# Creating a Virtual Member of a Community of Practice

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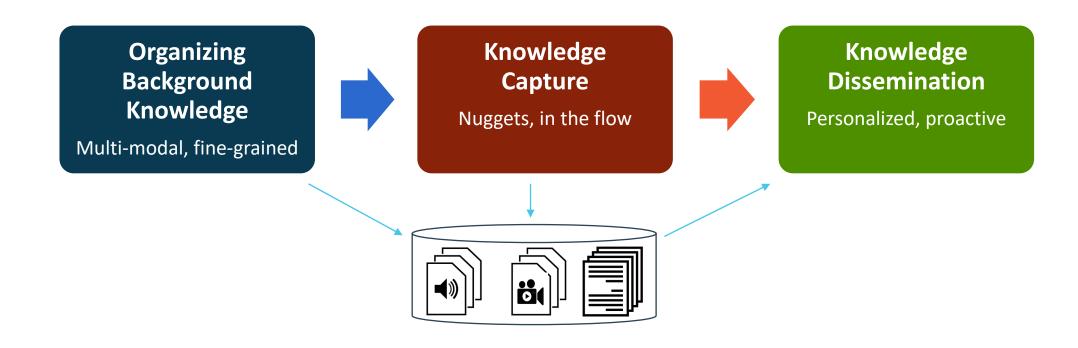


## i2k Connect



## DARPA KMASS: Knowledge Management at Scale and Speed

**Objective**: develop technology to enable effective use of documented knowledge, acquisition of new knowledge as part of regular workflows, and application of knowledge when and where it is required – *just in time*, *just enough*, *just for me*.





## DARPA KMASS – Knowledge Management at Scale and Speed

## Military replacements, rotations, and retirements occur frequently

- One third of US forces rotate in and out of theater every year
- For many deployments, rotations may be as frequent as every 1-12 months
- 30% of DoD personnel are eligible for retirement in 5 years



Face-to-face Knowledge Transfer

https://www.darpa.mil/news-events/2021-06-21

## Well Planning: Knowledge Management Challenges

- 1. Diversity of required knowledge: technical, corporate policies, regulatory requirements, training materials, journal articles, news articles
- 2. Real-world experience matters: large gaps in knowledge between new and expert engineers / geoscientists
- 3. Local context matters: an expert with experience in one part of the world needs local input to operate safely and efficiently elsewhere
- 4. High personnel turnover
- 5. Scale of the CoPs: 100s 1000s of members



https://www.drillingsoftware.com/

## ATHENA Concept



### ATHENA is a virtual member of your CoP

- It knows your industry, company, workflows, and tasks.
- It delivers multimodal info text, diagrams, videos, and people to call.
- By observing your CoP at work, it can push the right information, at the right time, to the right person.

## **Expected Impact**

- 1-2 orders of magnitude savings in time and effort
- Halving time to competency for new personnel

