

Overview of in-context learning

Mehmet F. Demirel

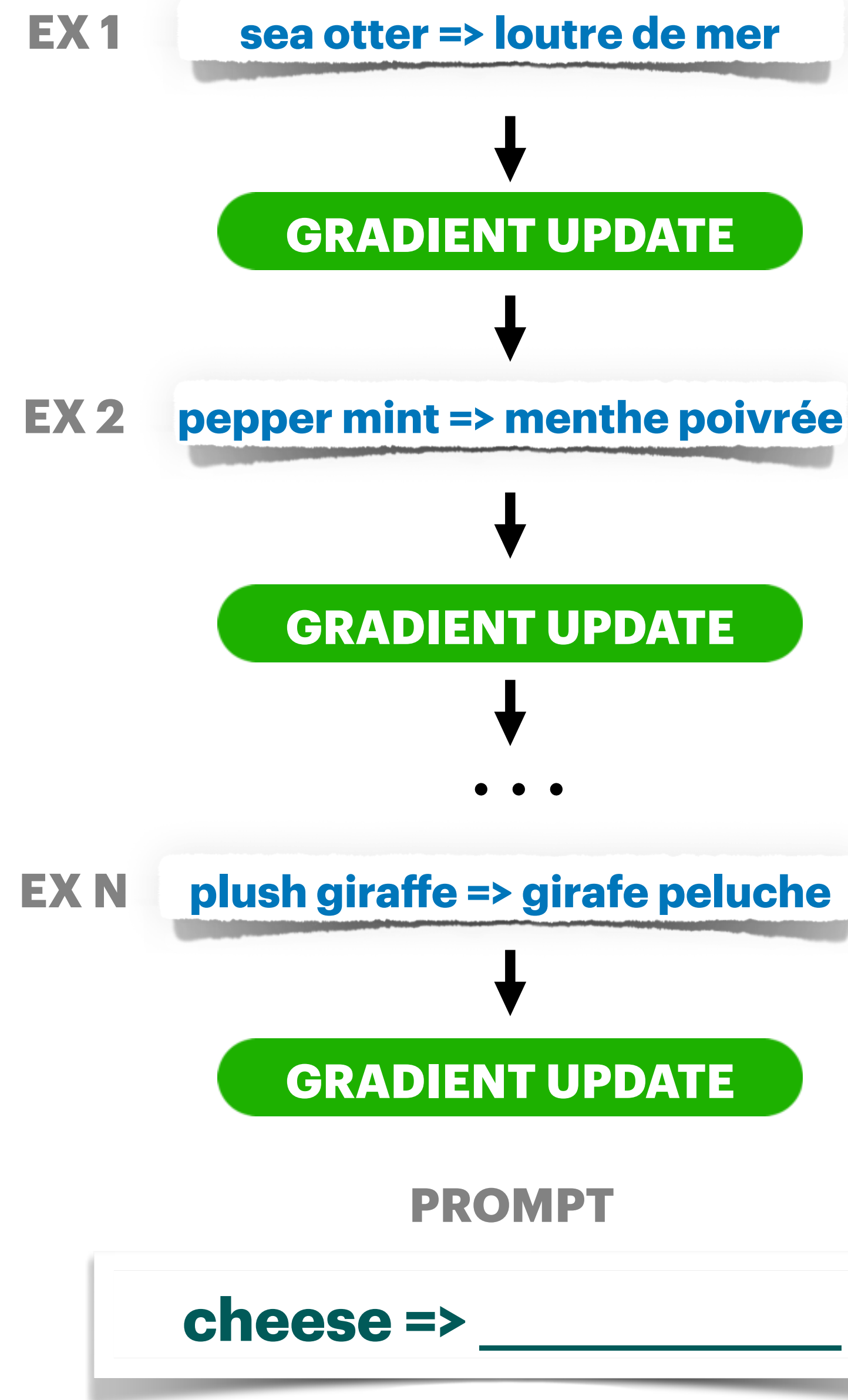
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The common part: unsupervised learning

- Pretrain a language model on a large corpus of linguistic data
- predict missing word —> **Sam took the ____ for a walk.**
- predict next word —> **Sam took the dog for a ____.**

