

Shifting the Baseline: Single Modality Performance on Visual Navigation & QA Jesse Thomason, Daniel Gordon, and Yonatan Bisk University of Washington

Unimodal Baselines

We demonstrate the strength of unimodal baselines in multimodal domains. We argue that unimodal approaches better capture and reflect dataset biases than random or majority class baselines, and therefore provide an important comparison when assessing the performance of multimodal techniques.

Evaluation Framework

Punchline

Unimodal Evaluation

Language-only and vision-only VLN and QA models outperform published baselines and even beat their multi-modal counterparts!

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Best unimodal in **bold**, blue indicates better than baseline; and * indicates better than full model.

Visual Navigation.

	Matter
Model	Seen
Full Model <u>م</u>	27.1
Baseline	15.9
A	18.5

	Matterport [↑]		IQA↑		EQA↓	
	(%)		(%)		(<i>m</i>)	
Model	Seen	Un	Seen	Un	Un	
Full Model	27.1	19.6	77.7	18.08	4.17	
Baseline	15.9	16.3	2.18	1.54	4.21	
$\overline{\mathcal{A}}$	18.5	17.1	4.53	2.88	4.53	
$\frac{1}{2}\mathcal{A}+\mathcal{V}$	21.2	16.6	35.6	7.50	* 4.11	
$\mathcal{A} + \mathcal{L}$	23.0	* 22.1	4.03	3.46	4.64	
∆Uni – Base	+7.1	+5.8	+33.4	+5.96	-0.10	

We ablate benchmark models from:

- 1. Matterport Room-2-Room Navigation – (*Anderson et al.*, *CVPR'18*);
- 2. **THOR** Interactive Question Answering – (*Gordon et al., CVPR'18*);
- 3. EQA Embodied Question Answering – (*Das et al., CVPR'18*). We define three ablations:
- Full Model is $\mathcal{M}(\mathcal{V}_t, \mathcal{L}, a_{t-1}; W)$ is $\mathcal{M}(\vec{0}, \vec{0}, a_{t-1}; W)$ \mathcal{A} is $\mathcal{M}(\mathcal{V}_t, \mathbf{0}, \mathbf{a}_{t-1}; W)$ $\mathcal{A} + \mathcal{V}$ is $\mathcal{M}(\mathbf{0}, \mathcal{L}, \mathbf{a}_{t-1}; \mathbf{W})$ $\mathcal{A} + \mathcal{L}$
- with \mathcal{A} ction inputs, \mathcal{V} ision inputs, and Language inputs.
- At each timestep an agent receives an observation and produces an action.

ble gains from single-modality biases in multimodal datasets irrespective of training and architecture details.

Recommendation for Best Practices:

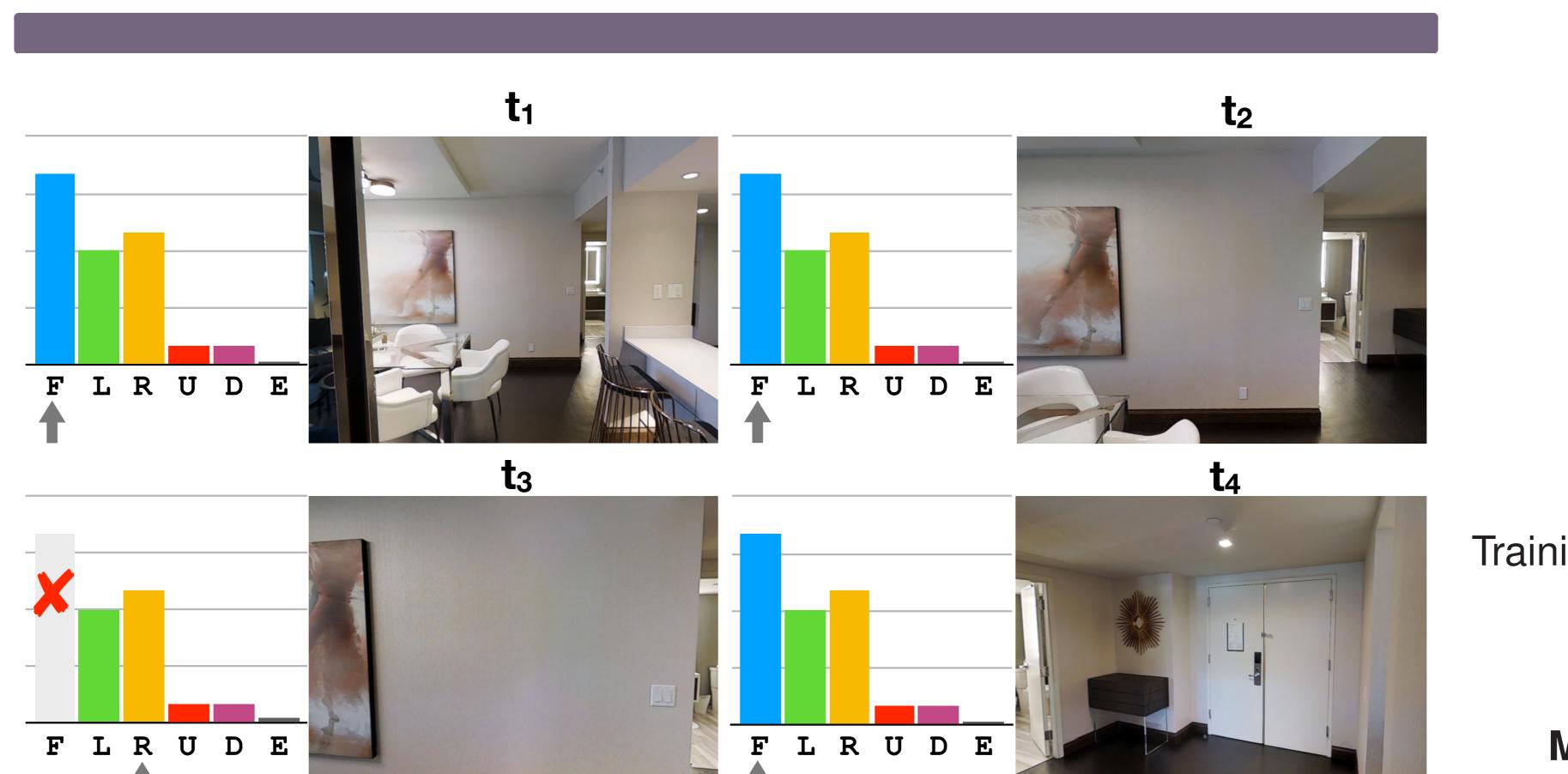
While many papers ablate either language