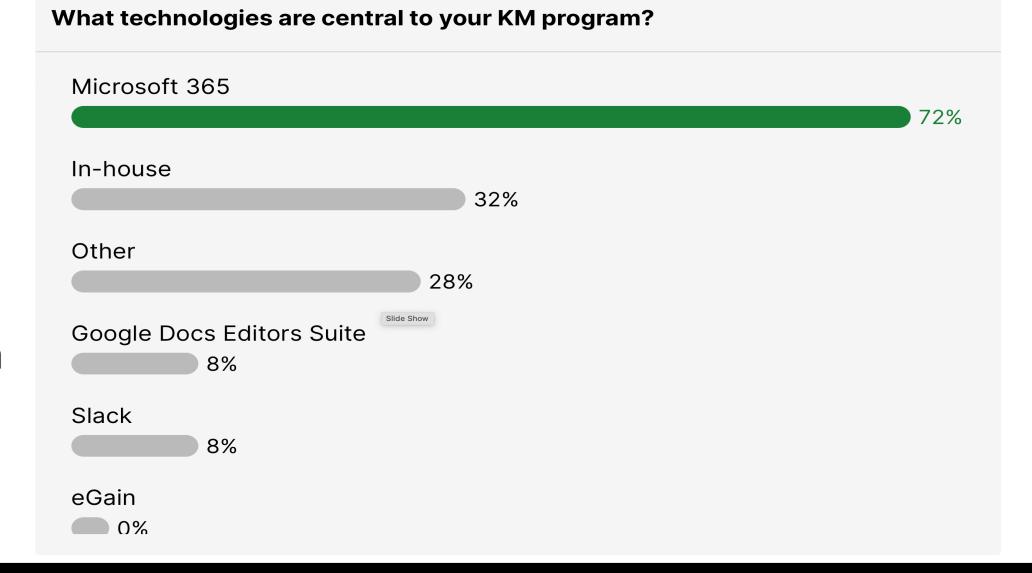
## **Technology Disruption**



Source: Large language models: What your data must include. Webz.io. (2023, March 22). https://webz.io/blog/machine-learning/large-language-models-what-your-data-must-include/



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ones? GPT family (ChatGPT, GPT3.5, GPT4) 67% CoPilot 40% Other 27% LLAMA 13% Claude 13% Mixtral

Are you using a Large Language Model (LLM) in your KM program? If so, which

## Background Knowledge

#### Goal

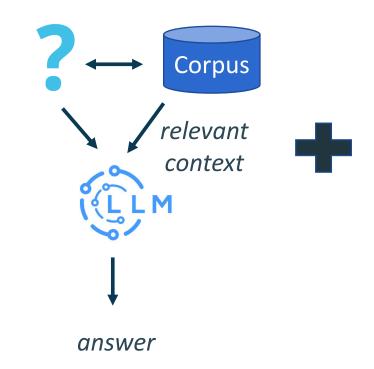
 Ingest and process documents to enable fine-grained retrieval and 'zero-click' answer delivery

#### ATHENA Approach

- Multimodal processing
- Hybrid Search: LLM + RAG + Agents

Current commercial / open-source RAG approaches are insufficient for QA

## Retrieval-Augmented Generation (RAG)

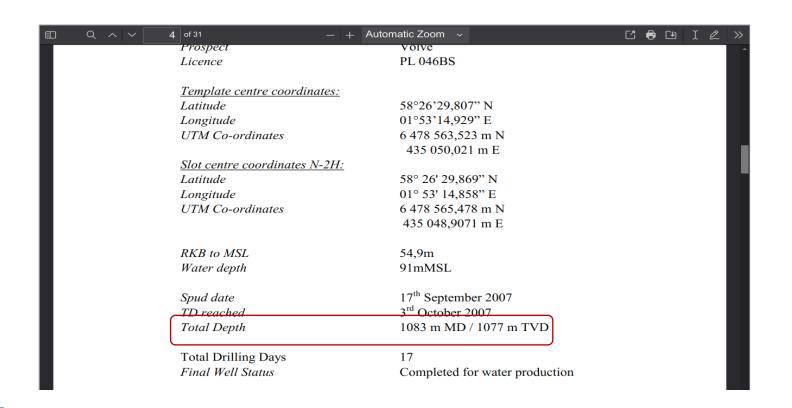


#### **Search Agents**

**Named Entity** Recognition Classifiers Geotagging Dates answer

## **Answers Rather than Documents**

What is the total depth of 15/9-F-7? The total depth of 15/9-F-7 is 1083 m MD / 1077 m TVDFinal Well Report for Well 15/9-F-7



#### Final Well Report For



Well: 15/9-F-7



Rig: Mærsk Inspirer

Geoservices		StatoilHydro	
Date Signed	:	Date Signed	:

Revision number: Date:

i2k Connect Pacific Science & Engineering 10

## LLMs on their own: Bing/Copilot

## Lacks domain understanding



You

what is the total depth of 15/9 f-7?



