

## Vision-Language-Action (VLA) Models

A Review of Recent Progress

#### Xiangyu Li

Institute for AI Industry Research (AIR), Tsinghua University lixiangy22@mails.tsinghua.edu.cn

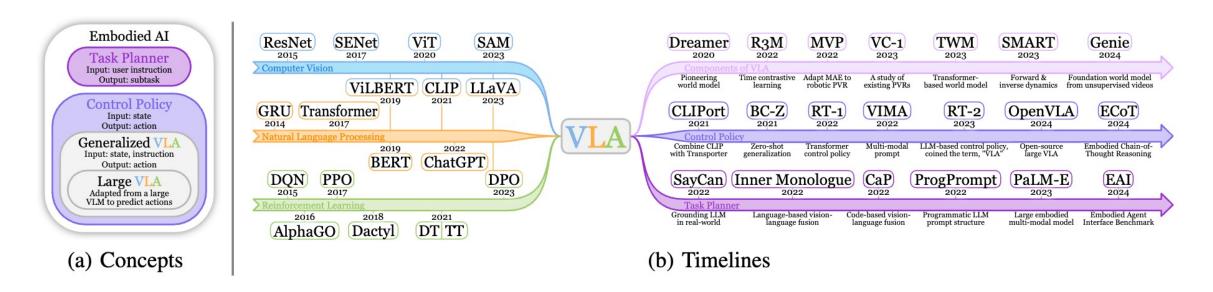


- > Background & Overview
- > Descrete VLA

>Continuous VLA

## What is a Vision-Language-Action (VLA) Model?

- A multi-modal foundation model for embodied Al.
  - Input modalities: vision (observation) & language (instruction)
  - Output modalities: action (low-level robot control policy)
- A VLA model utilizes a VLM for VL-conditioned action generation.

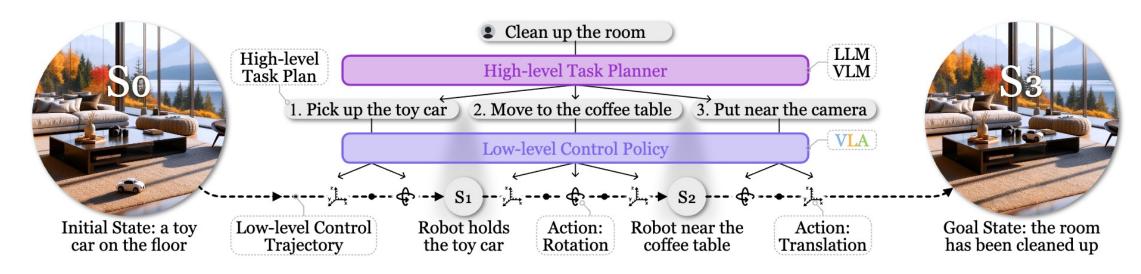


The concepts (a) and timelines (b) related to the development of VLA.

[1] Ma, Yueen, et al. "A Survey on Vision-Language-Action Models for Embodied AI." arXiv preprint arXiv:2405.14093 (2024).

## VLA + Task Planner for Long-Horizon Tasks

- VLA is initially optimized for low-level robot control policy.
- To complete complex, long-horizon tasks, an effective approach is to add a LLM/VLM-based task planner to decompose them to simple subtasks.
  - Earlier works usually adopt <u>a separate model</u> as the task planner.
  - Recent works utilize <u>a shared VLM</u> for both task planning and control policy (i.e., *dual-system*).



