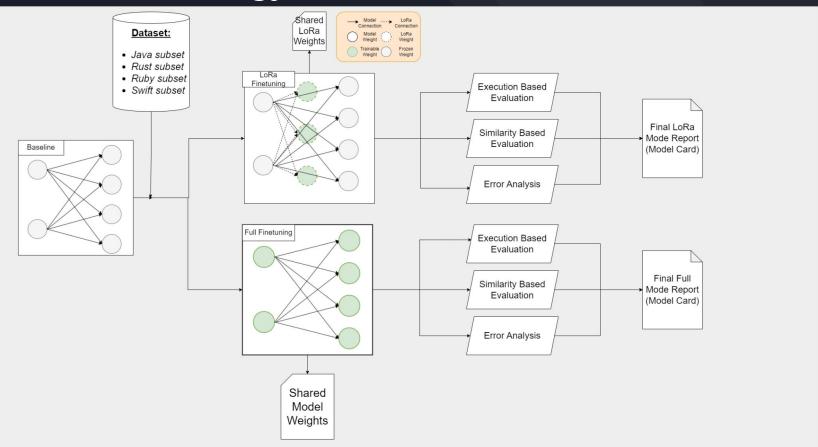


### Objectives

- Sharing models weights is not enough to enable equitable access to Code LLMs because of the Resource Limitations and the Knowledge Gap.
- This project aims to:
  - a. Provide a set of small code language models that can perform code completion in multiple programming languages (resource limitations)
  - b. Provide a framework that practitioners can follow to fine-tune code LLMs to their own needs and according to their own constraints (knowledge gap)

## Methodology



#### Baseline & Dataset

<u>Dataset</u>: 1 Million files of target programming languages, cleaned and filtered.

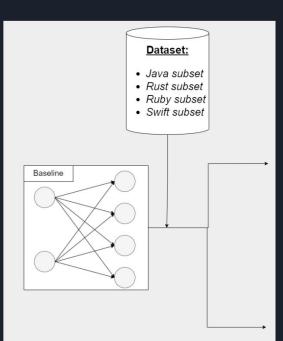
<u>Baseline</u>: CodeGen-350M-Mono Autoregressive Code LLM pretrained on Python.

<u>Task</u>: Casual Language Modelling

$$L_{CLM}^{(x)} = -\frac{1}{|x|} \sum_{i=1}^{|x|} log P(x_i/x_{< i})$$

whore

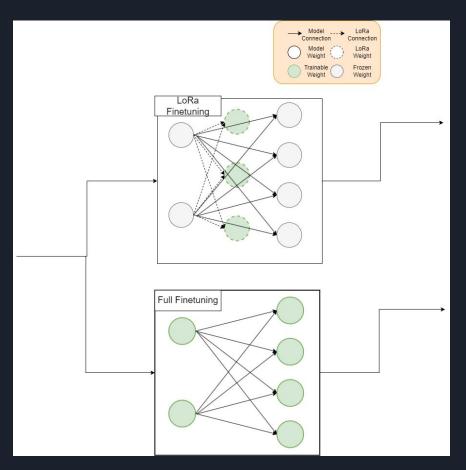
$$x = \{x_1, x_2, x_3, \dots, x_{|x|}\}$$
 represents a sequence 
$$x_{< i} = x_1, x_2, x_3, \dots x_{i-1}$$



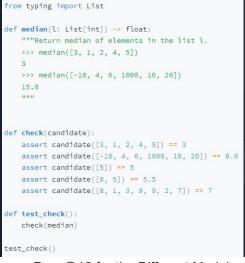
#### Finetune Methods

<u>Full</u>: All the parameters of the model are trainable.

Low Rank Adaptation (LoRa): Small set of Parameters (inserted at specific layers) are trainable



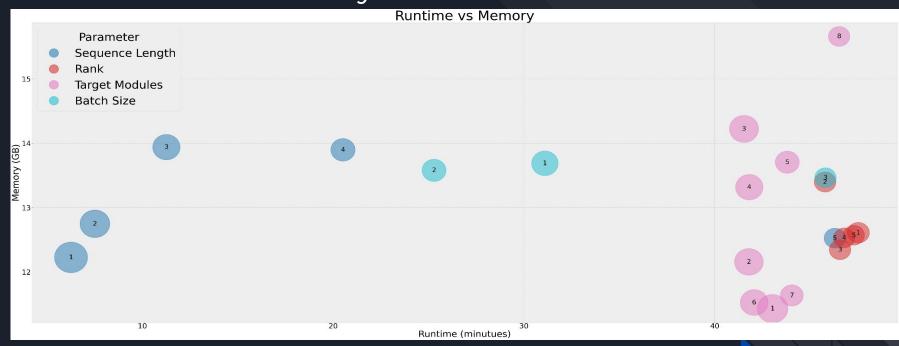






Model

# Trade-Off Analysis



Variable	Initial Value	Range
LoRa Rank	64	[8, 16, 32, 64, 128]
Batch Size	2	[1, 2, 4]
Sequence Length	1024	[128, 256, 512, 1024, 2048]
Learning Rate	5e-5	[5e-6, 1e-6, 5e-5, 5e-4]
Target Modules	Conf-6	[Attention Modules, Language Feature Modules, Fully Connected Module]

### Open-Source (Inference Demo)

