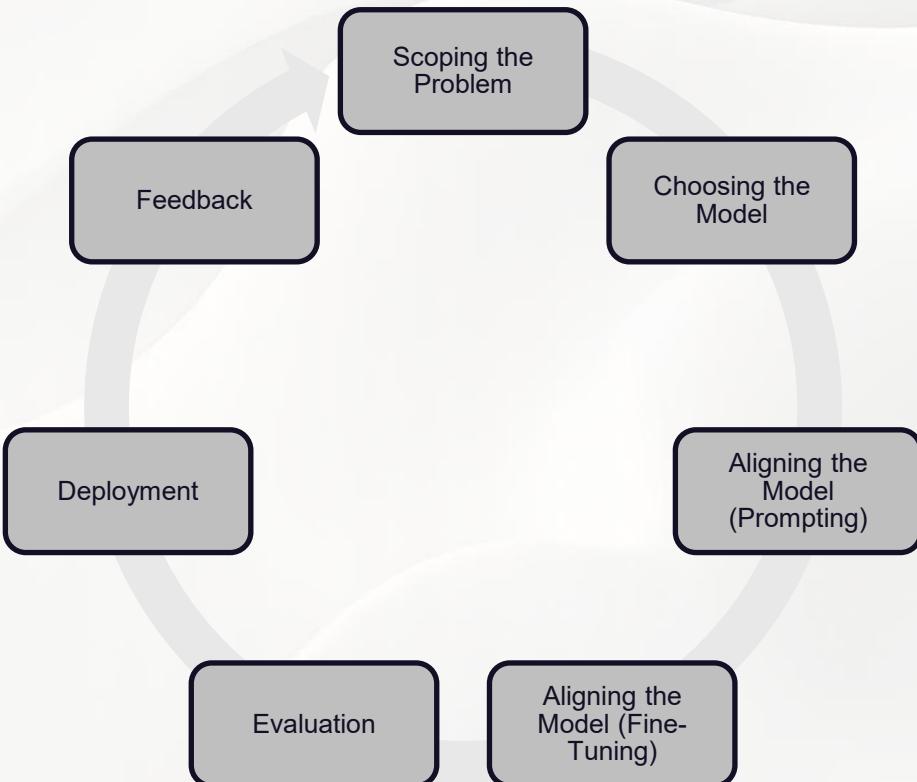


# AI Project Lifecycle





# 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	<b>Using PICF framework</b>
	<b>Iterative refinement through conversation</b>
	<b>No code or training required</b>
	<b>Fastest path to results</b>
When to use	<b>General tasks, varied requirements, rapid iteration</b>



# AI Project Lifecycle

Scoping  
Model  
**Alignment (PE)**  
**Alignment (FT)**

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## Alignment through Fine Tuning

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

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**Teaches it your domain, style, or specialized knowledge**

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**Requires: Hundreds to thousands of examples**

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## Example

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**Teaching ChatGPT your company's writing style**

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## 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

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**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

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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"**