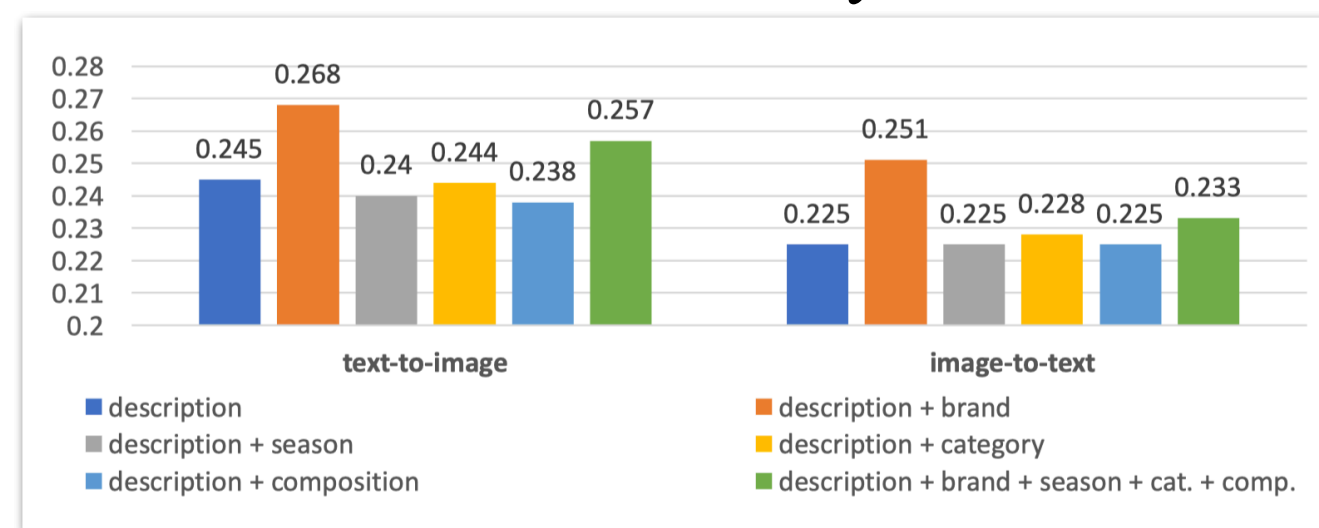


Motivations:

1) Words come up with special meanings in e-commerce.

General Domain	E-Commerce Domain
<p>Tankstelle Diesel</p> <p>Diesel</p> <p>"diesel" "fuel" "garage" "gas" "pump" "diesel" "fuel" "gasoline" "spirit" "car"</p>	<p>T-DIEGO-CUTY logo T-shirt</p> <p>Brand: diesel Composition: Cotton 100% Category: t-shirt Fit: relaxed Washing instruction: machine wash</p>

2) Meta data contributes unevenly in cross-modal retrieval.



CLIP in the Causal View:

Confounders $z = g(a, b)$, entity a takes the semantics b .
 X : text, Y : image.

Contrastive learning of CLIP:

$$P(Y|X) = \sum_z P(Y, z|X) = \sum_z P(Y|X, z) \underline{P(z|X)}$$

Interve X with do-calculus

mitigate bias towards commonsense in general domain

$$P(Y|do(X)) = \sum_z P(Y|X, z) \underline{P(z)}$$

e.g.