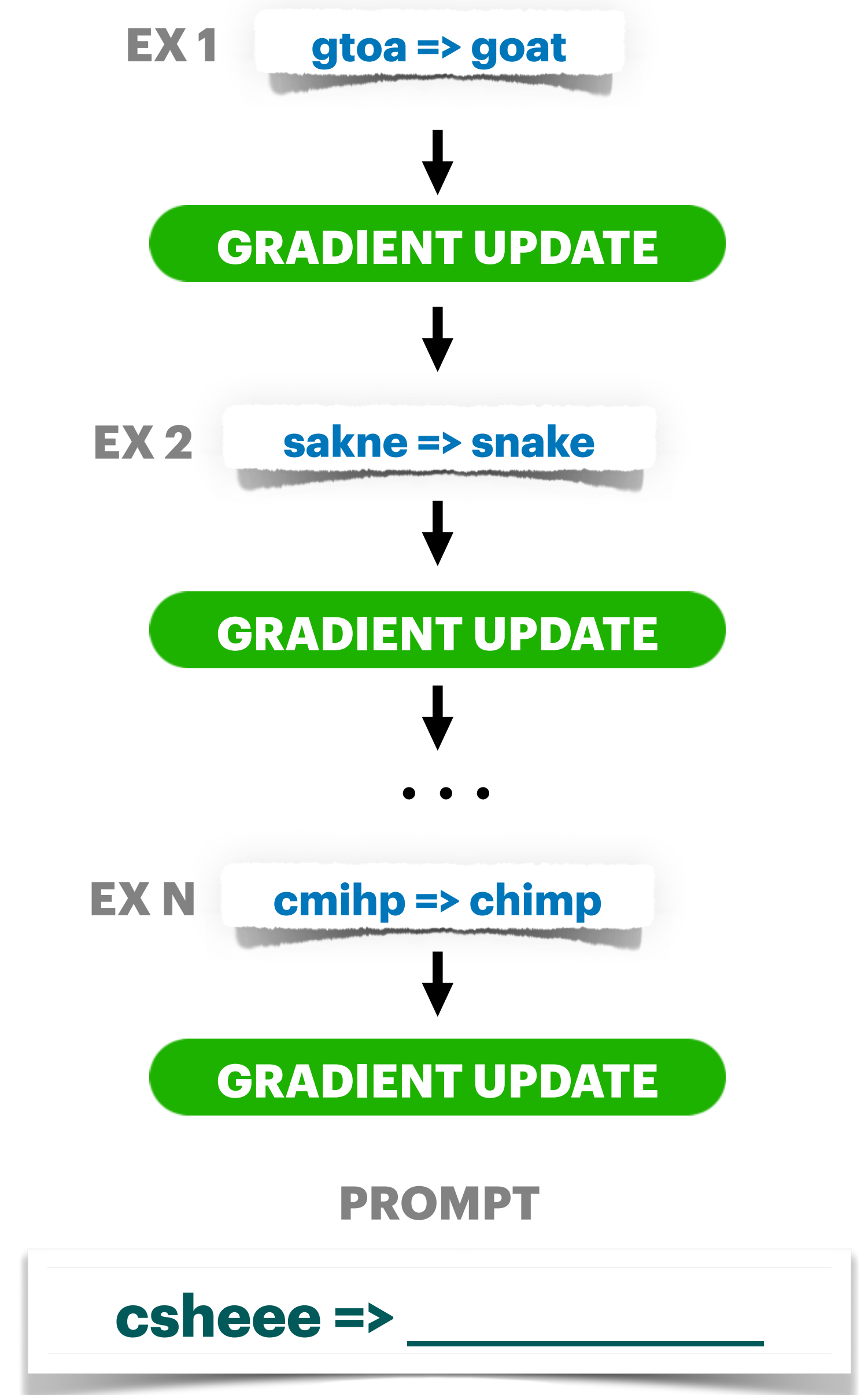
The traditional way: Fine-tuning

- Fine-tune the parameters of the pre-trained model for a specific downstream task using a large (thousands to hundreds of thousands) corpus of **labeled data**.
- Keep training the model via repeated gradient updates.
- **Strong performance on many benchmarks.**
- **Need a new large dataset for each task.**
- **Potential for poor out-of-distribution generalization**
- **Potential to explore spurious features of the data**



