### UNIFIED-IO: A UNIFIED MODEL FOR VISION, LANGUAGE, AND MULTI-MODAL TASKS



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### Single-Task Model vs. Unified Model

Single-Task Model A set of parameters that optimize for single tasks. Shared parameters + task specific heads for multiple tasks. Multi-Task Model Unified Model

Shared parameters + shared heads for multiple tasks.



(1) Contrastive pre-training



(2) Create dataset classifier from label text



CLIP: Connecting Text and Images [Radford et.al. 2021]



) Create dataset classifier from label text



an illustration of a baby daikon radish in a tutu walking a dog



Edit prompt or view more images↓

Zero-Shot Text-to-Image Generation (DALLE) [Ramesh et.al. 2021]



(1) Contrastive pre-training



CLIP: Connecting Text and Images [Radford et.al. 2021]



Semantic Segmentations



Image classifications

an illustration of a baby daikon radish in a tutu walking a dog



Edit prompt or view more images↓

Zero-Shot Text-to-Image Generation (DALLE) [Ramesh et.al. 2021]

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Edit prompt or view more images↓

Zero-Shot Text-to-Image Generation (DALLE) [Ramesh et.al. 2021]



Image classifications

### Prior Work



GPV [Gupta et.al. 2021]



#### Perceiver-IO [Jaegle et.al. 2021]



OFA [Wang et.al. 2022]



### Prior Work



Pixel2seq [Chen et.al. 2021]



(a) **Stage I** training: we train the base model f, which is guided by the code produced by the *restricted oracle* model  $\Omega$ . The oracle has access to the ground-truth label, but is only allowed to communicate with f by passing a short discrete sequence, which we call a *guiding code*.



(b) Stage II training: we train a *language model* (LM) to output a *guiding code* by learning to mimic the oracle, but using only the image input.

UViM [Kolesnikov et.al. 2022]

## Vision has diverse output format



Image classifications



**Object detections** 



Semantic Segmentations



**Depth Estimation** 



**Pose Estimation** 

& MORE



### Vision has diverse output format









Image c

Convert diverse vision inputs/outputs into sequences.



Depth Estimation



**Pose Estimation** 





• NLP Task Pipeline





### • NLP Task Pipeline



• Vision Task Pipeline



### • NLP Task Pipeline



• Vision Task Pipeline



• Image Quantization Using VQ-VAE



• Image Quantization Using VQ-VAE



• Image Quantization Using VQ-VAE



• Task with 2D structured outputs.



• Task with 2D structured outputs.



What is the segmentation of " giraffe " ?



• Task with 1D structured outputs



• Task with 1D structured outputs

Sequence length:  $17 \times 3 = 51$  per person

Take bounding box as input and detect one person each time.

human pose estimation --> detection + single human pose estimation















#### Segment the cat

What is the color of the cat?



What is the depth map of the image?

Localize the cat





















### Model Details

- Follow T5 implementation with minimum modification.
- Use pretrained VQ-GAN from Imagenet.
- Use patch encoding for image inputs (no CNN involved)
- Absolute position encoding is important for vision injecting the position embedding at input
- Relative position encoding within text and image.

#### [5, 4, 4, 3, 2, 1, 0, 1, 2, 3, 4, 4, 5, 5],

Relative encoding for the text

 $[ [45, 44, 44, 43, 42, 41, 40, 41, 42, 43, 44, 44, 45, 45], \\ [37, 36, 36, 35, 34, 33, 32, 33, 34, 35, 36, 36, 37, 37], \\ [37, 36, 36, 35, 34, 33, 32, 33, 34, 35, 36, 36, 37, 37], \\ [29, 28, 28, 27, 26, 25, 24, 25, 26, 27, 28, 28, 29, 29], \\ [21, 20, 20, 19, 18, 17, 16, 17, 18, 19, 20, 20, 21, 21], \\ [13, 12, 12, 11, 10, 9, 8, 9, 10, 11, 12, 12, 13, 13], \\ [ 5, 4, 4, 3, 2, 1, 0, 1, 2, 3, 4, 4, 5, 5], \\ [13, 12, 12, 11, 10, 9, 8, 9, 10, 11, 12, 12, 13, 13], \\ [ 13, 12, 12, 11, 10, 9, 8, 9, 10, 11, 12, 12, 13, 13], \\ [ 13, 12, 12, 11, 10, 9, 8, 9, 10, 11, 12, 12, 13, 13], \\ [ 14, 20, 20, 19, 18, 17, 16, 17, 18, 19, 20, 20, 21, 21], \\ [ 29, 28, 28, 27, 26, 25, 24, 25, 26, 27, 28, 28, 29, 29], \\ [ 37, 36, 36, 35, 34, 33, 32, 33, 34, 35, 36, 36, 37, 37], \\ [ 37, 36, 36, 35, 34, 33, 32, 33, 34, 35, 36, 36, 37, 37], \\ [ 45, 44, 44, 43, 42, 41, 40, 41, 42, 43, 44, 44, 45, 45], \\ [ 45, 44, 44, 43, 42, 41, 40, 41, 42, 43, 44, 44, 45, 45] \\$ 

