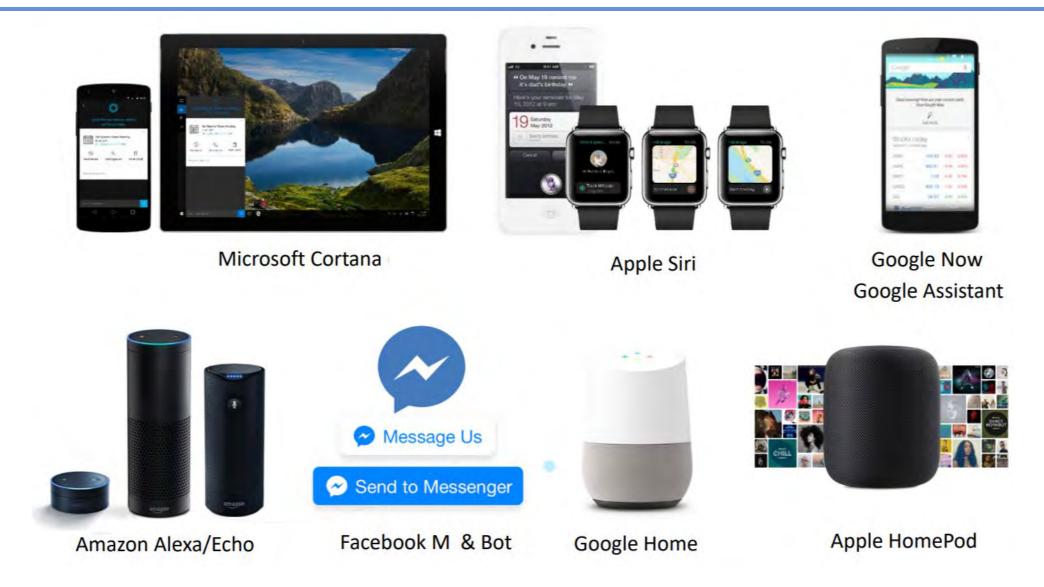
# Grounding Natural Language for Building Embodied Agents

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Acknowledgements:

Xin Wang (UCSB), Qiuyuan Huang (MSR), Dinghan Shen (Duke), Xiujun Li (MSR) Jianfeng Gao (MSR), Lei Zhang (MSR), Yonatan Bisk (MSR), William Wang (UCSB)

### Language Empowering Intelligent Agents



### Adapting Agents to Physical Environments



Image Source : Henderson Biomedical

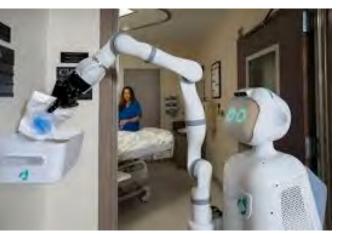


Image Source : boingboing.net



Image Source : Boston Dynamics

# Outline

- Language grounding in visual environments
  - Visual Language Navigation Task
  - Self-supervised imitation learning [CVPR 2019]
- Ongoing Work
  - Navigation and Dialog
  - situated and bi-directional

#### Our Goal $\rightarrow$ Build intelligent agents

- Communicate with people
  - Follow natural language instructions
- Understand the dynamics of the perceptual environment
- Alignment between the two !

## Language Grounding in Situated Environments

Linguistic symbols



Perceptual experiences and actions

#### (Noun Phrase) dog reading newspaper



(Verb) sleeping



(Verb Phrase) climb on chair to reach switch







## Understanding Visually Grounded Language



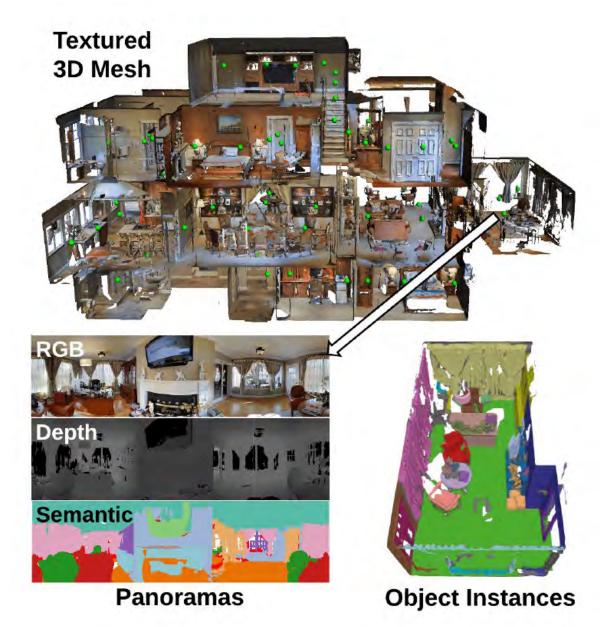
#### Understanding Visually Grounded Language

#### **TASK:** Vision & Language Navigation (VNL)

Navigating an agent inside real 3D environments by following natural language instructions.

