

# OPEN SCENE GRAPHS FOR OPEN-WORLD OBJECT GOAL NAVIGATION





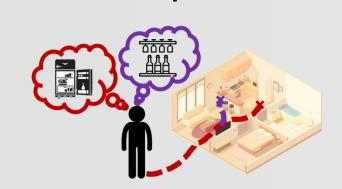
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Joel Loo\* [joell@u.nus.edu], Zhanxin Wu\* [zhanxinwu@u.nus.edu], David Hsu [dyhsu@comp.nus.edu.sg]

## Open-World ObjectNav

Can we search novel scenes for an open-set object class, in *any* environment, with *any* embodiment?

 Requires semantic reasoning, common-sense priors



 Must handle diverse instructions, environments & embodiments



Approach: Compose an ObjectNav robot system purely built from Foundation Models (FMs)

Problem: Need a structured scene memory to retain information for FMs, that is also built with FMs

### OpenSearch, a general system for Open-World ObjectNav





Open-set instructions





Different embodiments

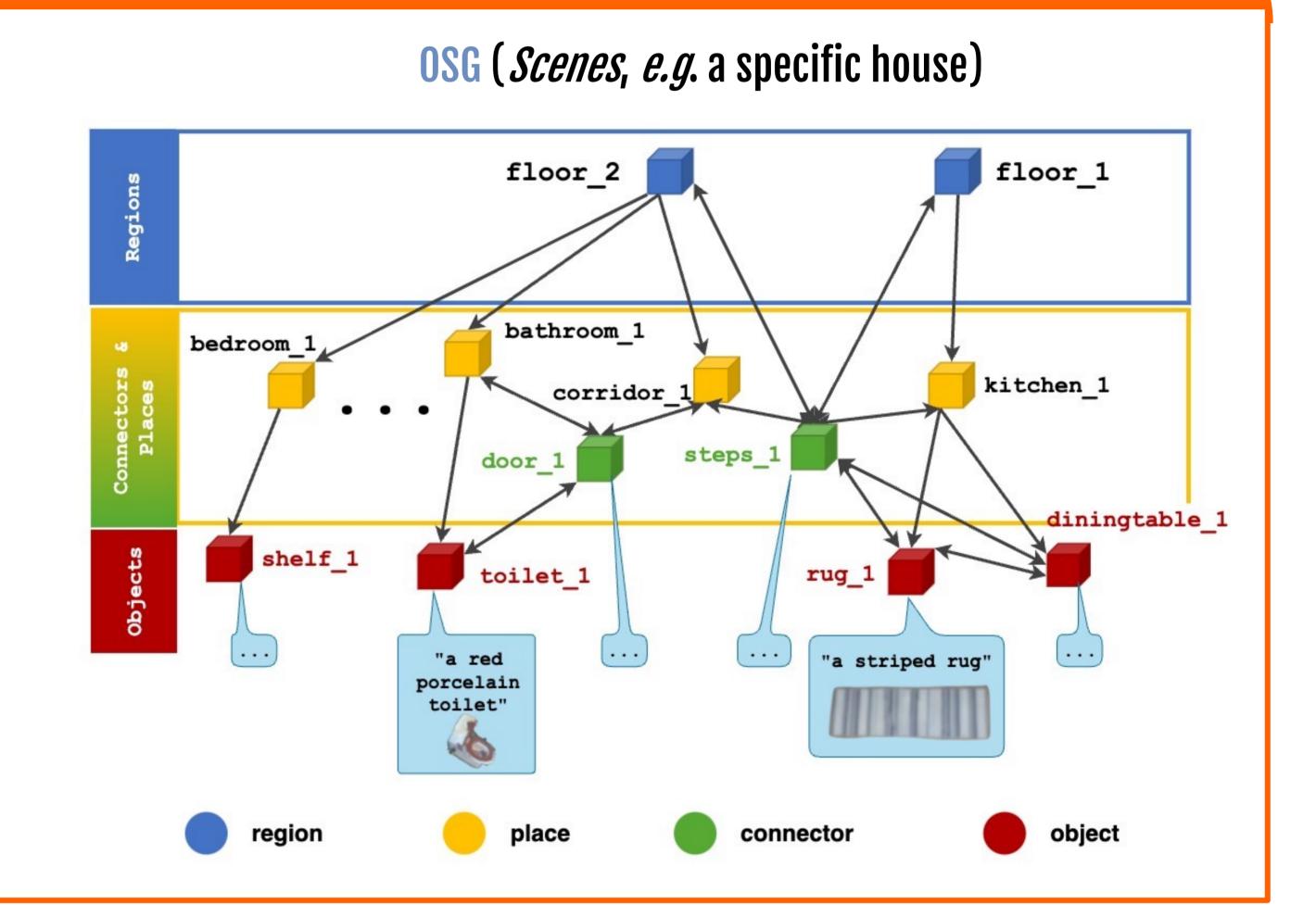
### Open Scene Graphs

### OSG Schema (*Environment types*, e.g. homes)

```
"RegionAbstraction1_Layer": {
```

- # [Optional] Semantically meaningful spatial abstractions
- "layer\_type": "floor",
- "Places Layer": {
- # [Required] Smallest semantically meaningful regions
  - "layer type": "room",
- "Connectors\_Layer": {
- #[Optional] Connective structures between regions
  - "layer\_type": ["stairs", "doors"]
- "Objects\_Layer": {
- # [Required] Task-relevant semantic features

# specifies structure of



### OSG Mapper: Builds OSGs online, following OSG spec

- Uses only FMs: i.e., LLM, open-vocab VQA, open-set Object Detector
- Uses objects as features for data association

```
 \begin{array}{l} \text{fn ImageParser} (I_t) \text{:} \\ & // \textit{Extract Place, Objects, Connectors from observations} \\ P_t = \texttt{LABELPLACE\_VLM\_VQA}(I_t) \\ D_t = \texttt{DETECTOBJECTSCONNECTORS\_VLM\_OD}(I_t) \\ O_t, C_t = \texttt{CLASSIFYOBJECTSANDCONNECTORS\_LLM}(D_t) \\ \text{for } o \ in \ O_t \cup C_t \ \text{do} \\ & \mid o.attr = \texttt{LABELWITHTEXTUALATTRIBS\_VLM\_VQA}(o) \\ \text{return } P_t, O_t, C_t \\ \end{array}   \begin{array}{l} \text{fn