Relative encoding for the Image

### Objective

- Pre-Training
  - Text span denoising (15%):

An image of a <M> is lying on the <M>.

<M> dog <M> ground.

• Mask image denoising (75%):



- Multi-Task Training
  - Jointly train with 80 vision, vision language and language datasets/sets.

Tasks



### Tasks

	Example		Siz	ze			Input M	odalities	8	Output Modalities					
	Source	Datasets	Size	Percent	Rate	Text	Image	Sparse	Dense	Text	Image	Sparse	Dense		
Image Synthesis		14	56m	43.0	18.7	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	-	$\checkmark$	-	-		
Image Synthesis from Text	RedCaps	9	55m	41.9	16.7	$\checkmark$	-	-	-	-	$\checkmark$	-	-		
Image Inpainting	VG	3	1.2m	0.9	1.5	$\checkmark$	$\checkmark$	$\checkmark$	-	-	$\checkmark$	-	-		
Image Synthesis from Seg.	LVIS	2	220k	0.2	0.6	$\checkmark$	-	-	$\checkmark$	-	$\checkmark$	-	-		
Sparse Labelling		10	8.2m	6.3	12.5	$\checkmark$	$\checkmark$	$\checkmark$	-	-	-	$\checkmark$	-		
Object Detection	<b>Open Images</b>	3	1.9m	1.5	3.6	-	$\checkmark$	-	-	-	-	$\checkmark$	-		
Object Localization	VG	3	6m	4.6	7.1	$\checkmark$	$\checkmark$	-	-	-	-	$\checkmark$	-		
Keypoint Estimation	COCO	1	140k	0.1	0.7	-	$\checkmark$	$\checkmark$	-	-	-	$\checkmark$	-		
Referring Expression	RefCoco	3	130k	0.1	1.1	$\checkmark$	$\checkmark$	-	-	-	-	$\checkmark$	-		
Dense Labelling		6	2.4m	1.8	6.2	$\checkmark$	$\checkmark$	-	-	-	-	-	$\checkmark$		
Depth Estimation	NYU Depth	1	48k	0.1	0.4	-	$\checkmark$	-	-	-	-	-	$\checkmark$		
Surface Normal Estimation	Framenet	2	210k	0.2	1.1	-	$\checkmark$	-	-	-	-	-	$\checkmark$		
Object Segmentation	LVIS	3	2.1m	1.6	4.7	$\checkmark$	$\checkmark$	-	-	-	-	-	$\checkmark$		
Image Classification		9	22m	16.8	12.5	-	$\checkmark$	$\checkmark$	-	$\checkmark$	-	-	-		
Image Classification	ImageNet	6	16m	12.2	8.1	$\checkmark$	$\checkmark$	-	-	$\checkmark$	-	-	-		
Object Categorization	COCO	3	6m	4.6	4.4	-	$\checkmark$	$\checkmark$	-	$\checkmark$	-	-	-		
Image Captioning		7	31m	23.7	12.5	-	$\checkmark$	$\checkmark$	-	$\checkmark$	-	-	-		
Webly Supervised Captioning	CC12M	3	26m	19.7	8.8	-	$\checkmark$	-	-	$\checkmark$	-	-	-		
Supervised Captioning	VizWiz	3	1.4m	1.1	1.7	-	$\checkmark$	-	-	$\checkmark$	-	-	-		
Region Captioning	VG	1	3.8m	2.9	2.0	-	$\checkmark$	$\checkmark$	-	$\checkmark$	-	-	-		
Vision & Language		16	4m	3.0	12.5	$\checkmark$	$\checkmark$	$\checkmark$	-	$\checkmark$	-	-	$\checkmark$		
Visual Question Answering	VQA 2.0	13	3.3m	2.5	10.4	$\checkmark$	$\checkmark$	$\checkmark$	-	$\checkmark$	-	-	-		
Relationship Detection	VG	2	640k	0.5	1.9	-	$\checkmark$	$\checkmark$	-	$\checkmark$	-	-	-		
Grounded VQA	VizWiz	1	6.5k	0.1	0.1	$\checkmark$	$\checkmark$	-	-	$\checkmark$	-	-	$\checkmark$		
NLP		31	7.1m	5.4	12.5	$\checkmark$	-	-	-	$\checkmark$	-	-	-		
Text Classification	MNLI	17	1.6m	1.2	4.8	$\checkmark$	-	-	-	$\checkmark$	-	-	-		
Question Answering	SQuAD	13	1.7m	1.3	5.2	$\checkmark$	-	-	-	$\checkmark$	-	-	-		
Text Summarization	Gigaword	1	3.8m	2.9	2.5	$\checkmark$	-	-	-	$\checkmark$	-	-	-		
Language Modelling		2	-	-	12.5	$\checkmark$	-	-	-	$\checkmark$	-	-	-		
Masked Language Modelling	<i>C4</i>	2	-	-	12.5	$\checkmark$	-	-	-	$\checkmark$	-	-	-		
All Tasks		95	130m	100	100	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$		