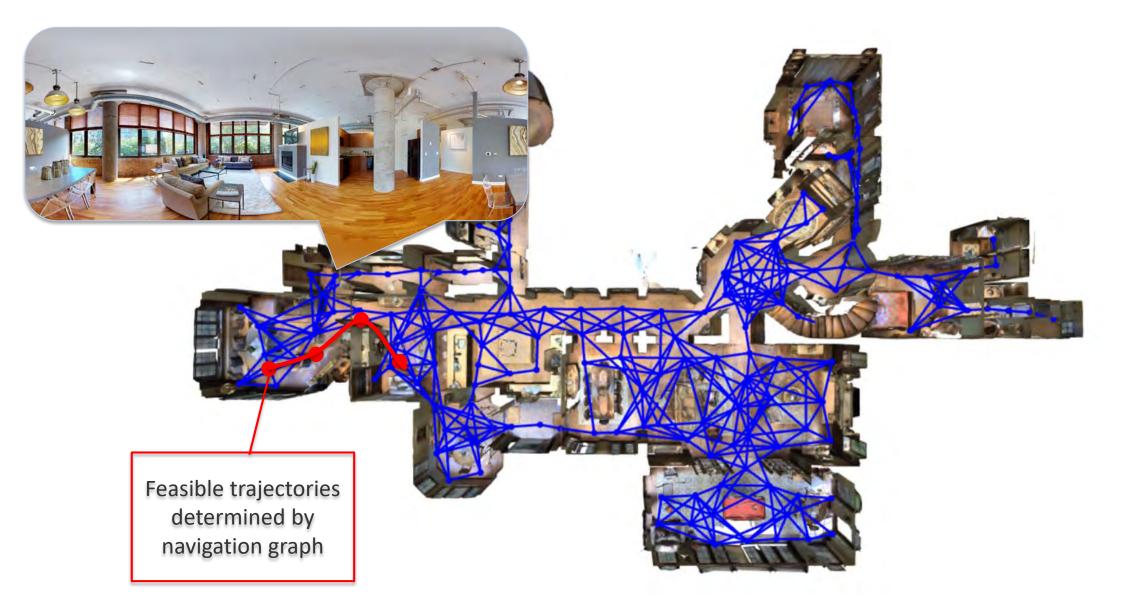


### Matterport 3D Dataset



- 10,800 panoramic views based on 194K RGB-D images
- 90 building-scale scenes (avg. 23 rooms each)
- Includes textured 3D meshes with object segmentations
- Largest RGB-D dataset

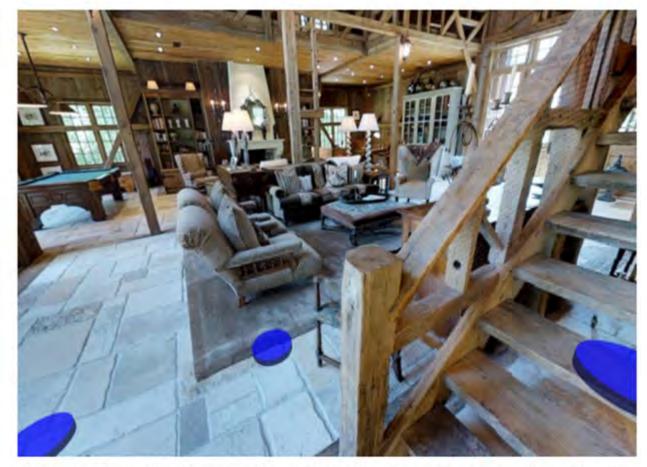
### Matterport 3D Simulator for VLN Task



## Matterport 3D Simulator for VLN Task

#### Room-to-Room (R2R) Dataset

- ~7K shortest paths
- 3 instructions for each path
  - Average instruction length 29 words
  - Average trajectory length is 10 meters
- Task: given natural language instructions, find the goal location!