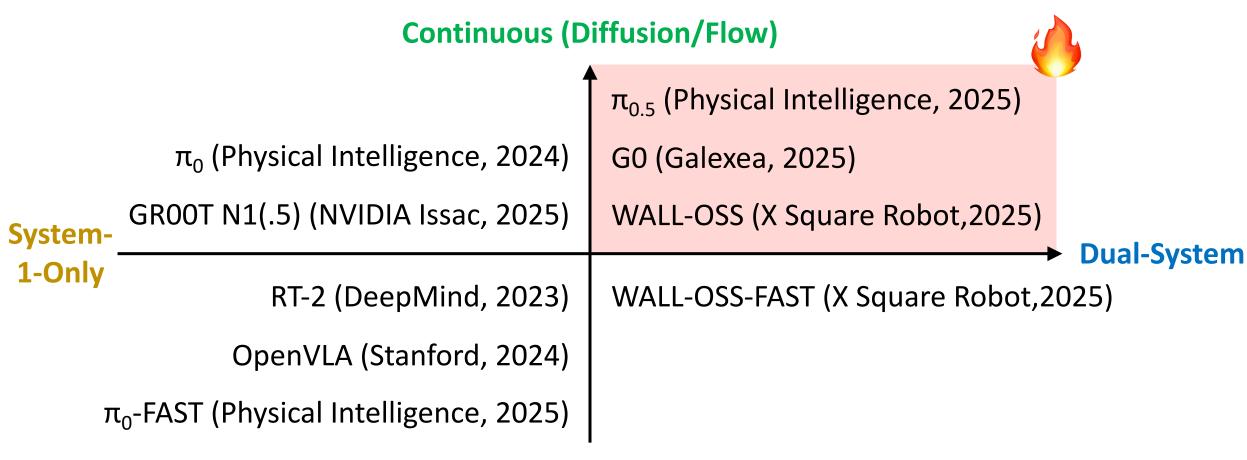
An example of a hierarchical robot policy (high-level planning + low-level control).

[1] Ma, Yueen, et al. "A Survey on Vision-Language-Action Models for Embodied AI." arXiv preprint arXiv:2405.14093 (2024).



## Recent Progress of VLA: an Overview

Trend: dual-system (planning + control) with continuous action generation.



**Discrete (Autoregressive)** 

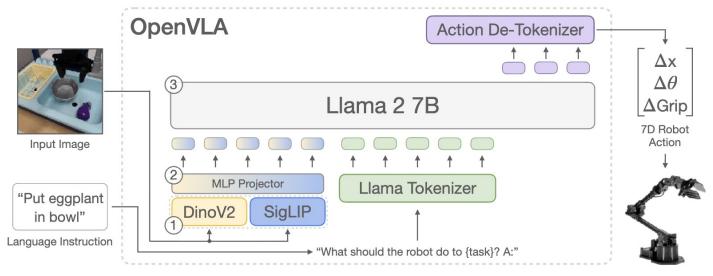
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## Discrete (Autoregressive) Action Generation

- Train the VLM to generate discrete tokens that directly map to actions.
  - Pros: <u>friendly for autoregressive VLM</u> to learn and generate.
  - Cons: <a href="high-latency">high-latency</a> and <a href="low FPS">low FPS</a>, not suitable for high frequency control.
- Representative methods:
  - RT-2 (ViT + PALI-X/PALM-E): a pioneer work that proposes and popularizes the term "VLA".
  - OpenVLA (DinoV2 & SigLIP + Llama2): an influential open-source VLA model (3.8k stars).
  - **FAST**: an action tokenizer that compresses action sequences with DCT.

Example: OpenVLA



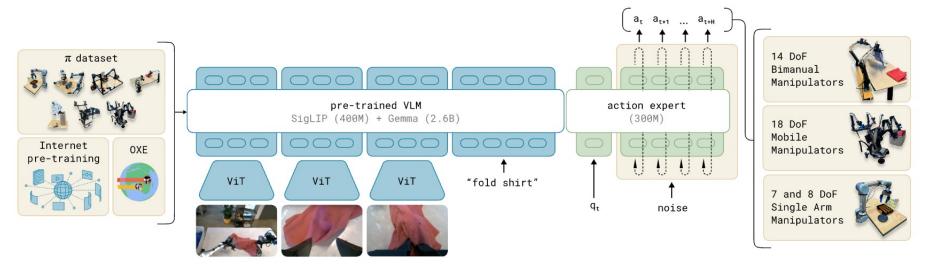
[1] Kim, Moo Jin, et al. "OpenVLA: An Open-Source Vision-Language-Action Model." arXiv preprint arXiv:2406.09246 (2024).