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### Tasks Distribution



### Tasks Distribution



### Evaluation





Benchmark

General Robust Image Task



GRIT: General Robust Image Task Benchmark [Gupta et.al. 2022]

# GRIT requires diverse skills

Task	Input Image	Input Query / Options	Output
Categorization		[open_images_categories]	drill
Localization		kitchen & dining room table	
Visual Question Answering		Does this sofa have armrests?	yes
Referring Expressions		man on end black suit	
Segmentation		dolphin	
Pose Keypoints		person	
Surface Normals			

	Inputs	Outputs
$ \rightarrow $	Bounding box	Class name
$\Rightarrow$	Class name	Bounding box
	Questions	Answers
$\Box$	Phrases	Bounding box
$\Rightarrow$	Class name	Segmentation
$ \rightarrow $		Joints + visibility
		Surface normal
		Ai2

# GRIT requires diverse skills

Task	Input Image	Input Query / Options	Output	
Categorization		[open_images_categories]	drill	
Localization		kitchen & dining room table		$\Rightarrow$
Visual Question Answering		Does this sofa have armrests?	yes	$\Box$
Referring Expressions		man on end black suit		$\Box$
Segmentation	The second secon	dolphin		$\Box$
Pose Keypoints		person		
Surface Normals				

Inputs	Outputs
Bounding box	Class name
Class name	Bounding box
Questions	Answers
Phrases	Bounding box
Class name	Segmentation
	Joints + visibility
	Surface normal

# GRIT requires diverse skills

Task	Input Image	Input Query / Options	Output	
Categorization		[open_images_categories]	drill	$\Box \rangle$
Localization		kitchen & dining room table		$\Box \rangle$
Visual Question Answering		Does this sofa have armrests?	yes	$\Box \rangle$
Referring Expressions		man on end black suit		$\square \rangle$
Segmentation	The second	dolphin		$\square \hspace{-0.5ex} \searrow$
Pose Keypoints		person		
Surface Normals				

Inputs	Outputs
Bounding box	Class name
Class name	Bounding box
Questions	Answers
Phrases	Bounding box
Class name	Segmentation
	Joints + visibility
	Surface normal



