

# Large Knowledge Graphs for Construction, Manufacturing and Mining: A work in Progress

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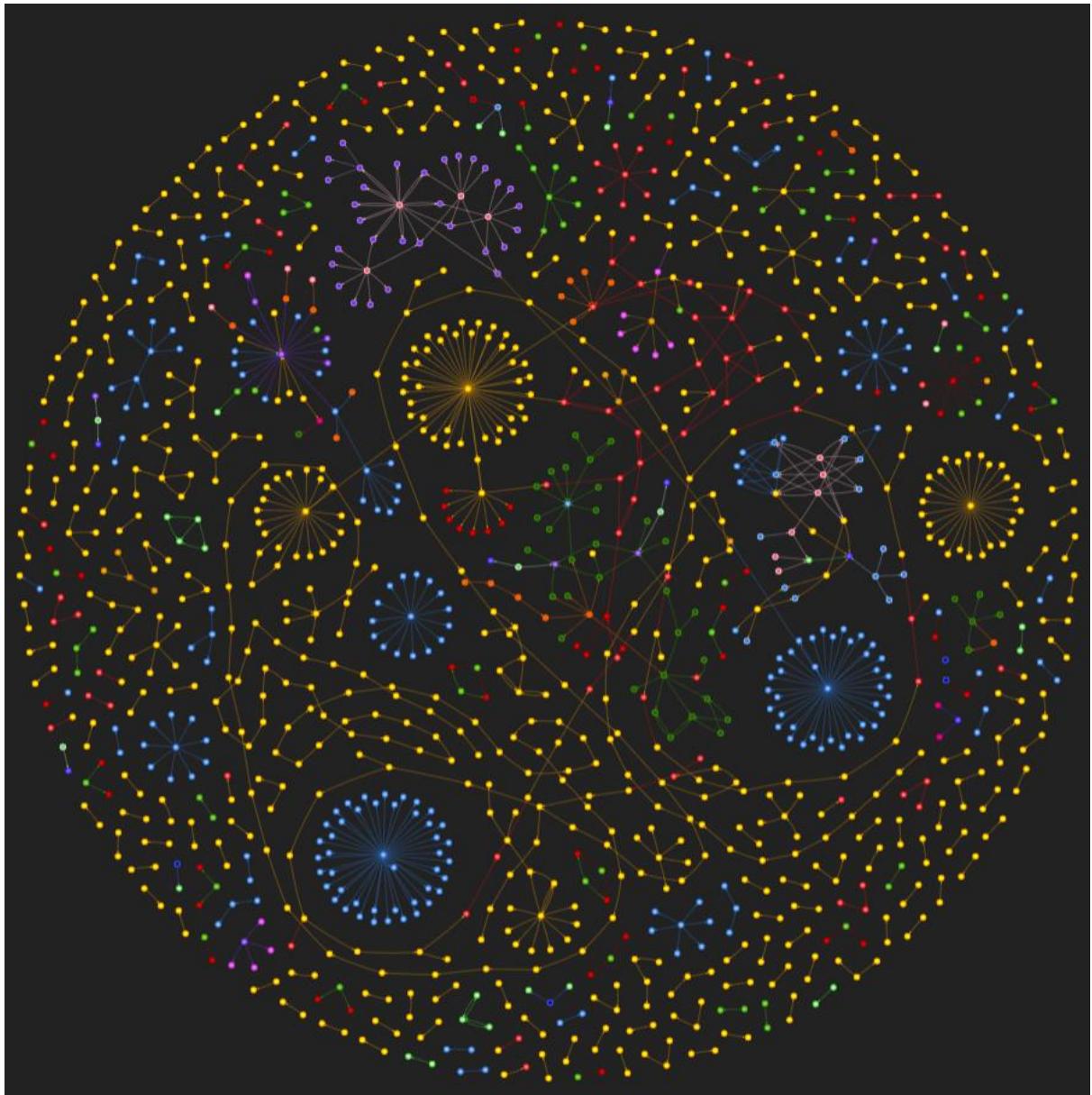
The manufacturing industry has seen great revamps in recent times through the introduction of IoT and AI integration. Autonomous dark factories are on the rise, not needing human intervention at all in many processes. From another end, breakthroughs in different AI products/frameworks/methodologies have made multiple tasks easier to automate. In this regard, Knowledge Graphs (KGs) have been playing an interesting role in different application areas. The ability to store information accurately as Tripartite Graphs [1] have proved to be extremely useful especially in domains like biology and medicine, but also affecting interconnected industrial processes like construction, manufacturing, mining, etc.