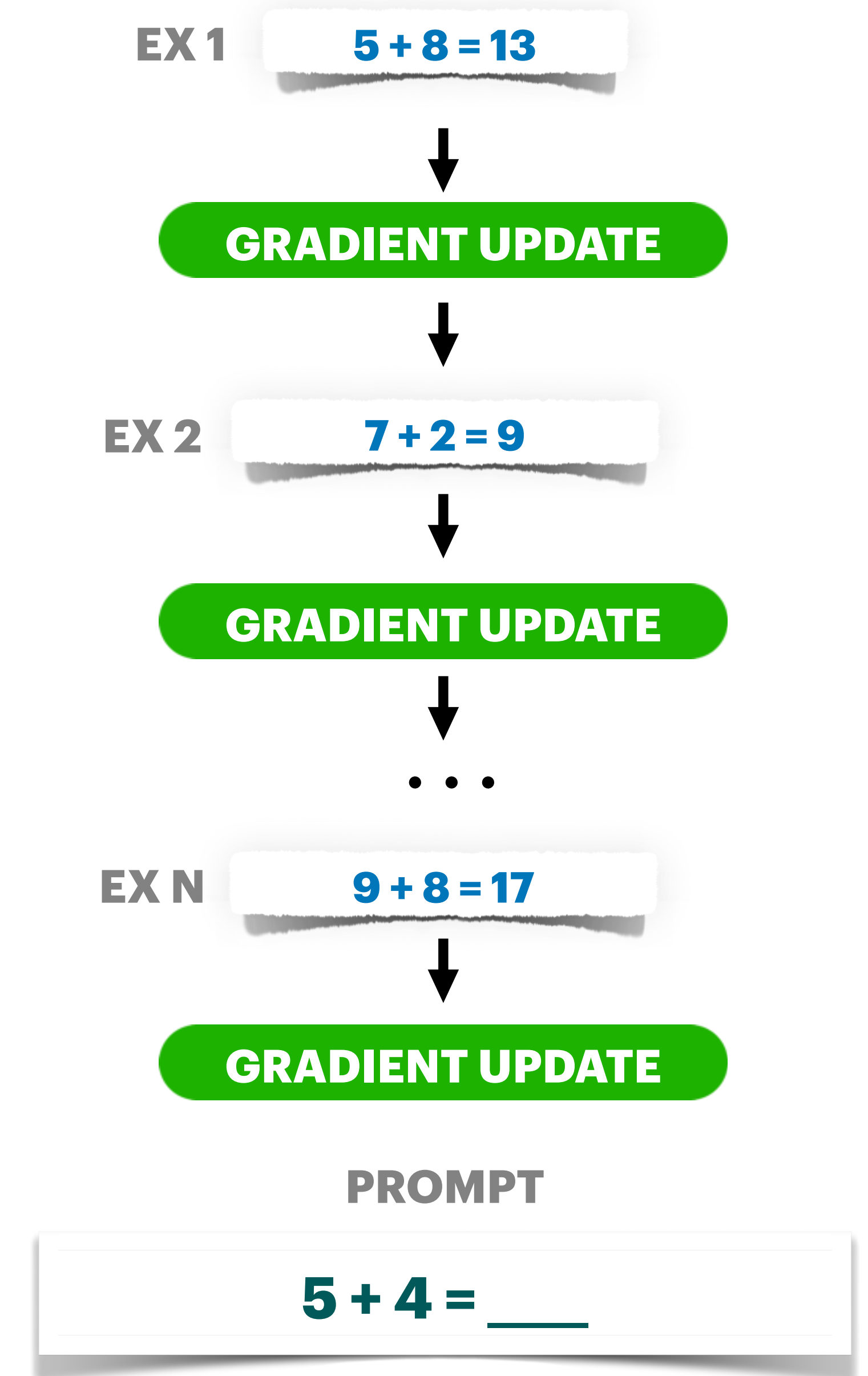
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In-context learning

- No training or optimization of the model parameters in the “adaptation step”.
- Simply give the model **a task description** as well as **none/one/few examples** as the input at inference time.
 - Only the task description: **ZERO-SHOT**
 - TD + one examples : **ONE-SHOT**
 - TD + a few examples: **FEW-SHOT**
- No gradient updates are performed.

FEW-SHOT

Translate English to French

sea otter => loutre de mer

peppermint => menthe poivrée

plush giraffe => girafe peluche

cheese => _____

ONE-SHOT

Translate English to French

sea otter => loutre de mer

cheese => _____

ZERO-SHOT

Translate English to French

cheese => _____

PROVIDE THE EXAMPLE(S)

IN THE CONTEXT OF

THE LANGUAGE MODEL

FEW-SHOT

Perform mathematical addition

$$5 + 8 = 13$$

$$7 + 2 = 9$$

$$9 + 8 = 17$$

$$5 + 4 = \underline{\quad}$$

ONE-SHOT

Perform mathematical addition

$$5 + 8 = 13$$

$$5 + 4 = \underline{\quad}$$

ZERO-SHOT

Perform mathematical addition

$$5 + 4 = \underline{\quad}$$

PROVIDE THE EXAMPLE(S)

IN THE CONTEXT OF

THE LANGUAGE MODEL

Few-shot

- **Give K examples of context and completion, and one final context whose prompt we want the model to predict.**
- **Major reduction in the need for task-specific data.**
- **Reduced potential to learn an overly narrow distribution from a large but narrow fine-tuning dataset.**
- **Still not as good as the fine-tuning SOTA, but competitive (GPT-3).**
- **Still need a few task-specific data.**

One-shot

- **Similar to few-shot, but with only one example**
- **Most closely matches the way in which some tasks are communicated to humans.**

Zero-shot

- **Provides maximum convenience (no task-specific example needed)**
- **Potential for robustness**
- **Potential for avoidance of spurious correlations**
- **Most challenging**
- **Even for humans, it is often hard to understand a task without an example.**

Foundation models

- The survey indicates that there is an emergence of **functionalities (such as in-context learning)** in foundation models.
- Rather than task-specific data and carefully-engineered features, NLP foundation models (such as GPT-2 and GPT-3) can make inference for given tasks whose task-specific examples are provided **in the context of the language model** as an input at inference time with no parameter optimization required.