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#### $\pi_0$ | Physical Intelligence ( $\pi$ ) | Oct. 2024 (Opened in Feb. 2025)

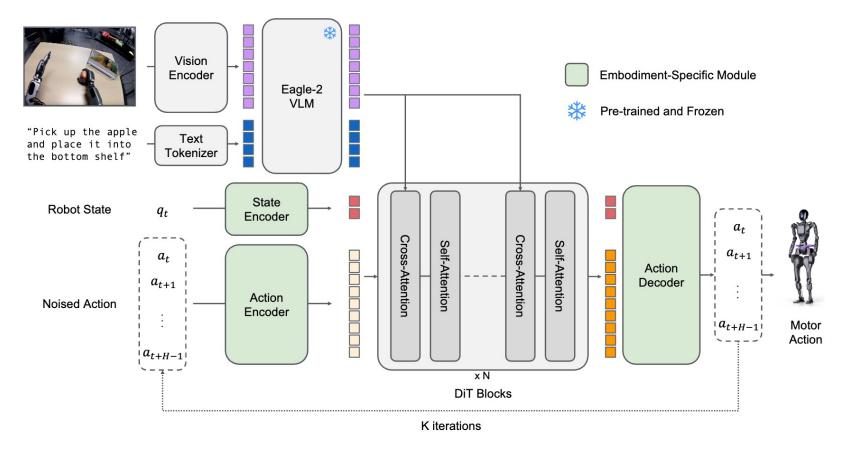
- The first to integrate a flow-matching action expert to a pre-trained VLM.
  - Pretrained VLM: Internet-scale semantic understanding.
  - Flow-matching action expert: high-frequency (up to 50Hz) control policy.
- Training recipe matters!
  - VLM pre-training: Internet-scale dataset.
  - VLA pre-training: Open X Embodiment dataset +  $\pi$  Cross-Embodiment Robot dataset.
  - Post-training (optional): high-quality post-training data (difficult/unseen tasks).



[1] Black, Kevin, et al. "\$\pi 0 \$: A Vision-Language-Action Flow Model for General Robot Control." arXiv preprint arXiv:2410.24164 (2024).

### GROOT N1(.5) | NVIDIA Isaac | Mar. 2025 (N1.5 in Jun. 2025)

- The first open foundation model for generalist humanoid robots.
  - Similar to  $\pi_0$  in model architecture and training method.
  - A difference: only passes hidden states from a certain VLM layer ( $\pi$  is each layer).



[1] Bjorck, Johan, et al. "GR00T N1: An Open Foundation Model For Generalist Humanoid Robots." arXiv preprint arXiv:2503.14734 (2025).

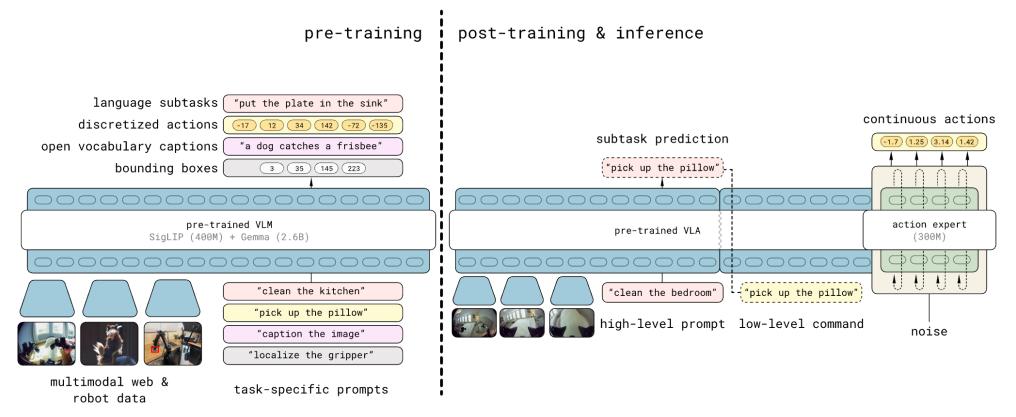


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#### $\pi_{0.5}$ | Physical Intelligence ( $\pi$ ) | Apr. 2025 (Opened in Sep. 2025)