or vision, researchers should ablate both.

These unimodal baselines expose possi-



Training via behavior cloning.



Matterp	ort
(%)	

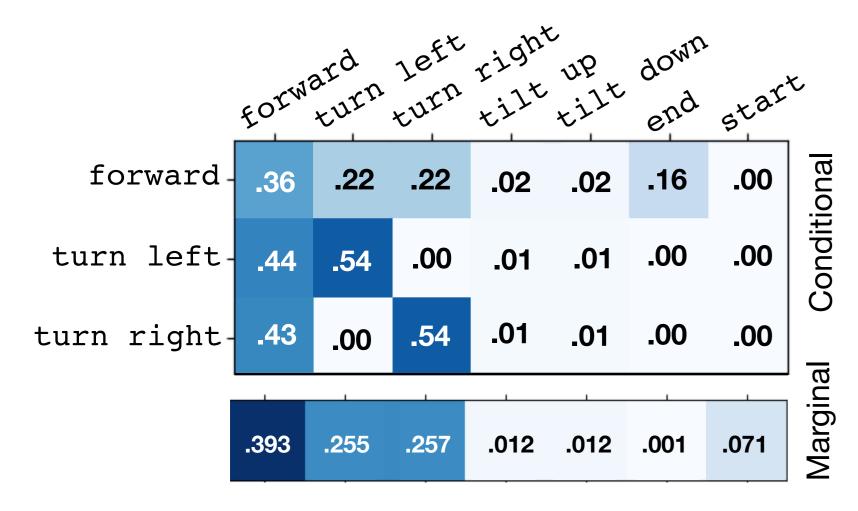
	$(, \mathbf{U})$			
Model	Seen	Un		
Full Model و	38.6	21.8		
Baseline	15.9	16.3		
\mathcal{A}	4.1	3.2		
$\overline{\boldsymbol{\zeta}} \mathcal{A} + \mathcal{V}$	30.6	13.3		
$-\mathcal{A}+\mathcal{L}$	15.4	13.9		
Δ Uni – Base	+14.7	-2.4		
	c 1			

Training via student forcing.

 $\mathbf{d}_{\mathsf{T}}\downarrow$ (m)Model I_{-10} *I*₋₅₀ I_{-30} Full ن 0.971 4.17 8.83

 $a_t \leftarrow \mathcal{M}(\mathcal{V}_t, \mathcal{L}, a_{t-1}; W)$

(1)

