

# Research, RAG & Knowledge Readiness Checklist

Assess whether your organisation's knowledge infrastructure is ready to support grounded AI — including retrieval-augmented generation (RAG), AI-assisted research, and organisational knowledge systems.

IT &amp; Data Leaders

Knowledge Managers

## THREE CONCEPTS THIS CHECKLIST COVERS

### AI-Assisted Research

Using AI tools to discover, synthesise, and extract information from internal and external sources — faster and more consistently than manual research.

*e.g. Perplexity for market research, NotebookLM for document synthesis, ChatGPT for competitive analysis*

### RAG (Retrieval-Augmented Generation)

A technique where AI retrieves relevant content from a defined knowledge source before generating its response — so answers are grounded in your actual documents, not general training data.

*e.g. AI that answers questions using your company policies, contracts, or product documentation*

### Organisational Knowledge Systems

The infrastructure for capturing, organising, maintaining, and accessing institutional knowledge — including SOPs, policies, training materials, and operational guides.

*e.g. SharePoint, Notion, Confluence — or the AKE (Agentic Knowledge Engine)*

## 1 Document & Knowledge Infrastructure

Score: / 5

- Core internal documents exist in a digital, findable format** CRITICAL  
SOPs, policies, procedures, and guides are written down — not just in people's heads — and stored somewhere accessible
- Documents are stored in a central location that staff can access** CRITICAL  
Not spread across personal drives, email threads, or scattered SharePoint folders with no clear hierarchy
- Documents are version-controlled — old versions are clearly marked or archived** IMPORTANT  
AI retrieving outdated policy versions is a governance and accuracy risk — version clarity is non-negotiable for RAG
- Someone owns document maintenance — updates are made when processes change**  
Knowledge that is not maintained becomes a liability when AI retrieves it as current information
- Documents are written clearly enough for an AI to extract accurate answers**  
Vague, jargon-heavy, or inconsistently structured documents produce poor RAG results regardless of system quality

## 2 Data Quality for RAG

Score: / 5

- Key terms and definitions are consistent across all documents** CRITICAL  
"Revenue", "customer", "region", "approved" must mean the same thing everywhere a RAG system will search
- Documents are regularly reviewed for accuracy — not just created once** IMPORTANT  
A quarterly review cycle minimum — AI retrieving outdated regulatory or product information causes real harm
- Sensitive documents are clearly labelled with access classifications**  
Not all knowledge should be equally accessible. Clear classification prevents AI from surfacing confidential content inappropriately

**Structured data (tables, databases) has documented field definitions**

When AI accesses structured data, ambiguous column names or undocumented calculations produce unreliable answers

**You can identify which knowledge sources are "trusted" vs "informal"**

Official policy documents vs Slack conversations vs personal notes – RAG systems must be fed from trusted sources only

### 3 AI-Assisted Research Readiness

Score: / 5

- Staff know the difference between general AI answers and grounded AI answers** CRITICAL  
ChatGPT answering from training data vs a RAG system answering from your documents are fundamentally different in reliability
- Research tasks have a defined verification step before conclusions are used** IMPORTANT  
AI research output — from any tool — must be reviewed and fact-checked before being included in reports or decisions
- Staff can evaluate the reliability of AI research sources**  
They understand that Perplexity citing a low-quality blog is different from NotebookLM citing your own internal report
- Research prompts are structured with clear scope and source constraints**  
e.g. "Summarise only from the uploaded documents" vs open-ended research — staff know which to use for which task
- Research outputs are saved and attributed — not lost after the session ends**  
AI research sessions have no memory by default. Important findings should be documented, not relied upon in the same chat thread days later

### 4 RAG System Readiness

Score: / 5

- You can identify the 5-10 knowledge sources the AI must access to be useful** CRITICAL  
Core policies, product guides, pricing documents, FAQs — you know specifically what the AI needs to retrieve to answer real questions
- Access controls are in place — not everyone should see everything** CRITICAL  
A RAG system must respect existing organisational permissions. HR documents should not be surfaced to all staff via an AI assistant
- A process exists for updating the knowledge base when content changes** IMPORTANT  
When a policy changes, who updates the RAG knowledge source? Without this process, the AI answers from outdated information
- You can test and validate that the AI is retrieving from the right sources**  
You have (or will establish) a way to audit what the AI actually retrieved when it generated a response — traceability is a governance requirement
- A named person or team is responsible for the knowledge pipeline**  
Without ownership, knowledge bases degrade. The RAG system is only as reliable as the content fed into it

#### SCORE SUMMARY



**Score 0 – 8**  
**Not RAG-Ready**  
Focus on document consolidation, version control, and knowledge ownership before implementing any RAG system. A knowledge consolidation sprint is the right first step.

**Score 9 – 14**  
**Partially Ready**  
You can begin a limited RAG pilot on a specific, well-documented knowledge domain. Identify your most reliable document set and start there. Build governance alongside the pilot.

**Score 15 – 20**  
**RAG-Ready**  
Strong foundations. You are ready to design and deploy a governed RAG system or evaluate the Agentic Knowledge Engine (AKE) for your organisation.

SCORED 15+ AND READY TO BUILD?

## Agentic Knowledge Engine (AKE) — Limited Beta

A governed RAG-based intelligence layer that connects LLMs to your organisation's proprietary documents, policies, and structured data — with access controls, source traceability, and human oversight built in.

[denagentical.com](https://denagentical.com)  
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