#### Copilot



The total depth of the expression 15/9 f-7 depends on the context in which it is used. Let's break it down:

- 1. 15/9: This fraction simplifies to 5/3.
- f: The variable "f" represents an unknown value or function. Without further context, we cannot determine its specific value.
- -7: Subtract 7 from the result of the previous steps.

To calculate the total depth, we need more information about the value of "f." If you have additional context or specific values for "f," please provide them, and I'll be happy to assist further!

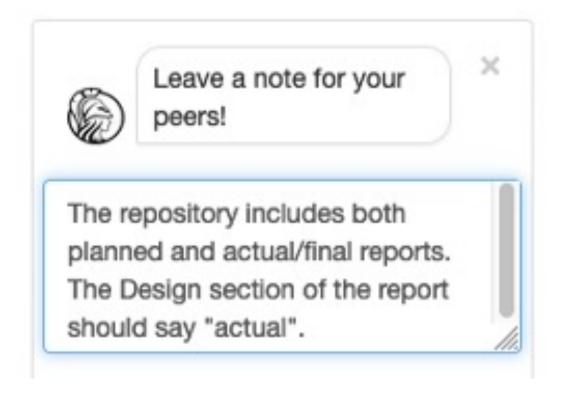
## Knowledge Capture

#### Goal

- Capture knowledge "nuggets" in-the-flow
  - i.e., local knowledge, lessons learned, tips/insights

#### ATHENA Approach

- Spoken/NL to enable capture with minimal disruption to current task
- Implicit capture of relevant context

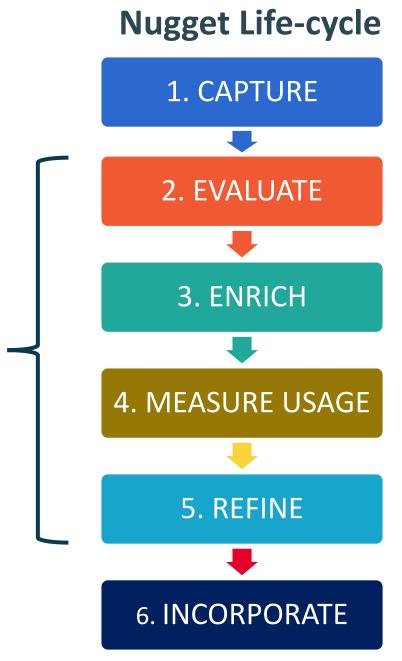


Nuggets captured "in the wild" have quality issues

## LLM-assisted Nugget Improvement

**Objective**: improve consistency, accessibility, and comprehensiveness of captured nuggets

**Approach:** use Large Language Models combined with research on knowledge elicitation to improve raw nuggets



## Audience Activity: Discussion (10 minutes)

1. How do you curate / verify captured knowledge?



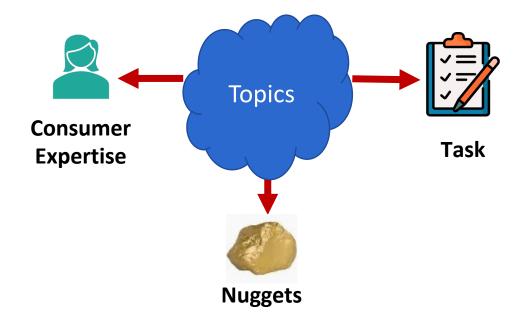
## Dissemination

#### Goal

• Deliver appropriate knowledge when and where it is required - *just in time*, *just enough*, *just for me*.