Some important literature highlighting the evolution of knowledge graphs include discussions on how LLMs can be used for KG generation [2], triple-set prediction by LLMs [3], capabilities of LLMs for constructing KGs [4], usage of KG Embeddings in LLMs [5], information retrieval and reasoning [6], task-oriented KG reasoning [7]; additionally multiple other resources highlight structured reasoning with KGs [8], AI Agent usage to traverse through KGs to find innovative subgraphs for hypothesis generation [9] and more. Interestingly, not only can KGs be used for novel hypothesis generation by letting a swarm of AI Agents loose on the KG from nearby nodes with slightly different prompts/goals to develop starkly different subgraphs, similar (backtracking/causal-analysis) approaches can be used in Root Cause Analysis (RCA); RCA could be useful not only in manufacturing or maintenance, but also in cybersecurity applications [10] [11] [12].

A recent paper in this regard focuses on utilizing KGs for smart manufacturing [13]. Apart from a whole lot of intriguing aspects which are discussed in the paper, the authors also focus on KG-reasoning mentioning that *“integrating structured ontologies with KGs enhances deductive reasoning, making implicit connections explicit and generating new insights to augment the existing knowledge base in” Smart Manufacturing*. They also discuss *“KG fusion at the conceptual layer ... merging disparate ontologies and schemas from different KG”*s. The article goes on to discuss applications of KGs in Engineering Design, Optimization and Scheduling, Quality Control, Supply Chain Management and Predictive Maintenance. In general KGs can be extremely useful for Risk Assessment, for usage alongside tools like (Design) Failure Modes and Effects Analysis, and allow novel frameworks to be developed on top of them for Safety Violation Assessment.