**Instruction:** Head upstairs and walk past the piano through an archway directly in front. Turn right when the hallway ends at pictures and table. Wait by the moose antlers hanging on the wall.

#### Room-to-Room Dataset Examples



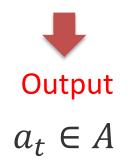
Leave the bedroom, and enter the kitchen. Walk forward, and take a left at the couch. Stop in front of the window.

#### Input: Instruction

turn completely around until you face an open door with a window to the left and a patio to the right, walk forward, ... ...

#### Input: Panoramic View





#### Visual-Language Navigation Task Challenges

#### (1) cross-modal grounding

**Instruction:** Go towards the *living room* and then turn right to the *kitchen*. Then turn left, pass a *table* and enter the *hallway*. Walk down the *hallway* and turn into the *entry way* to your right. Stop in front of the *toilet*.

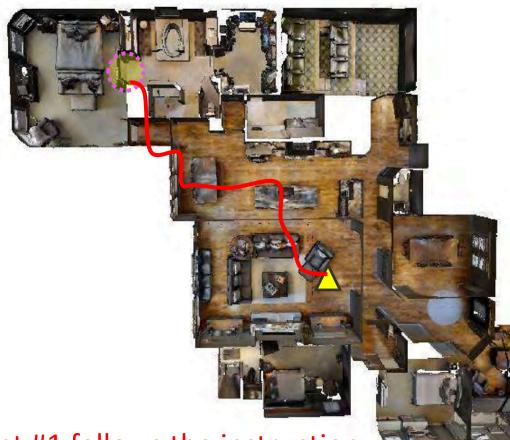


Global trajectory in top-down view (NOT visible to the agent)

### Visual-Language Navigation Task Challenges

# (1) cross-modal grounding(2) ill-posed feedback

**Instruction:** Go towards the *living room* and then turn right to the *kitchen*. Then turn left, pass a *table* and enter the *hallway*. Walk down the *hallway* and turn into the *entry way* to your right. Stop in front of the *toilet*.



Agent #1 follows the instruction and reaches the destination.

Agent #2 randomly walks insides the house and reaches the destination.

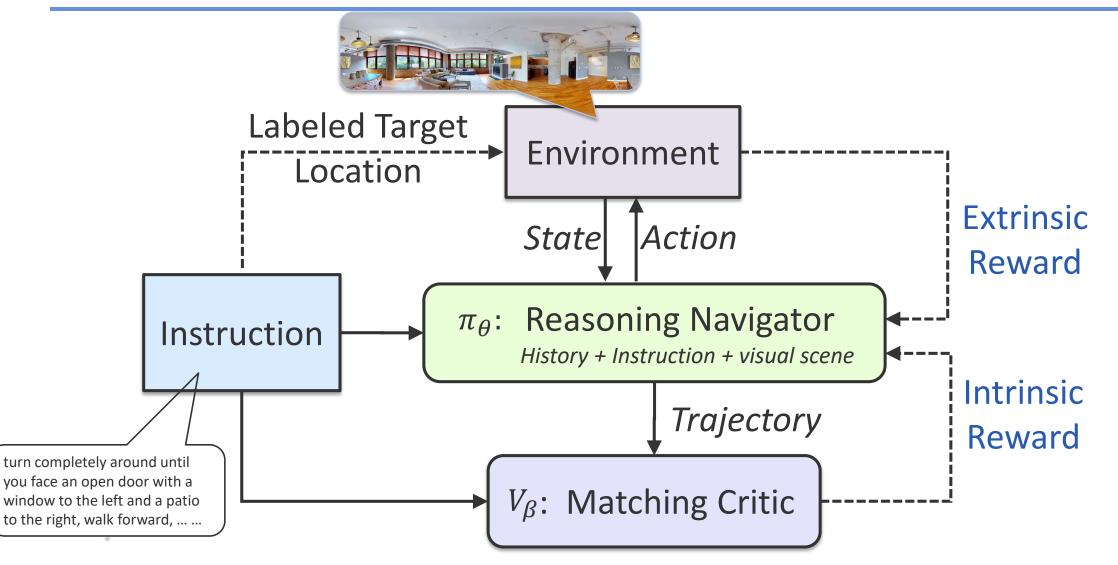
Both trajectories are considered same in terms of the success signal.

