RAG (Su et al., 2025)

LLMs in Training and Learning Applications

- A **Virtual training coach provided near human sports coaching advice** with minimal historic data (Silacci et al., 2020).
- Multitasking capabilities of LLMs helped **improve learners' performance** (Baillifard et al., 2025), in **medical learning, sports training**, and other HR processes (Wei et al., 2021).
- Transforming wide range of HR processes with **personalised learning experiences** with IBM's Watson AI, Engazify, Google's Project Oxygen, Clustree, Visier, and platforms such as eDoer and Duolingo's Birdbrain (Benabou and Touhami, 2025).

LLMs applications in Professional Consultancy

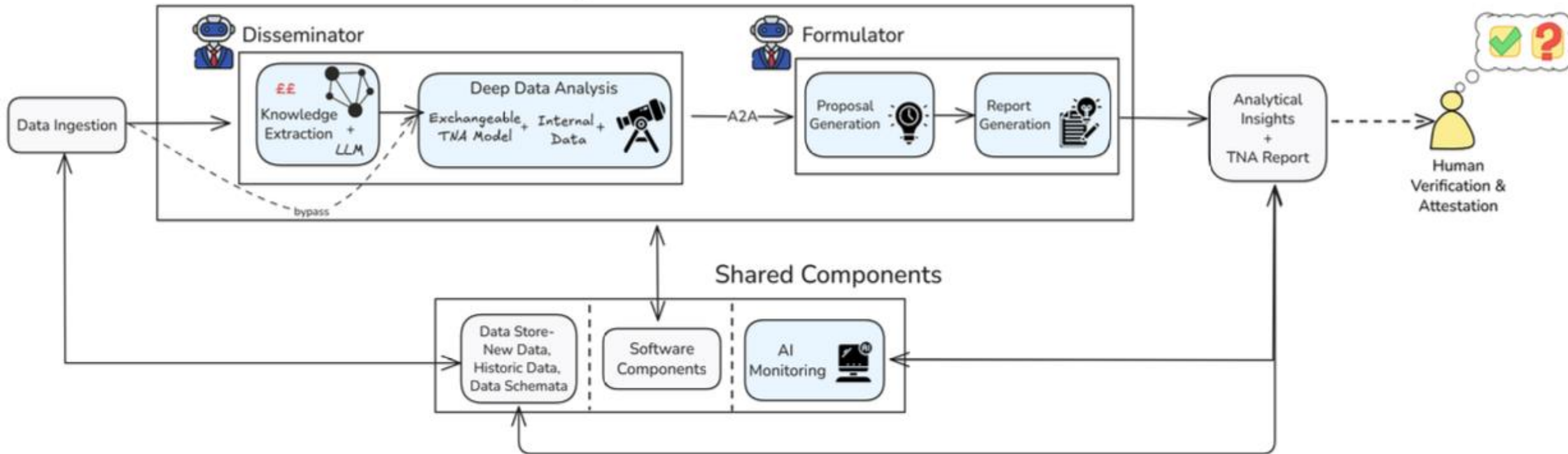
- The **financial consulting industry is rapidly adopting GenAI** integration with examples from IBM and others (Mohamad Ali, 2024).
- **Strategic advising in management consulting** by providing reliable decision-making support (Malberg et al., 2025).
- Encouraging **success in medical consulting**, although *challenges such as accurate interpretation of numerical data remain unsolved* (Dell'Acqua et al., 2023).
- Significantly **improved Q&A results on complex datasets such as US tax law** with RAG and few-shot prompting (Nayet al., 2024).

Challenges in Industry

- World Economic Forum, 2025 Demand trends: **85% of employers to prioritise workforce upskilling**, including hiring for new skills (70%), internal role reallocation (50%), and, in some instances, workforce reduction due to skill obsolescence (40%). – will require significant resources for training
- Market survey- **Existing TNA products or platforms** are either deployed for in-house usage or lack cognitive and creative aids, are **inflexible**, and mostly **rely on a conventional** approach to conduct TNA.
- Key attributes for any HR system: Fairness, Explainability, Transparency, Data Privacy, and Human Inclusiveness – often ignored due to **technological and practical issues**
- Unique HR challenges, such as conflicts between human management and AI automation
- Major limitations of LLMs:
 - **Hallucinations, fabrication of facts** via knowledge overshadowing, and **ethical risks** with AI-enabled Knowledge Management (Rhem, 2021).
 - Challenges in **compliance, data processing, system evaluation** (Brehme et al., 2025)
 - Limited understanding of the **impact of context retrieval on answer correctness** (Joren et al., 2025) limit simple RAG implementations for knowledge intensive use-cases.