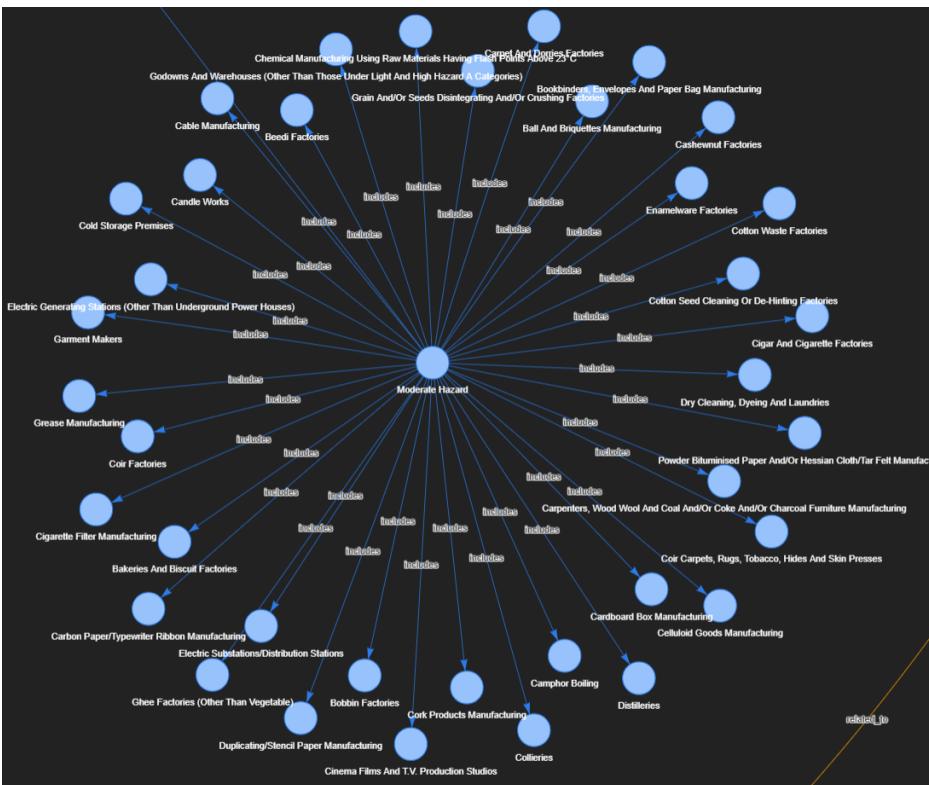
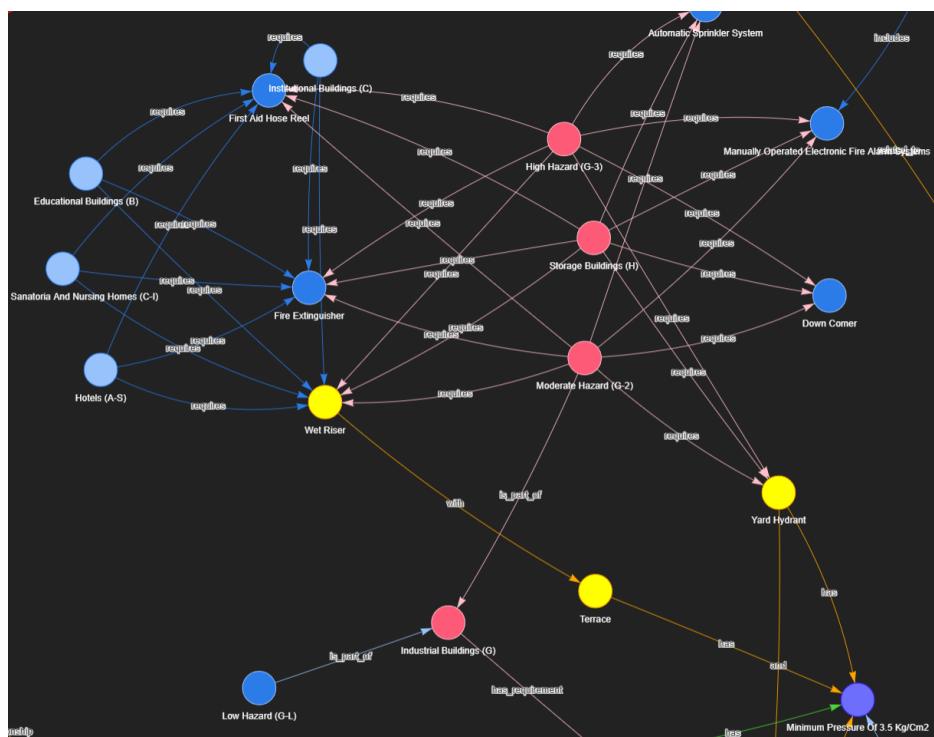
In SenseiAlgo, we understand the importance of being the first movers in using Large Knowledge Graphs (LKGs) for core industrial processes, especially to map and store complex indirect relationships in construction, mining and the manufacturing industries. In this regard we approach developing a LKG for Indian National Building Codes (INBC) 2016; the INBC

can be found in different volume-wise PDFs (more than 1000 pages each). We use the GPT-4o model using OpenAI APIs to develop small KGs for each chunk of the PDF text (we use *text\_splitter* from *langchain* to split the PDF files) and continue merging these chunks-wise KG; an intermediate LKG during this process is shown in Figure 1. It is observed that many clusters of good data connections gets formed, however additionally the LKG needs to be processed at regular intervals during its formation, to continually remove bad nodes and edges which somehow also clustered together among themselves. We will keep the community posted of our findings and look forward to novel methodology releases from academic labs focusing on KG post-processing.



*Figure 1: An evolving LKG (under-development) from INBC 2016 (Volumes 1 and 2) followed by two zoomed in views of different portions of this LKG*

These images are hyperlinked to view the entire graph (HTML) in any browser window (warning: it takes a long time to load, but is worth the patience)



## References

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