#### Our Recent "Explanatory" Work on VLN

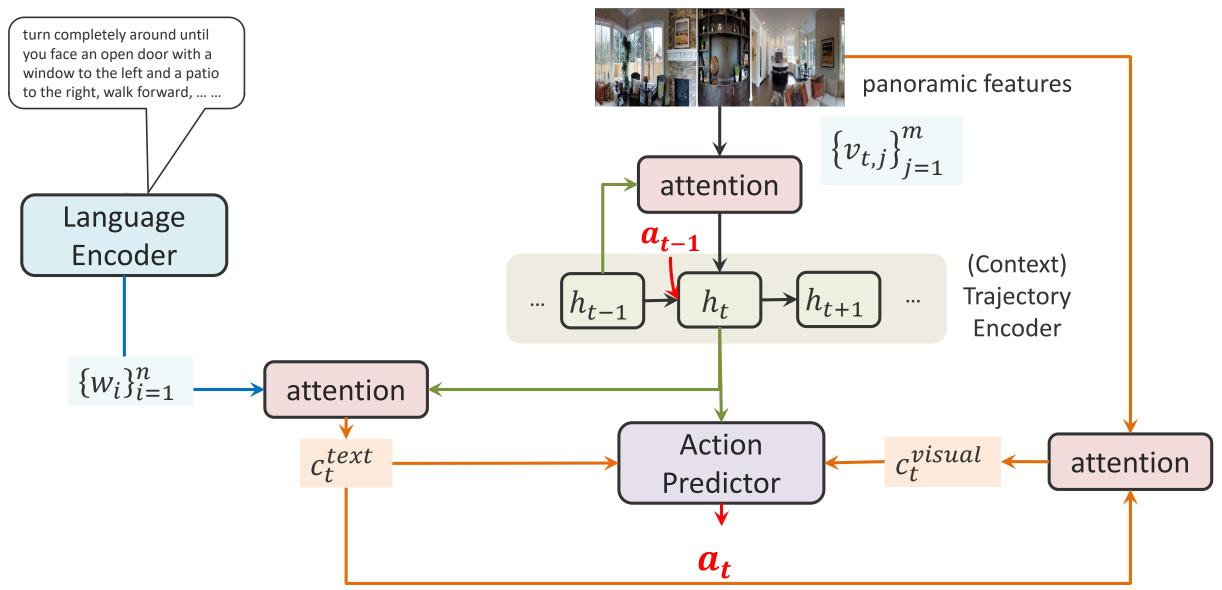
Reinforced Cross-Modal Matching and Self-Supervised Imitation Learning for Vision-Language Navigation, CVPR 2019

Learns to ground language in visual context using <u>RL</u> and <u>Self-Supervised</u> <u>Imitation Learning</u>

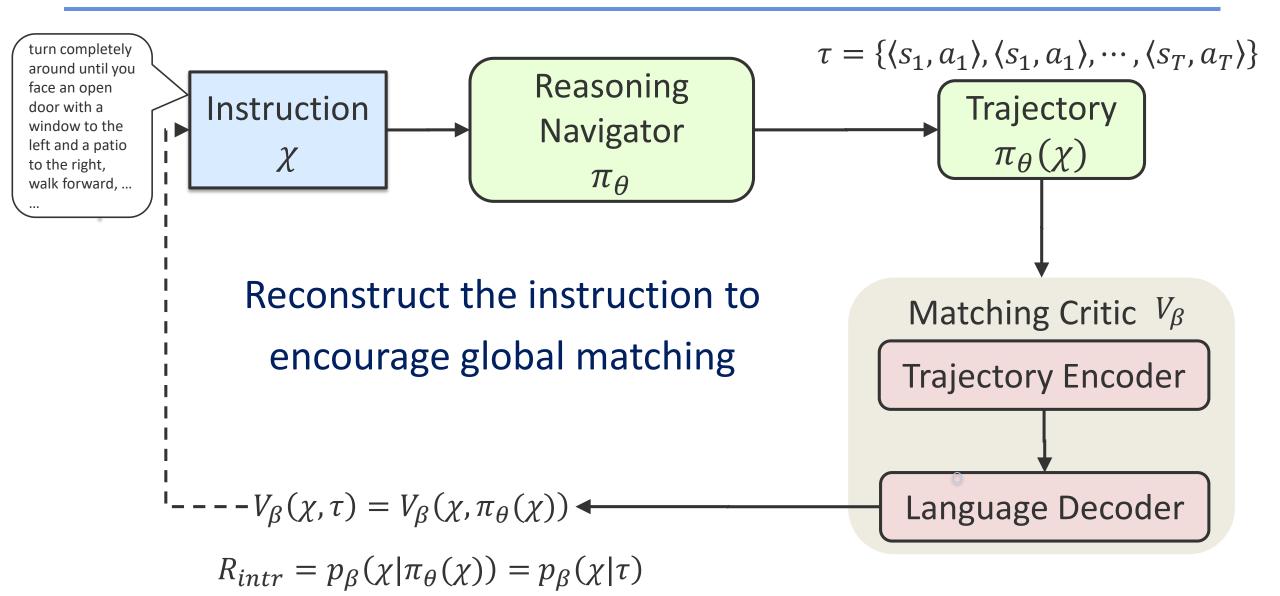
## Reinforced Cross-modal Matching (RCM)