		Categor	ization	Localiz	ation	VQ	A	Refe	xp	Segmen	tation	Keypo	oint	Norm	nal	All	l
		ablation	test	ablation	test	ablation	test	ablation	test	ablation	test	ablation	test	ablation	test	ablation	test
0 NLL-Ang	MF [3]	-	-	-	-	-	-	-	-	-	-	-	-	49.6	50.5	7.2	7.1
1 Mask R-C	NN [29]	-	-	44.7	45.1	-	-	-	-	26.2	26.2	70.8	70.6	-	-	20.2	20.3
2 GPV-1 [26	5]	33.2	33.2	42.8	42.7	50.6	49.8	25.8	26.8	-	-	-	-	-	-	21.8	21.8
3 CLIP [56]		48.1	-	-	-	-	-	-	-	-	-	-	-	-	-	6.9	-
4 <b>OFA</b> LARGE	[73]	22.6	-	-	-	72.4	-	61.7	-	-	-	-	-	-	-	22.4	-
5 GPV-2 [36	5]	54.7	55.1	53.6	53.6	63.5	63.2	51.5	52.1	-	-	-	-	-	-	31.9	32.0
6 UNIFIED-	$IO_{\text{SMALL}}$	42.6	-	50.4	-	52.9	-	51.1	-	40.7	-	46.5	-	33.5	-	45.4	-
7 UNIFIED-	$IO_{BASE}$	53.1	-	59.7	-	63.0	-	68.3	-	49.3	-	60.2	-	37.5	-	55.9	-
8 UNIFIED-	$IO_{LARGE}$	57.0	-	64.2	-	67.4	-	74.1	-	54.0	-	67.6	-	40.2	-	57.0	-
9 UNIFIED-	$IO_{XL}$	61.7	60.8	67.0	67.1	74.5	74.5	78.6	78.9	56.3	56.5	68.1	67.7	45.0	44.3	64.5	64.3

		Categorization		Localization		VQA		Refexp		Segmentation		Keypoint		Normal		All	
		ablation	test	ablation	test	ablation	test	ablation	test	ablation	test	ablation	test	ablation	test	ablation	test
0 1	NLL-AngMF [3]	-	-	-	-	-	-	-	-	-	-	-	-	49.6	50.5	7.2	7.1
1 l	Mask R-CNN [29]	-	-	44.7	45.1	-	-	-	-	26.2	26.2	70.8	70.6	-	-	20.2	20.3
2 (	GPV-1 [26]	33.2	33.2	42.8	42.7	50.6	49.8	25.8	26.8	-	-	-	-	-	-	21.8	21.8
3 (	CLIP [56]	48.1	-	-	-	-	-	-	-	-	-	-	-	-	-	6.9	-
4 (	OFA <sub>LARGE</sub> [73]	22.6	-	-	-	72.4	-	61.7	-	-	-	-	-	-	-	22.4	-
5 (	GPV-2 [36]	54.7	55.1	53.6	53.6	63.5	63.2	51.5	52.1	-	-	-	-	-	-	31.9	32.0
61	UNIFIED-IO $_{\text{SMALL}}$	42.6	-	50.4	-	52.9	-	51.1	-	40.7	-	46.5	-	33.5	-	45.4	÷
71	UNIFIED-IO <sub>BASE</sub>	53.1	-	59.7	-	63.0	-	68.3	-	49.3	-	60.2	-	37.5	-	55.9	ł
8 1	UNIFIED-IO <sub>LARGE</sub>	57.0	-	64.2	-	67.4	-	74.1	-	54.0	-	67.6	-	40.2	-	57.0	
91	UNIFIED- $IO_{XL}$	61.7	60.8	67.0	67.1	74.5	74.5	78.6	78.9	56.3	56.5	68.1	67.7	45.0	44.3	64.5	64.3

		Categor	ization	Localiz	ation	VQA	4	Refe	хр	Segmen	tation	Keypo	oint	Norm	nal	All	
		ablation	test	ablation	test	ablation	test	ablation	test	ablation	test	ablation	test	ablation	test	ablation	test
0	NLL-AngMF [3]	-	-	-	-	-	-	-	-	-	-	_	-	49.6	50.5	7.2	7.1
1	Mask R-CNN [29]	-	-	44.7	45.1	-	-	-	-	26.2	26.2	70.8	70.6	-	-	20.2	20.3
2	GPV-1 [26]	33.2	33.2	42.8	42.7	50.6	49.8	25.8	26.8	-	-	-	-	-	-	21.8	21.8
3	CLIP [56]	48.1	-	-	-	-	-	-	-	-	-	-	-	-	-	6.9	-
4	OFA <sub>LARGE</sub> [73]	22.6	-	-	-	72.4	-	61.7	-	-	-	-	-	-	-	22.4	-
5	GPV-2 [36]	54.7	55.1	53.6	53.6	63.5	63.2	51.5	52.1	-	-	-	-	-	-	31.9	32.0
6	$Unified\text{-}IO_{\text{SMALL}}$	42.6	-	50.4	-	52.9	-	51.1	-	40.7	-	46.5	-	33.5	-	45.4	-
7	UNIFIED- $IO_{BASE}$	53.1	-	59.7	-	63.0	-	68.3	-	49.3	-	60.2	-	37.5	-	55.9	-
8	$UNIFIED\text{-}IO_{\text{LARGE}}$	57.0	-	64.2	-	67.4	-	74.1	-	54.0	-	67.6	-	40.2	-	57.0	-
9	$UNIFIEDIO_{\text{XL}}$	61.7	60.8	67.0	67.1	74.5	74.5	78.6	<b>78.9</b>	56.3	56.5	68.1	67.7	45.0	44.3	64.5	64.3