$X = "a T-shirt of golden goose"$

most of the likelihood is assigned to

$P(Y|X, z = g(\text{golden goose}, "animal"))$, as

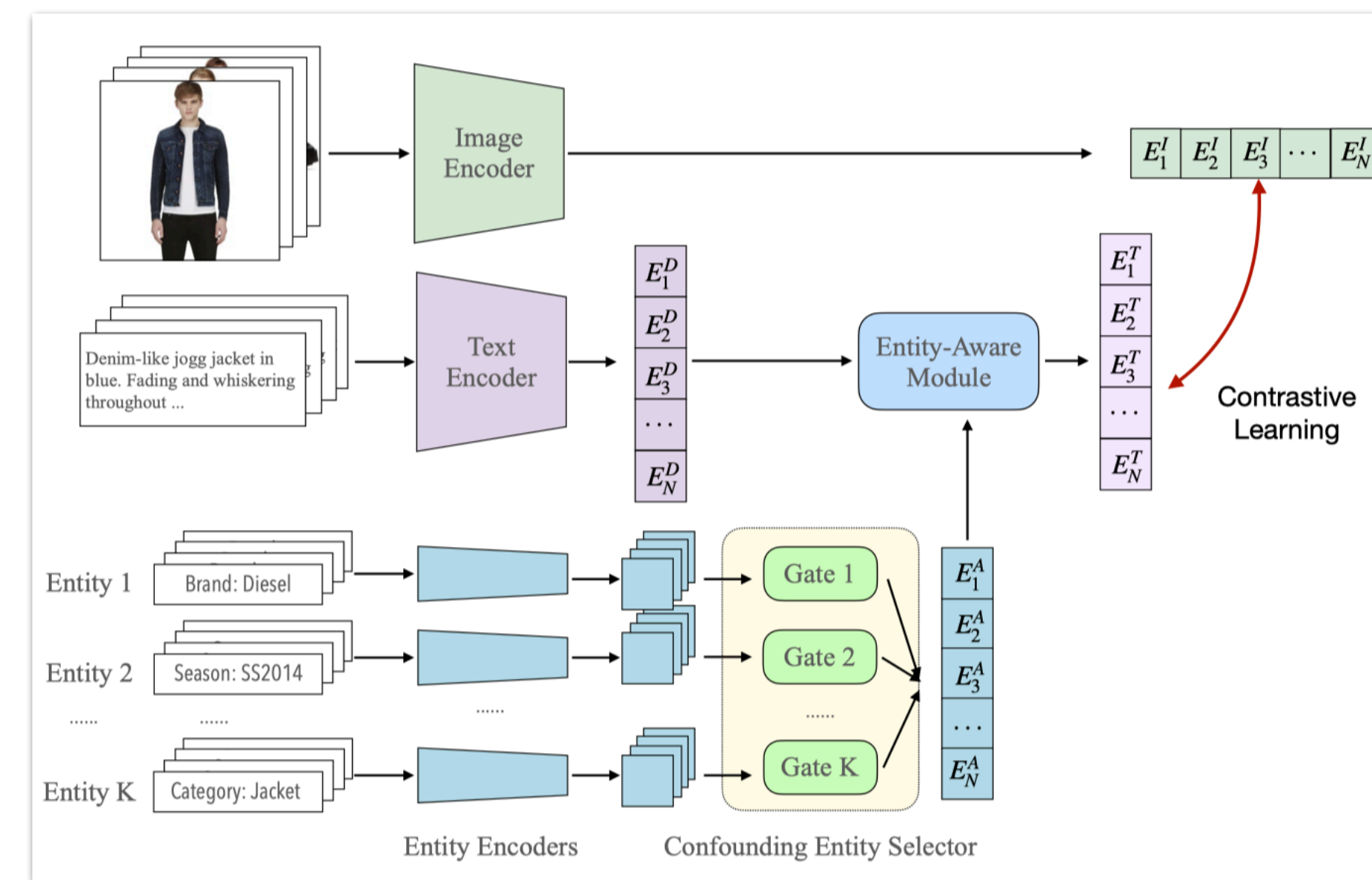
$P(z = g(\text{golden goose}, "animal") | X)$ is large in the general domain



Framework (EI-CLIP):

EA-Learner: explicitly capture each entity information

CE-Selector: select important entities



Results:

Fashion-Gen

	Image-to-text			Text-to-image			Sum R
	R@1	R@5	R@10	R@1	R@5	R@10	
①	9.4	24.5	33.5	10.7	26.8	35.8	141
②	22.5	49.5	62.0	24.5	51.1	63.6	273
③	23.3	51.5	64.6	25.7	53.9	66.5	285
④	25.2	52.6	64.8	28.2	56.6	68.4	296
⑤	25.7	54.5	66.8	28.4	57.1	69.4	302
↑	10.3%	5.8%	3.1%	10.5%	5.9%	4.4%	6.0%

Brand: Diesel. Long sleeve denim jacket in black. Fading, distressing, stitched detailing, and multicolor appliques throughout. Spread collar. Button closure at front. Flap pockets at chest. Seam pockets at waist. White logo embroidered at front hem. Adjustable buttoned tabs at back hem. Silver-tone hardware. Tonal stitching.



Contribution Summaries:

- 1) The pioneering work to tackle the challenges introduced by e-commerce special entities.
- 2) The first to formulate the entity-aware retrieval task in causal view.
- 3) We propose an Entity-aware Intervention-based contrastive learning framework (EI-CLIP), which achieves competitive performance on e-commerce benchmark dataset Fashion-Gen.

References:

Radford, Alec, et al. "Learning transferable visual models from natural language supervision." *IICML*, 2021.