#### **Cross-Modal Reasoning Navigator**



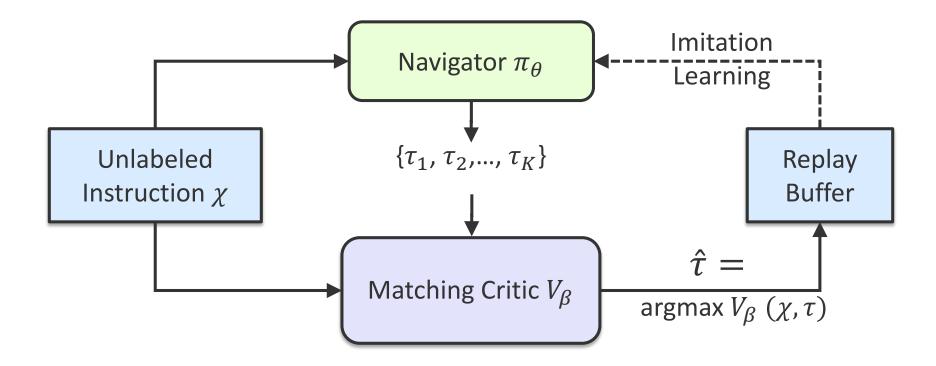
## Matching Critic $\rightarrow$ Intrinsic Reward



### Visual-Language Navigation Task Challenges

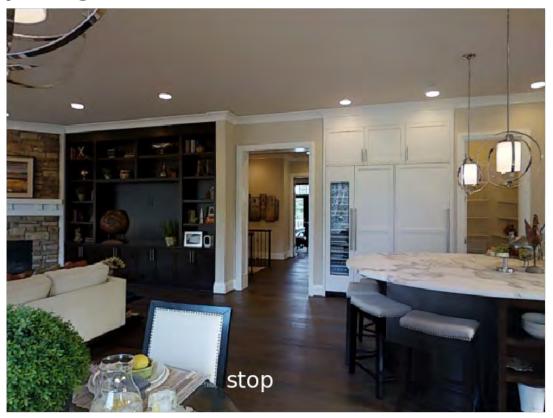
(1) cross-modal grounding(2) ill-posed feedback(3) generalization

## Self-supervised Imitation Learning (SIL)



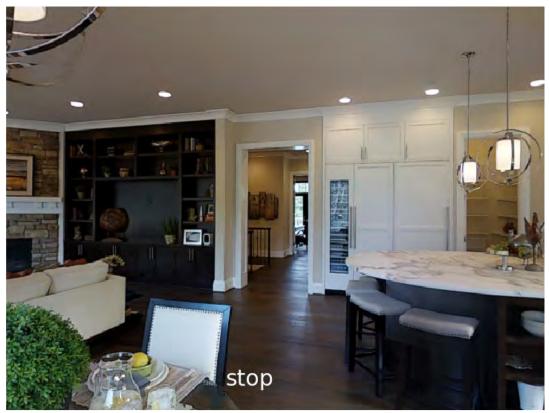
Learning from its previous good behaviors  $\rightarrow$  better policy that adapts to new environments

**Instruction:** Turn right and head towards the kitchen. Before you get to the kitchen, turn left and enter the hallway. ... Walk forward and stop beside the bottom of the steps *facing the double white doors*.