## ATHENA Approach

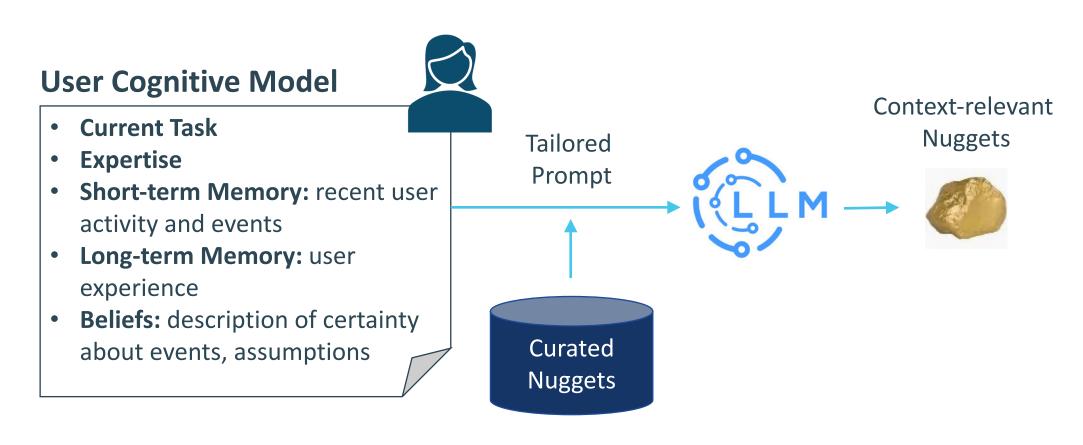
- When possible, use <u>Push</u> (volunteer assistance)
- Also enable <u>Pull</u> (assist by responding to questions)



Topic modeling connects Consumers with Nuggets for a given task

## User Modeling via LLMs

**Idea:** augment LLM-based retrieval with a "Cognitive Model" of a user to enable personalization and context relevance



Hi there! I'm ATHENA, your new Al assistant. I'm here to help with your tasks today. I'll send you a message if I know anything that could help you out.

Summary information about the well can typically be found in the Final Well Report, also called the End of Well Report.





Total Depth (abbreviated as TD) can typically be found in a table with other key well properties such as drilling rig, license, and operator.





Leave a note for your peers!



What is the total depth for 15/9-F-7?

Q

#### Answer

#### The total depth for 15/9-F-7 is 1083 m MD / 1077 m TVD

Final Well Report for Well 15/9-F-7

#### Source(s)



☐ Videos

#### Final Well Report for Well 15/9-F-7

/files/corpus/files/Volve/Well\_logs/14.DIV. REPORTS/15\_9-F-7/15-9-f-7\_nr\_002\_FWR\_MUD.pdf



page 2 - ...Statoil Well: ...7 Contents 1. INTRODUCTION...3 2. WELL & RIG DATA...

page 3 - ... Statoil Well: ... 7 1. Introduction 1.1. Purpose of the project ... 7 is desig...





Volve Nov-26-2008, 20:02:58 GMT



#### Recommendation to Drill: Volve Field, Well NO 15/9-F-7 and NO 15/9-F-9

/files/corpus/.../15\_9-F-7/14.DIV.REPORTS/15\_9\_F\_7\_F\_9\_Recommendation\_to\_Drill.doc.pdf

page 1 - ... Volve field Recommendation to drill NO 15/9-... 7 NO ... 9 Volve RA 00116 Classi ...

page 3 - ... Volve field Doc. no. Recommendation to drill Volve RA 00116 NO 15/9-...7 Valid from Rev. ...



Volve Apr-12-2007, 07:30:45 GMT



#### Final Well Report: Drilling and Completion of Well NO 15/9-F-7

/files/corpus/.../14.DIV. REPORTS/15\_9-F-7/Volve 15\_9\_F-7 Drilling and Completion EOWR.pdf

page 1 - ...FINAL WELL REPORT Drilling and Completion Licence no: PL046BS ...Well: NO ...

page 2 - ... Statoil Hydro FINAL WELL REPORT Drilling and Completion Licence no: PL046BS ... Well: NO .....

page 3 - ...FINAL WELL REPORT Doc no StatoilHydro Drilling and Completion Licence no: PL046BS ...Well: NO ....



