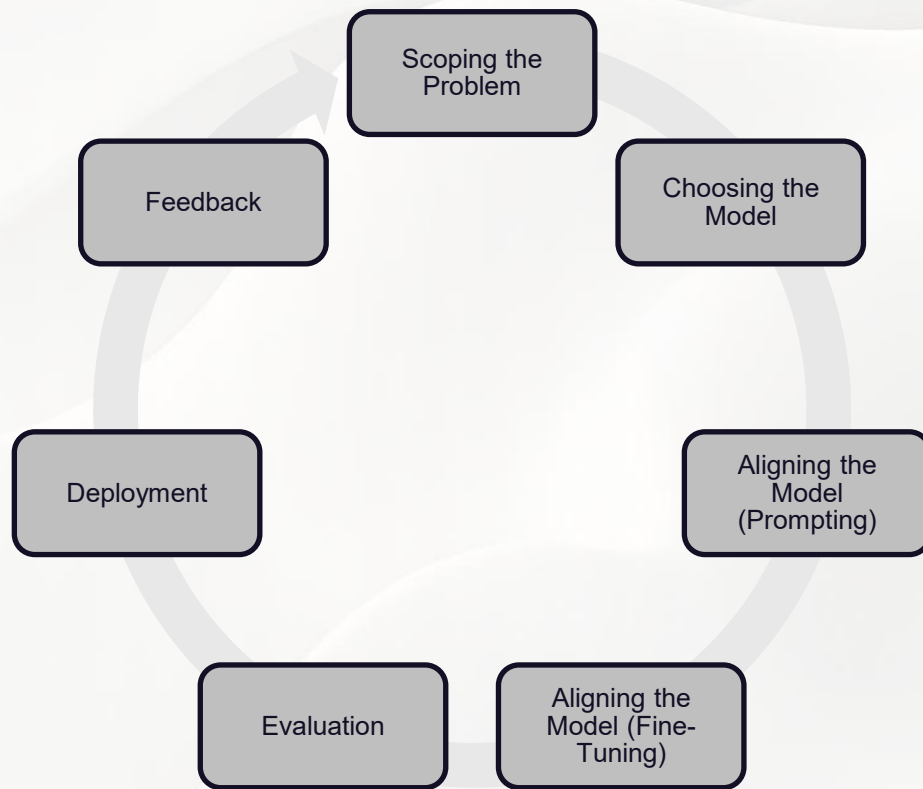


AI Project Lifecycle



AI Project Lifecycle

Scoping



What problem are we solving?

Be specific

Why does this problem matter?

Business impact

How will we measure success?

Quantifiable metrics

What are our constraints?

Time, budget, data, regulations

Is AI the right solution?

or is it a process problem?

Bad Scoping

Use AI to improve sales

No clear metrics

Undefined timeline

No consideration of available data

Good Scoping

Predict which leads are most likely to convert within 30 days, achieving 75% accuracy, to help sales prioritize outreach

Clear metric: 75% accuracy

Defined timeline: 30 days

Implies data needed: lead characteristics and conversion outcomes

AI Project Lifecycle

Scoping Model



**1. Pre-trained
Generative
AI (ChatGPT,
Claude,
Gemini)**

Pros: Ready to use, no training needed, versatile

Best for: Content generation, analysis, general tasks

**2. Custom
Machine
Learning**

Pros: Optimized for specific task, full control

Best for: Unique requirements, specialized predictions

**3. Fine-tuned
Models**

Pros: Combines general knowledge with specific expertise

Best for: Domain-specific content, consistent style/format

AI Project Lifecycle

Scoping
Model

Alignment (PE)



Alignment
through
Prompt
Engineering

Using PICF framework

Iterative refinement through conversation

No code or training required

Fastest path to results

When to use

General tasks, varied requirements, rapid iteration

AI Project Lifecycle

Scoping
Model

Alignment (PE)
Alignment (FT)



Alignment
through
Fine
Tuning

Taking a pre-trained model and training it further on your specific data

Teaches it your domain, style, or specialized knowledge

Requires: Hundreds to thousands of examples

Example

Teaching ChatGPT your company's writing style

Prompting

What it is: Giving the AI instructions and examples at the moment you ask the question.

Analogy: Giving a smart, general-purpose assistant (like a new intern) a very good set of instructions.

Cost: Cheap & Fast.

Rule: ALWAYS START HERE.

Fine-Tuning

What it is: A permanent update to the model's "brain" by re-training it on thousands of your private labeled data examples.

Analogy: Sending that intern to a 6-month specialized training course.

Cost: Expensive & Slow.

AI Project Lifecycle

Scoping
Model
Alignment (PE)
Alignment (FT)
Evaluation



Quantitative Metrics:

Accuracy, speed, cost per query

Qualitative Assessment:

Human review of outputs

A/B Testing:

Compare AI vs. current approach

User Feedback:

Are end-users satisfied?

Business Metrics:

Did it achieve the business goal?

Human-in-the-Loop (HITL)



Definition:

A process where a human must review, approve, or correct the AI's output before it becomes final.

Solution to Ethics

To fix Bias: An AI can suggest which resumes to review, but a human makes the hiring decision.

To fix Hallucinations: An AI can draft a report, but a human must fact-check and approve it.

NOT “optional”

It is the only way to use AI responsibly.

AI Project Lifecycle

Scoping
Model

Alignment (PE)

Alignment (FT)

Evaluation

Deployment



Integration:	How does AI fit into existing workflows?
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Scale:	Can it handle the expected volume?
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Monitoring:	Are we tracking performance?
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Rollout Strategy:	Pilot → Limited Release → Full Deployment
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User Training:	Do people know how to use it?
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AI Project Lifecycle

Scoping
Model
Alignment (PE)
Alignment (FT)
Evaluation
Deployment
Feedback



User Feedback:	What's working? What isn't?
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Performance Monitoring:	Track metrics over time
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Edge Case Discovery:	Find scenarios where AI struggles
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Model Updates:	Refine prompts, retrain, or fine-tune
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Loop Back:	New insights lead to new scoping
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Common Lifecycle Pitfalls



**Skipping
scoping**

"Let's just try AI and see what happens"

**Over-
engineering
early**

Fine-tuning before testing with prompts

**Ignoring
evaluation**

Deploying without testing

**No feedback
mechanism**

AI stagnates

**Unrealistic
timelines**

"We need this in 2 weeks"