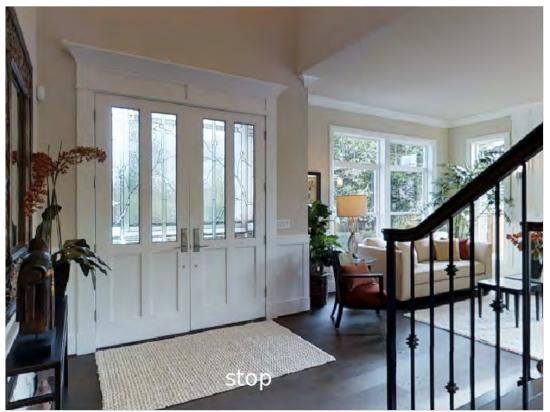


Reinforcement Learning Only

**Instruction:** Turn right and head towards the kitchen. Before you get to the kitchen, turn left and enter the hallway. ... Walk forward and stop beside the bottom of the steps *facing the double white doors*.



Reinforcement Learning Only



#### RL + Self-Supervised Imitation Learning

step 1 panorama view



step 2 panorama view



step 3 panorama view



step 4 panorama view



#### step 5 panorama view



relative heading from the agent

**Instruction:** Go up the stairs to the right, turn left and go into the room on the left. Turn left and stop near the **mannequins**.

#### Intrinsic Reward : 0.51 Result : Failure (error = 3.1m)



#### What's next ?

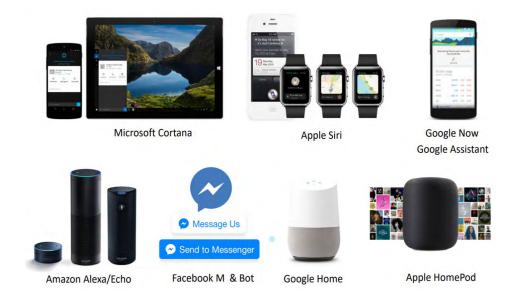


Image Source : Microsoft/HoloLens web page



Image Source : mirror.co.uk

#### Situated Reasoning Machines

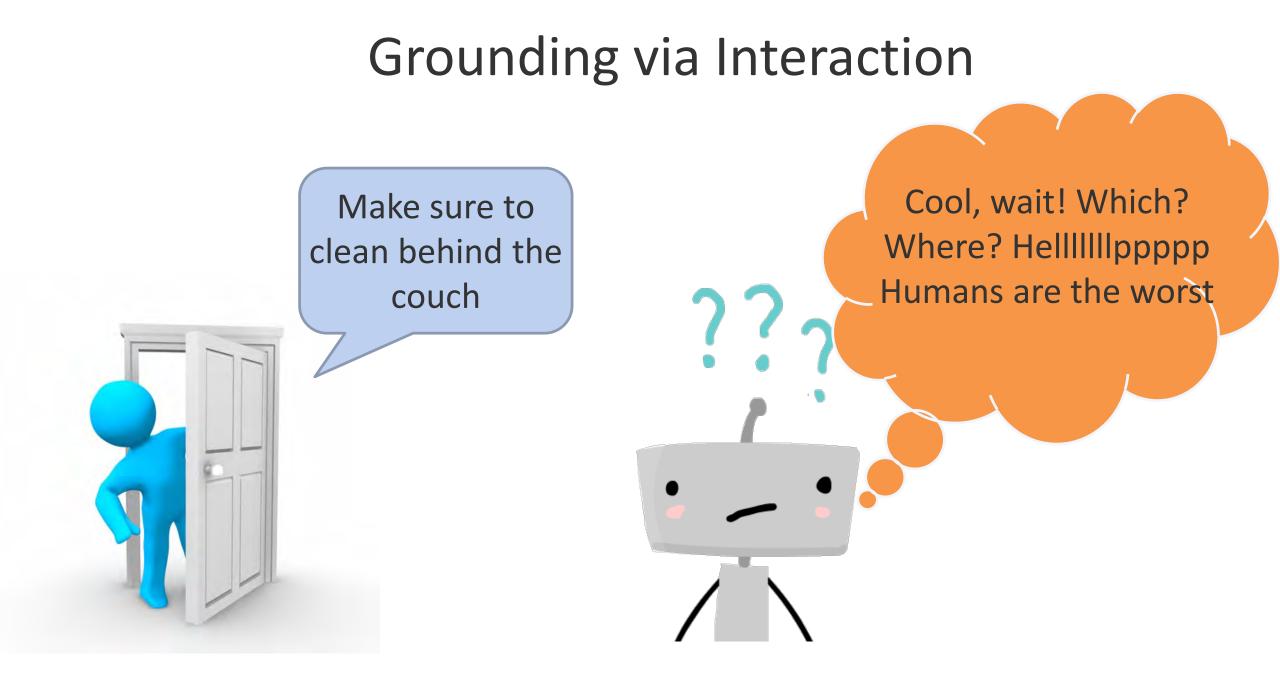


#### Language Empowered Agents Bi-directional but not situated!



#### Situated Language Empowered Agents Situated but uni-directional !

Situated Reasoning Machines Bi-directional and situated!