	Catego	rization	Localiz	ation	VQ	A	Refe	хр	Segmen	tation	Кеурс	oint	Norm	nal	All	
	ablation	test	ablation	test	ablation	test	ablation	test	ablation	test	ablation	test	ablation	test	ablation	test
0 NLL-AngMF [3	8] -	-	-	-	-	-	-	-	-	-	-	-	49.6	50.5	7.2	7.1
1 Mask R-CNN [2	29] -	-	44.7	45.1	-	-	-	-	26.2	26.2	70.8	70.6	-	-	20.2	20.3
2 GPV-1 [26]	33.2	33.2	42.8	42.7	50.6	49.8	25.8	26.8	-	-	-	-	-	-	21.8	21.8
3 CLIP [56]	48.1	-	-	-	-	-	-	-	-	-	-	-	-	-	6.9	-
4 <b>OFA</b> <sub>LARGE</sub> [73]	22.6	-	-	-	72.4	-	61.7	-	-	-	-	-	-	-	22.4	-
5 GPV-2 [36]	54.7	55.1	53.6	53.6	63.5	63.2	51.5	52.1	-	-	-	-	-	-	31.9	32.0
6 UNIFIED-IO <sub>SMA</sub>	ALL 42.6	-	50.4	-	52.9	-	51.1	-	40.7	-	46.5	-	33.5	-	45.4	-
7 UNIFIED-IO <sub>BAS</sub>	<sub>E</sub> 53.1	-	59.7	-	63.0	-	68.3	-	49.3	-	60.2	-	37.5	-	55.9	-
8 UNIFIED-IO <sub>LAR</sub>	RGE 57.0	-	64.2	-	67.4	-	74.1	-	54.0	-	67.6		40.2	-	57.0	-
9 UNIFIED-IO <sub>XL</sub>	61.7	60.8	67.0	67.1	74.5	74.5	78.6	78.9	56.3	56.5	68.1	67.7	45.0	44.3	64.5	64.3

		Categor	ization	Localiz	ation	VQA	4	Refe	хр	Segmen	tation	Keypo	oint	Norm	nal	All	l
		ablation	test	ablation	test	ablation	test	ablation	test	ablation	test	ablation	test	ablation	test	ablation	test
0 NLL-Angl	MF [3]	-	-	-	-	-	-	-	-	-	-	-	-	49.6	50.5	7.2	7.1
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3 CLIP [56]		48.1	-	-	-	-	-	-	-	-	-	-	-	-	-	6.9	-
4 $OFA_{LARGE}$	[73]	22.6	-	-	-	72.4	-	61.7	-	-	-	-	-	-	+	22.4	-
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6 UNIFIED-	[O <sub>SMALL</sub>	42.6	-	50.4	-	52.9	-	51.1	-	40.7	-	46.5	-	33.5	-	45.4	-
7 UNIFIED-	$O_{BASE}$	53.1	-	59.7	-	63.0	-	68.3	-	49.3	-	60.2	-	37.5	+	55.9	-
8 UNIFIED-	<b>IO</b> LARGE	57.0	-	64.2	-	67.4	-	74.1	-	54.0	-	67.6	-	40.2		57.0	-
9 UNIFIED-	IO <sub>XL</sub>	61.7	60.8	67.0	67.1	74.5	74.5	78.6	78.9	56.3	56.5	68.1	67.7	45.0	44.3	64.5	64.3