Volve Sep-30-2022, 17:15:53 GMT





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## **Evaluation**



## **Findings**

- Strong preference for ATHENA over baselines
- Strong preference for LLM style interactions, answers
- Achieved high accuracy on granular retrieval (paragraphs 85%, video clips 82%)
- In-the-flow nugget capture substantially decreased capture time (62%)
- Expert nugget curation and delivery led to significant improvement in solution quality and task time
- Participants unable to succeed on any task with the Baseline LLM+RAG system

#### **Q1 2023**

- ~100 SPE volunteers
- Prescribed and volunteered questions
- Baseline: pre-LLM Q&A technology

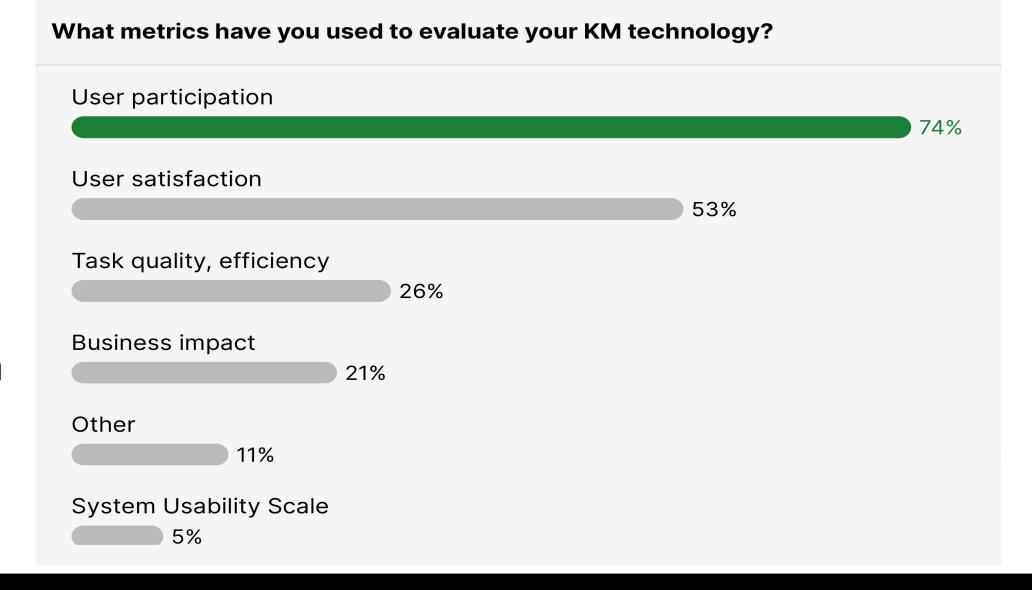
#### OQ2 2023

- ~80 SPE volunteers
- Prescribed tasks
- Expert nuggets, captured nuggets
- Baseline: SharePoint search

#### O Q1 2024

- 6 SRI/i2k staff
- Prescribed tasks
- Expert nuggets
- Baseline: Commercial LLM (Claude) + RAG

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## Take-aways on KM Technology

- Documents are no longer atomic: knowledge can be extracted from tables, sentences, video clips, etc.
- Capture of insights can be accomplished in-context with minimal distraction from task, but curation is needed to ensure utility
- Pushing personalized content based on task focus and expertise can yield significant performance improvement

LLMs have changed the landscape for Knowledge Management – and there's more to come!

## Considerations when Embracing LLMs

- Trust: LLMs are good at "reading" a document but you have to find it first
- Verification: necessary to check LLM results
- **Controllability**: even with instruction models, cannot guarantee that responses will respect the framing or local conventions; debugging is hard
- Latencies: may exceed user tolerance levels
- Cost: computational budget can explode
- Intellectual Property: data and documents may be retained or used for training