Vision and Dialog Navigation

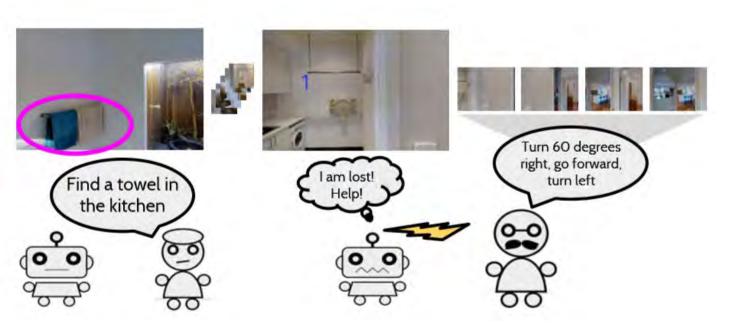
Connecting Language and Vision

What's the meaning of "the second door on the right?"

#### Modeling uncertainty

How does an agent know if it's lost or confused?

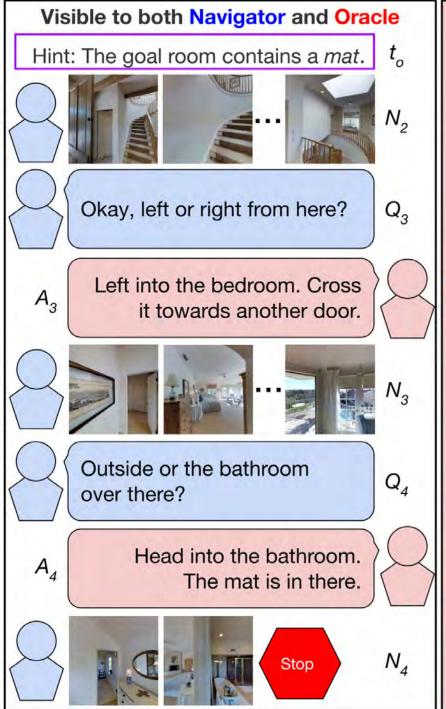
#### • NL Question and Answer generation Provide targeted feedback

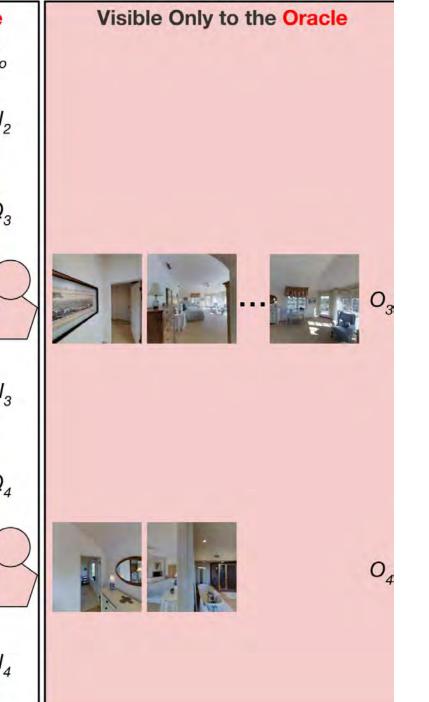


Vision-based Navigation with Language-based Assistance via Imitation Learning with Indirect Intervention CVPR 2019

Data + Model

K. Nyugen (UMD), **D. Dey** (MSR), C. Brocket (MSR), B. Dolan (MSR)





## Interaction Snapshot

Goal: Build both the Navigator and the Oracle systems

#### Current SOTA? 0% Brand new dataset!

# Briefly ...

- Situated Unidirectional Task: Visual Language Navigation
  - **RL agent** navigating **3D** environment
  - Cycle loss to evaluate local and global path behavior
  - Imitation learning via **self supervision**

- Situated Bi-directional Task : Visual+Dialog Navigation (VDN)
  - Learn to ask questions
  - Transfer from previous tasks : Unimodal Dialog, Visual Dialog, VLN, etc.
  - Meta-learn !

# Thank you !

#### QUESTIONS