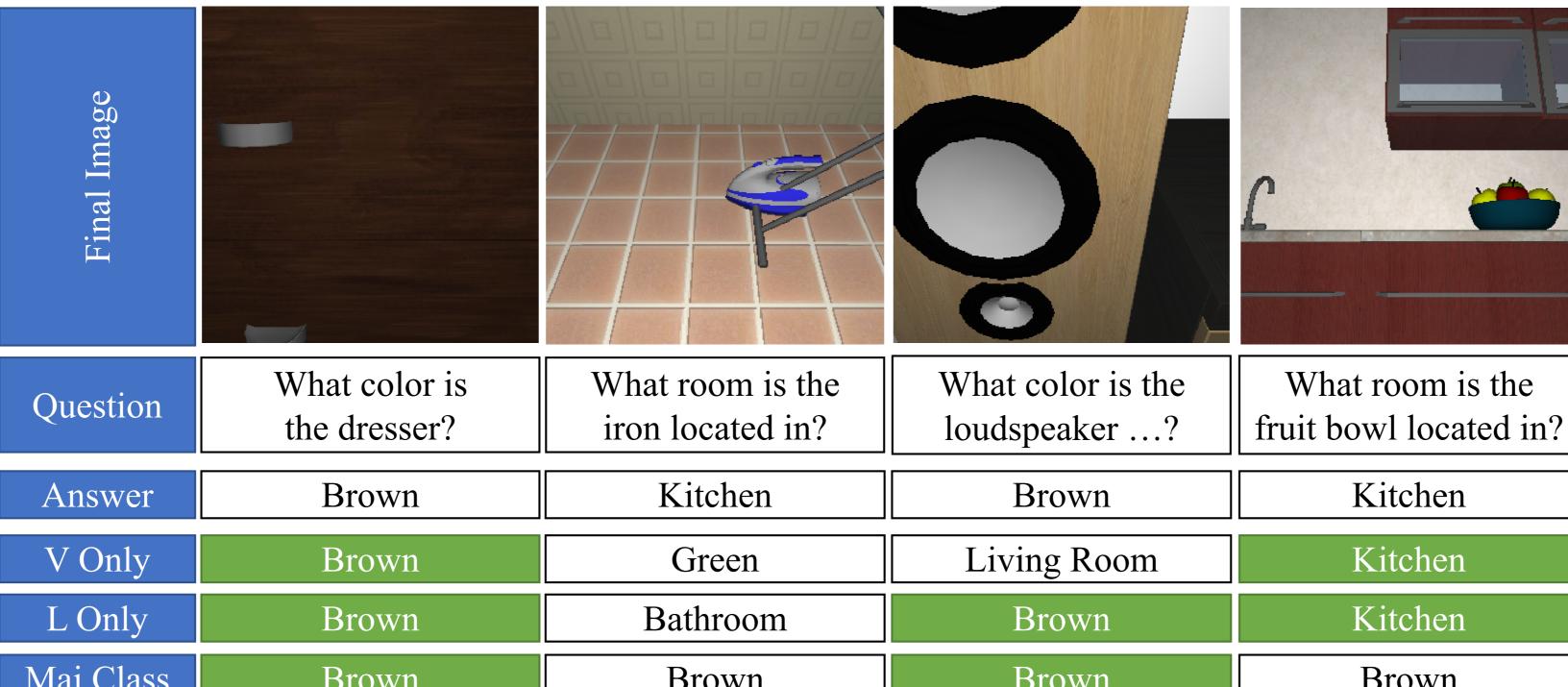


P(act = col|prev = row) and marginal action distributions in Matterport data reveal memorizable peakiness.



Actions: Forward, turn Left & Right, tilt Up & Down, End

In the Matterport Room-2-Room task, navigating without vision can lead to sensible navigation trajectories in response to commands like "walk past the bar and turn right". At t_3 , "forward" is unavailable as the agent would collide with the wall, rendering the visual context for the command unnecessary.



Baseline 1.020 4.21 8.73 ***0.893** 4.53 9.56 \mathcal{A} $\overleftarrow{\mathsf{D}} \mathcal{A} + \mathcal{V}$ [†]8.83 *0.951 ***4.11** 0.987 4.64 9.51 $\mathcal{A} + \mathcal{L}$ △ Uni – Base -0.127 -0.10 +0.10 EQA navigation final distance.

	$d_{min}\downarrow$				
	(<i>m</i>)				
Model	T_{-10}	T_{-30}	T_{-50}		
ج Full Baseline	0.291	2.43	6.45		
Baseline	0.293	2.45	6.38		
\mathcal{A}	*0.242	3.16	7.99		
$ \mathcal{A} + \mathcal{V}$	*0.287	2.51	*6.44		
$-\mathcal{A} + \mathcal{L}$	*0.240	3.19	7.96		
∆ Uni – Base	-0.053	+0.06	+0.06		
EQA navigation closest distance.					

Question Answering.

Relevant visual observations are made after navigating to an implicit goal point in QA tasks.

Iviaj Class	DIOWII	DIOWII	DIOWII	DIOWII				
Full Model	Brown	Bathroom	Brown	Kitchen		IQ/		EQA ↑
			Model	Un	Seen	Un		
Qualitative results on the EQA task illuminate some unimodal bi-			Full Model و	88.3	89.3	64.0		
ases in the data. The language only model can pick out the most			Baseline	41.7	41.7	19.8		
likely answer for a given question without visual context. The vi-			$ = \mathcal{V} ONLY $	43.5	42.8	44.2		
sion only model finds and reports salient color and room feature			$\supset \mathcal{L}$ only	41.7	41.7	48.8		
as answ	ers without be	ing aware of th	ne question.		∆ Uni – Base	+1.8	+1.1	+29.0
		-	-					

Question answering accuracy.

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https://arxiv.org/abs/1811.00613