	Categorization		Localization		VQA		Refexp		Segmentation		Keypoint		Normal		All	
	ablation	test	ablation	test	ablation	test	ablation	test	ablation	test	ablation	test	ablation	test	ablation	test
0 NLL-AngMF [3]	-	-	-	-	-	-	-	-	-	-	-	-	49.6	50.5	7.2	7.1
1 Mask R-CNN [29]	] -	-	44.7	45.1	-	-	-	-	26.2	26.2	70.8	70.6	-	-	20.2	20.3
2 GPV-1 [26]	33.2	33.2	42.8	42.7	50.6	49.8	25.8	26.8	-	-	-	-	-	-	21.8	21.8
3 CLIP [56]	48.1	-	-	-	-	-	-	-	-	-	-	-	-	-	6.9	-
4 OFA <sub>LARGE</sub> [73]	22.6	-	-	-	72.4	-	61.7	-	-	-	-	-	-	-	22.4	-
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8 UNIFIED- $IO_{LARGE}$	57.0	-	64.2	-	67.4	-	74.1	-	54.0	-	67.6	-	40.2	-	57.0	-
9 UNIFIED-IO <sub>XL</sub>	61.7	60.8	67.0	67.1	74.5	74.5	78.6	78.9	56.3	56.5	68.1	67.7	45.0	44.3	64.5	64.3

	NYUV2	ImageNet	Place365	VQAV2	okvqA	A-OkVQA	<sup>Vizwiz</sup> QA	VizWizGround	Swig	SNLI-VE	VisComet	Nocaps	coco	coco	MRPC	gloog	SciTail
Split	val	val	val	test-dev	test	test	test-dev	test-std	test	val	val	val	val	test	val	val	test
Metric	RMSE	Acc.	Acc.	Acc.	Acc.	Acc.	Acc.	IOU	Acc.	Acc.	CIDEr	CIDEr	CIDEr	CIDEr	F1	Acc	Acc
Unified SOTA	UViM	170	-	-	Flamingo	5	Flamingo	-	-	5	7	-		5	T5	PaLM	-
	0.467	-	-	-	57.8	-	49.8	-	-	-	-	-	-	-	92.20	92.2	-
UNIFIED-IO SMALL	0.649	42.8	38.2	57.7	31.0	24.3	42.4	35.5	17.3	76.5	-	45.1	80.1	-	84.9	65.9	87.4
UNIFIED-IO <sub>BASE</sub>	0.469	63.3	43.2	61.8	37.8	28.5	45.8	50.0	29.7	85.6	-	66.9	104.0	-	87.9	70.8	90.8
$UNIFIED\text{-IO}_{\text{LARGE}}$	0.402	71.8	50.5	67.8	42.7	33.4	47.7	54.7	40.4	86.1	-	87.2	117.5	-	87.5	73.1	93.1
$U\text{NIFIED-IO}_{\text{XL}}$	0.385	79.1	53.2	77.9	54.0	45.2	57.4	65.0	49.8	91.1	21.2	100.0	126.8	122.3	89.2	79.7	95.7
Single or fine-	BinsFormer	CoCa	MAE	CoCa	KAT	GPV2	Flamingo	MAC-Caps	JSL	OFA	SVT	CoCa	2	OFA	Turing NLR	ST-MOE	DeBERTa
tuned SOTA	0.330	91.00	60.3	82.3	54.4	38.1	65.7	27.3	39.6	91.0	18.3	122.4	-	145.3	93.8	92.4	97.7

Unified IO on other tasks

### Visualization – image synthesis



## Visualization – image synthesis



## Visualization – image synthesis



## Visualization – sparse labeling



## Visualization – sparse labeling

#### **OBJECT DETECTION**

Locate all objects in the image.



# Visualization – sparse labeling



### Visualization – dense labeling

#### DEPTH ESTIMATION

INPUT PREDICTION

What is the depth map of the image?

### Visualization – dense labeling

#### SURFACE NORMAL ESTIMATION

What is the surface normal of the image?



INPUT

### Visualization – dense labeling



### Referring expressions using different prompts

	Prompt	<b>Refexp Score</b>
0	Which region does the text "REFEXP " describe ?	78.9
1	Which region does the text "REFEXP" describe?	76.7
2	Which region matches the text "REFEXP "?	77.4
3	Locate the "REFEXP".	65.6
4	Which region can be described as "REFEXP "?	64.8
5	Locate the region described by "REFEXP".	43.2
6	Where is the "REFEXP"?	41.5
7	Where is the "REFEXP"?	0.1

### Summary

- Propose unified IO which is the first framework that can handle massive vision, vision language, and language tasks.
- We treat 2D image tasks as condition image generation tasks.
- We use pre-trained VQ-GAN to convert images into discrete sequences.
- We will release the Code + pre-trained model.

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- Al2 Reviz Team Great demo pages.



A new general-purpose model with **unprecedented breadth**, Unified-IO can perform a wide array of **visual** and **linguistic** tasks.



### https://unified-io.allenai.org