Proposed Design

Agentic Workflow



Disseminator: Knowledge Extraction (NER) | Deep Data Analysis (Reasoning & Tools)

Formulator: Proposal Generation (Idea Generation & Creative Planning) | Report Generation (Structured Output)

Disseminator

A2A →

Formulator

1. Knowledge Construction: Knowledge Graphs

Why knowledge graphs (KG)?

- Disseminates **unstructured contextual** data for **Global Search (Sensemaking)** over Local search (Edge et al., 2024) and for its **relational understanding**
- **Outperforms prompt based** methods for extraction and interpretation of unstructured scientific text (Dagdelen et al., 2024)
- Empirical knowledge **evidence**

Problems with graphs

- Resource intensive and high maintenance (Peng et al., 2024, Hofer et al., 2024)
- **Optional** for less knowledge intensive use cases

2. Deep Data Analysis: Deep Research

- Analysis of training needs based only on the constructed knowledge artifacts
- Multiturn analysis with **pluggable TNA models**
- Includes **tools**: Data Processing & Visualisation, Data transformation, Reasoning over numeric data etc

3. Proposal Generation

- Novel idea generation for training, **LLMs as idea-machine** with near-human creative capabilities of LLMs are leveraged (Girotra et al., 2023)
- The **Novelty – utility trade off** to keep in mind (Wang et al., 2025)

4. Report Generation

- **Structured output generations** (json, xml, office docs, MD) with dynamic data schema configuration
- Embedded with business assets such as policies, branding etc. and **reviewable by human experts** for business use
- Office software/downstream HR process friendly reports

AI Monitoring

- Enterprise applications **MUST** have **LLM observability**
- For real-world applications, LLMs often underperform, in large part due to **hallucinations** caused by **long-tailed data** and **the lack of robust publicly available evaluation benchmarks** (Patel et al., 2025; Shi et al., 2025) Leading to disparity in expectations between system developers and stakeholders (Calderon and Reichart, 2025).
- To avert the impact of **threats** such as prompt injections, jailbreaks, misinformation, invalid use, etc using **AI Guardrails** (Fu et al., 2024; Zhong et al., 2025).
- To track **inference costs, system logs, and other software engineering metrics** for effective technical maintenance (Pradeep Kumar Sambamurthy, 2024), capture business metrics, real-time and non-real-time feedback, organisational KPIs.(Ma et al., 2024; Gupta et al., 2024; Niimi, 2024).

Proposed System Evaluation

- Crucial for real-world AI implementation success, but existing benchmarks have **limitations for business specific uses** (Alonso and Church, 2024; Eriksson et al., 2025).
- Modular system design and AI agents are harder to evaluate due to their **context sensitivity, autonomy**, and inherent **stochastic nature of LLMs** (Moshkovich et al., 2025).
- Business-specific **datasets curated with the help of training professionals** to ensure alignment and compliance with business policies.
- Focus is on **qualitative measurements at the task level** (Shavit et al., 2023) , in addition to manual evaluations by experts and **quantitative metrics** such as task completion times, accuracy rates, resource utilisation, user satisfaction scores, ROI figures, and others (Prashant D. Sawant, 2025).

Future Work

- Implementation of the proposed system is ongoing
- Scan this QR to see sample outputs by the proposed design prototype:

[Demo website - https://fieri-leadership.github.io/AI-Enabled-TNA/](https://fieri-leadership.github.io/AI-Enabled-TNA/)

- Work to be considered:
 - Organisation compatible **data schemata**
 - Real-world **evaluation datasets** with the help of experts from **our industrial partner Fieri Leadership**.
 - Investigate - data and user privacy concerns, regional legal compliance, cost and compute optimisation, and improving product economics.
 - May also **explore alternative hybrid setups** with open-weighted **SLMs** with larger proprietary LLMs (Belcak et al., 2025).



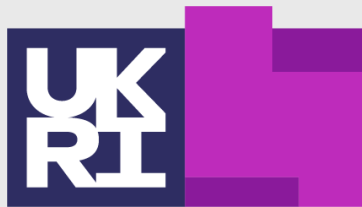


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