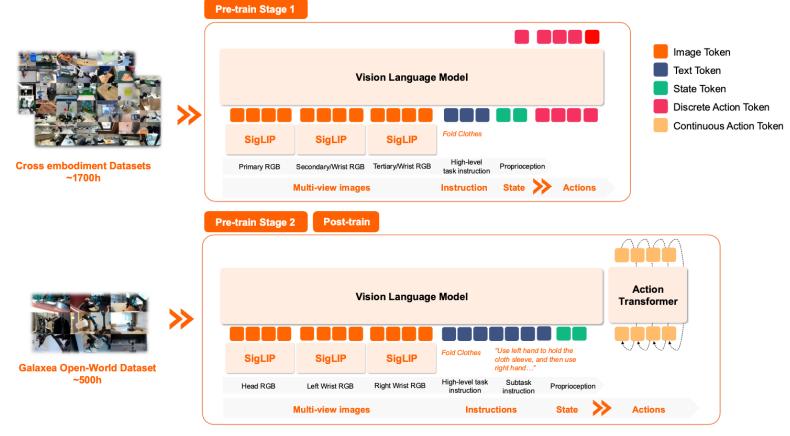
- What's new, compared to  $\pi_0$ ? for **open-world generalization**.
  - New training data: object detection, instructions & subtask commands, etc.
  - New inference flow: subtask prediction with the same model (earlier works use 2 separate models).
    - AKA the dual-system design: system 2 for high-level planning and system 1 for low-level control policy.



[1] Intelligence, Physical, et al. "\$\pi\_ {0.5} \$: a Vision-Language-Action Model with Open-World Generalization." arXiv preprint arXiv:2504.16054 (2025).

### G0 | Galaxea 星海图 | Aug. 2025 (Opened in Sep. 2025)

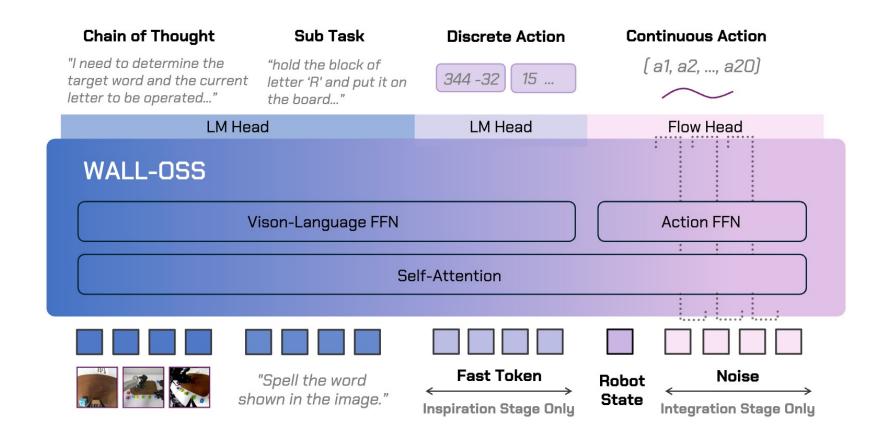
- Similar to  $\pi_{0.5}$  in model architecture and training method.
  - Single foundation model for subtask prediction, continous action generation, etc.
  - Only compared with  $\pi_0$  in paper. Model weights opened on the same day as  $\pi_{0.5}$ .



[1] Jiang, Tao, et al. "Galaxea Open-World Dataset and G0 Dual-System VLA Model." arXiv preprint arXiv:2509.00576 (2025).

### WALL-OSS | X Square Robot 自变量 | Sep. 2025 (Opened in Sep. 2025)

- Similar to  $\pi_{0.5}$ , too.
  - Only compared with  $\pi_0$ , too weights opened right before G0 and  $\pi_{0.5}$ .



[1] X Square Robot. "WALL-OSS: Igniting VLMs toward the Embodied Space." 2025, <a href="https://x2robot.cn-wlcb.ufileos.com/wall\_oss.pdf">https://x2robot.cn-wlcb.ufileos.com/wall\_oss.pdf</a>. White paper.

## **Summary & Future Look**

- VLAs evolve from discrete to continuous, from single-system to dual-system.
  - Only within 3 years (from RT-2 in 2023 to G0/WALL-OSS in 2025)!!
  - The future 3 years?
- Another trend (from my perspective): multi-tasking.
  - Model-side: SOTA VLAs are built upon <u>Internet-scale pre-trained VLMs</u>.
  - Task-side: SOTA VLAs are capable of task planning in human language.
  - A step forward: all VLM-capable tasks, with one single model (multi-expert).
    - Currently working on welcome to discuss & collaborate.



### Thanks

