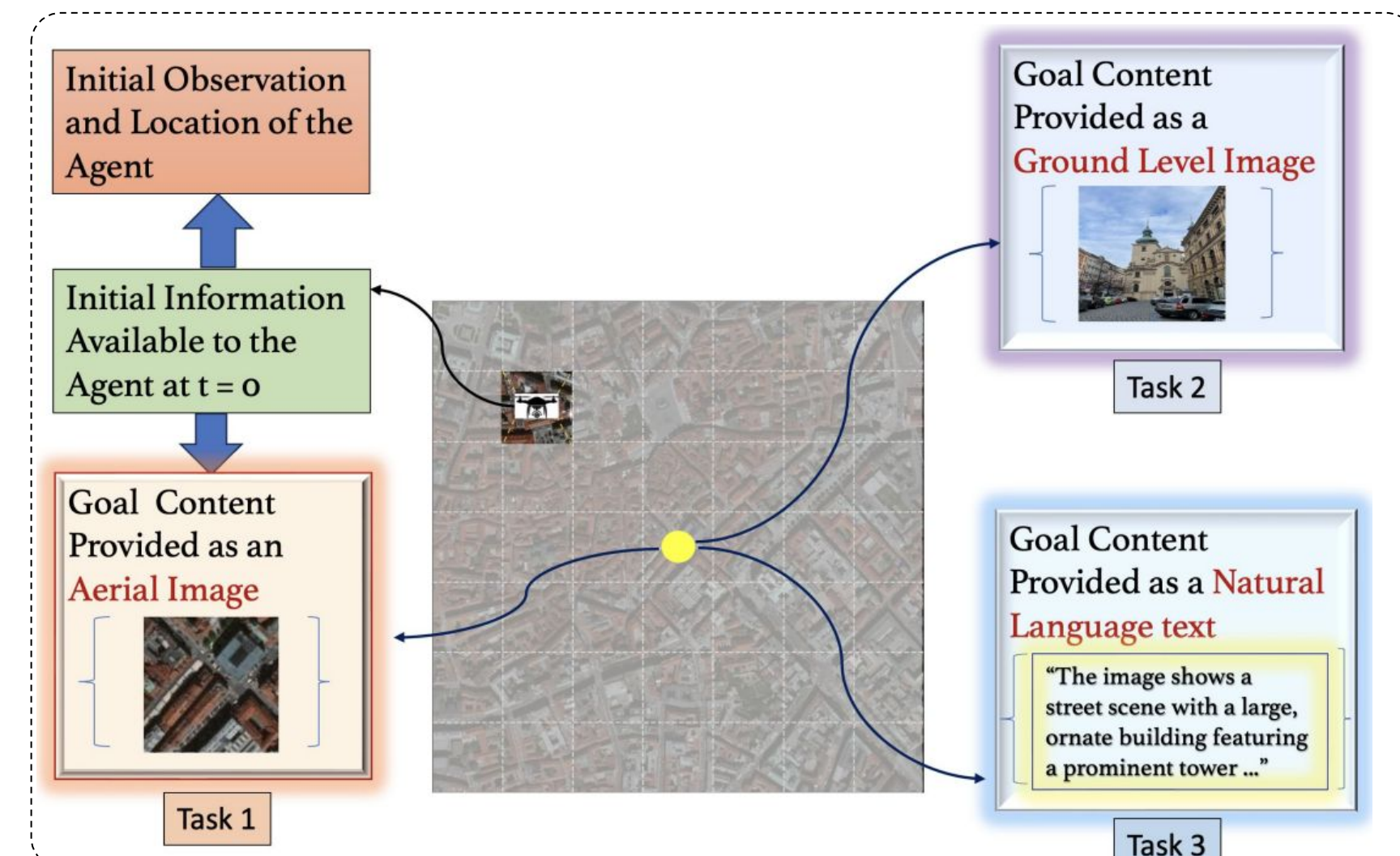


## Problem Setup



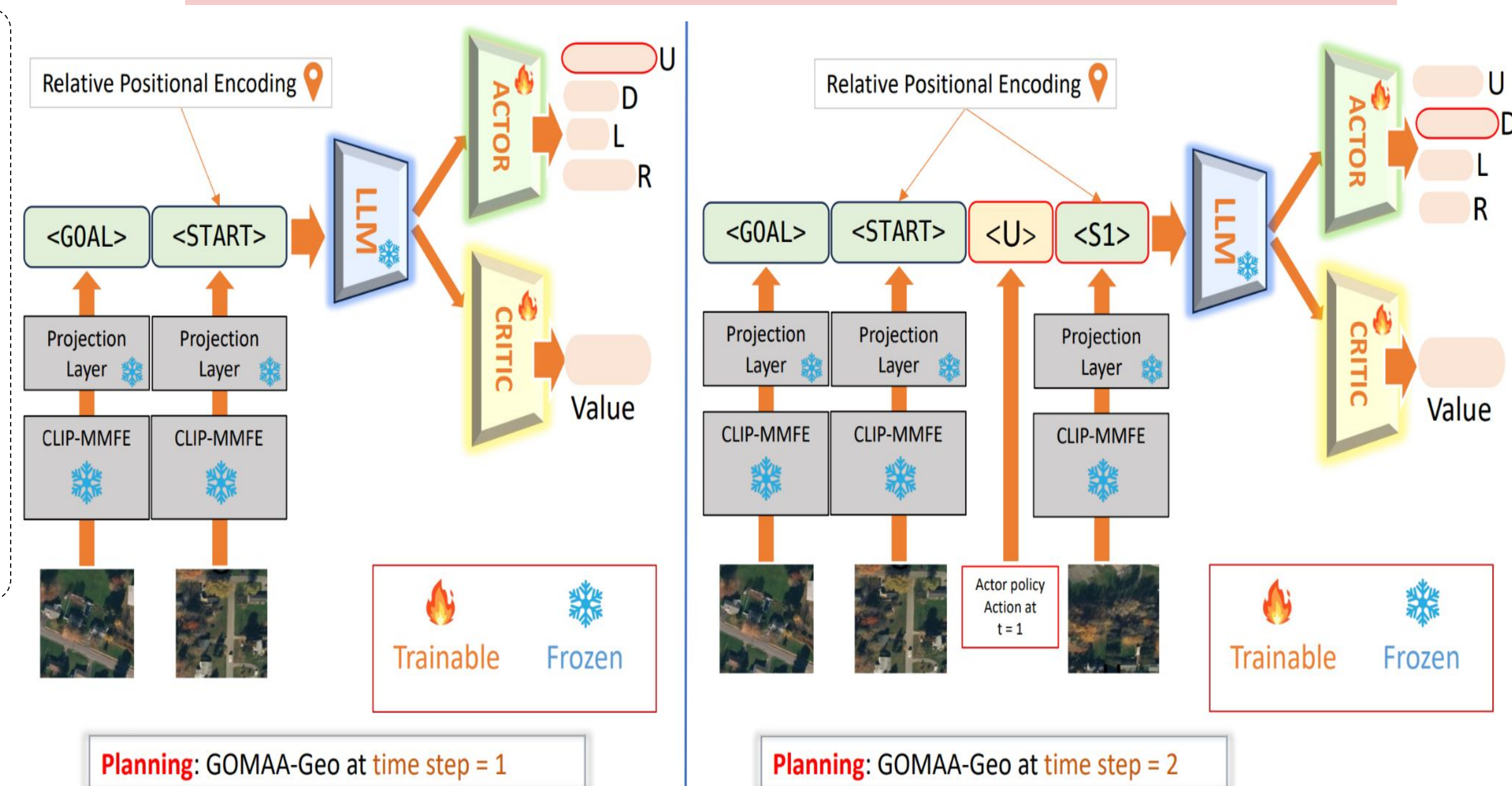
## Motivation: Towards automating search-and-rescue

- **Intensifying climate change leads** to increased frequency and severity of **natural disasters**.
- Thus also an **increased need for search-and-rescue missions**.
- **Practical constraints** (e.g. flight time, battery, expert pilots).
- A big potential for an automated approach.