StateEstimator} (P_t, O_t, C_t) \text{:} \\ & | // \textit{Estimate robot's current location (Place)} \\ P^{OSG} = \texttt{SORTBYDISTANCE} (all \textit{Place nodes in OSG}) \\ \text{for } p \ in \ P^{OSG} \ \text{do} \\ & | \ \text{if IsPLaceMatchedWithObs\_LLM}(p, O_t \cup C_t) \\ & | \ \text{return } p \\ \text{return None} \\ \end{array}
```

```
fn OSGUpdater (\hat{p}, O_t, C_t):
   // Update OSG with observed Place, Objects, Connectors
   if \hat{p} is None
       p_t = \text{AddPlaceNode}(\hat{p})
       ADDOBJECTCONNECTORLEAFNODES(O_t, C_t)
       ADDEDGES(s_t, O_t, C_t)
       INFERREGIONABSTRACTIONS_LLM(p_t)
   else
       p_t = \hat{p}
       for o in O_t \cup C_t do
           \mathcal{V}_{leaf} = \text{GetObjectConnectorLeafNodes}(p_t)
           v_{match} = \text{FINDMATCHEDLEAFNODE\_LLM}(o, \mathcal{V}_{leaf})
           if v_{match} is None
               AddNewLeafNode(o)
               UPDATELEAFNODE(o)
           UPDATEEDGES(o)
   return p_t
```

#### **OSG Navigator** Prev. Image state osg Open-vocab Specification Image State OSG Updater Estimator OSG Region Goal Proposer Proposer Pathfinder \$ $\mathcal{P} = [I_1, \dots, I_n]$ General Navigation Model (GNM) Actions $\rightarrow (v,\omega)$ Image-goal Visuomotor Policy

### **Experiments: Simulation**

Comparison With LLM-based ObjectNav Methods On HM3D

Method	Success (↑)	SPL (†)	<b>DTG</b> (m) (↓)
Greedy LLM (based on [11])	0.275	0.080	5.078
LFG [39]	0.675	0.389	2.411
Explorer-FMM-GT	0.775	0.380	1.702
Explorer-FMM	0.693	0.283	2.338

### Comparison With ObjectNav Methods On Gibson

Method	Success (†)	SPL (†)	DTG (\dagger)	TF	NM
SemExp [5]	0.657	0.339	1.474	×	×
PONI [31]	0.736	0.410	1.250	X	X
FBE [50]	0.641	0.283	1.780	1	×
SemUtil [8]	0.693	0.405	1.488	✓	X
Explorer-FMM	0.734	0.386	1.722	<b>✓</b>	<b>√</b>

(TF: training free. NM: Non-metric)



- LLM reasoning with Open Scene Graphs lets OpenSearch strongly outperform LLM-based methods
- LLMs' rich semantic priors lets the zero-shot OpenSearch perform on par with task-specific learned methods