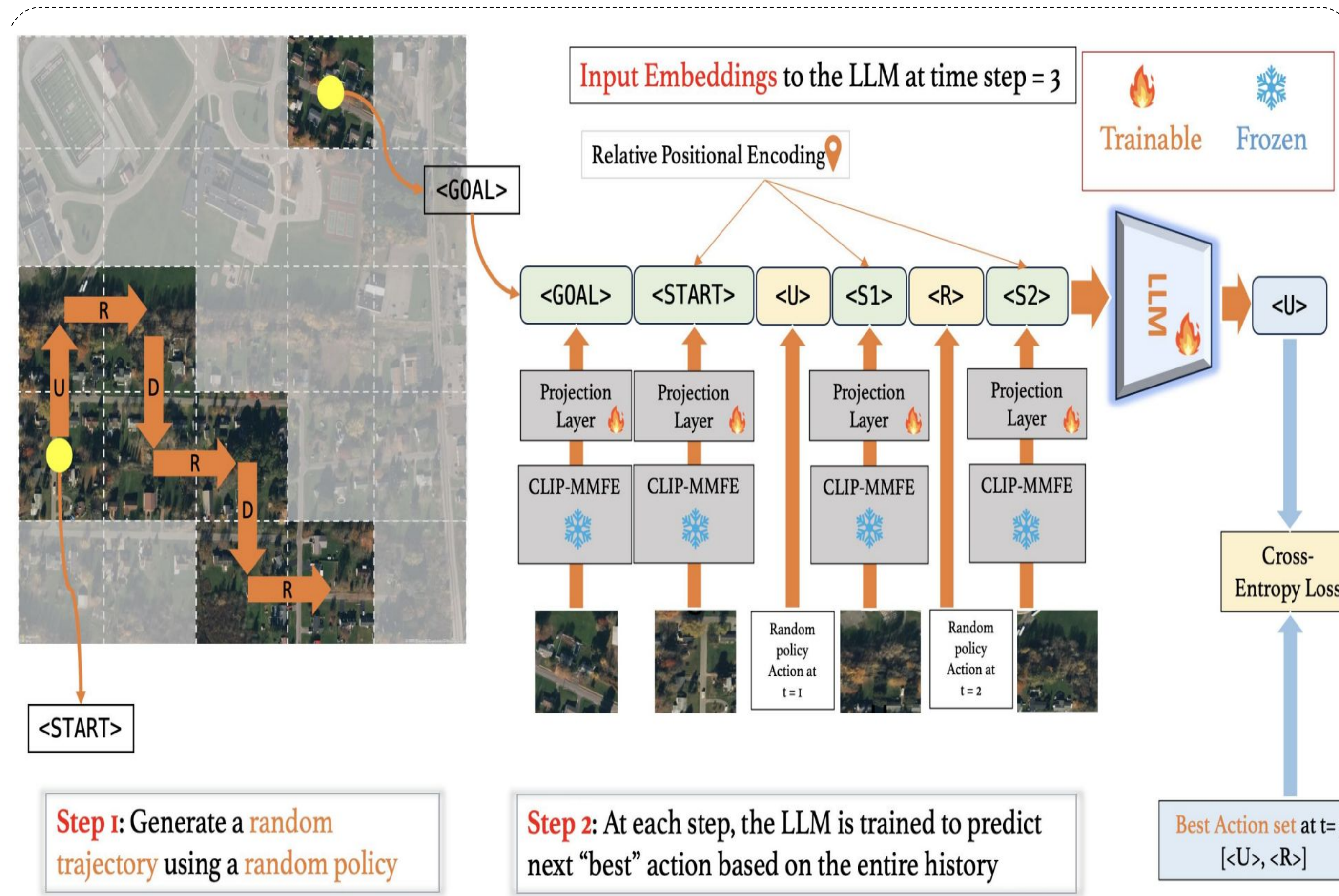
## Key steps and components of GOMAA-Geo:

- (1) Representation alignment across modalities
- (2) RL-aligned representation learning using Goal Aware Supervised LLM Pretraining (*GASP*)
- (3) Planning using RL (PPO)

## Overview of GOMAA-Geo:



**Goal modality agnostic:** Zero-shot generalization between:  
 aerial view | ground level view | natural language text



## Experimental results

**GOMAA-Geo generalizes well across goal modalities**  
 (only trained using aerial images as goal modality!)

Goal Modality	$C = 4$	$C = 5$	$C = 6$	$C = 7$	$C = 8$
Text	0.4000	0.4978	0.6766	0.7702	0.6595
Ground Image	0.4383	0.5150	0.6808	0.7489	0.6893
Aerial Image	0.4085	0.5064	0.6638	0.7362	0.7021

**RL-based planning is crucial** (LLM-based predictions alone are not sufficient for solving the AGL task)

Method	$C = 4$	$C = 5$	$C = 6$	$C = 7$	$C = 8$
LLM-Geo	0.2331	0.2591	0.3121	0.3967	0.4051
<b>GOMAA-Geo</b>	<b>0.4090</b>	<b>0.5056</b>	<b>0.7168</b>	<b>0.8034</b>	<b>0.7854</b>

**GOMAA-Geo outperforms alternative methods**

(tested on post-disaster, but trained on pre-disaster data)

Method	$C = 4$	$C = 5$	$C = 6$	$C = 7$	$C = 8$
Random	0.1412	0.0584	0.0640	0.0247	0.0236
PPO	0.1132	0.1146	0.1292	0.1665	0.1953
DiT	0.1012	0.2389	0.3067	0.3390	0.3543
AiRLoc	0.1201	0.1298	0.1507	0.1631	0.1989
<b>GOMAA-Geo</b>	<b>0.4002</b>	<b>0.4632</b>	<b>0.6553</b>	<b>0.7391</b>	<b>0.6942</b>

## Visualization of GOMAA-Geo

