Towards Multi-Modal Smart Information Retrieval Systems

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The Mercateo Group and its platforms

Facts and figures

Mercateo

mercateo

Procurement platform and Europe's leading B2B marketplace with > 100 million items

Growth

Revenue of €343 million in 2020

9 % year-on-year growth

Unite



Neutral B2B network with pre-integrated Mercateo Shop

Partners

> 100,000 active business customers > 700 suppliers > 50 system partners

Made in Germany

Founded in 2000 in Munich

International

> 600 employees in 14 subsidiaries in Europe



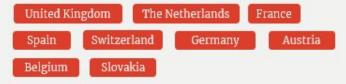
Unite in Europe

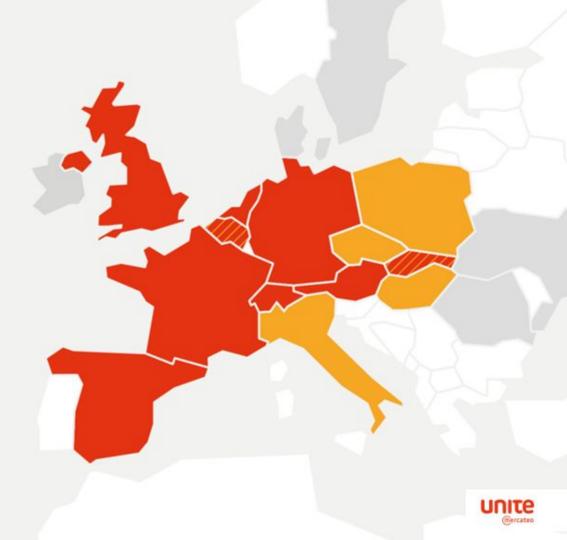
Availability and new countries (2021)

Unite available via Mercateo:



Unite available via other e-procurement partners:





The Problem

Motivation



Motivation





Motivation



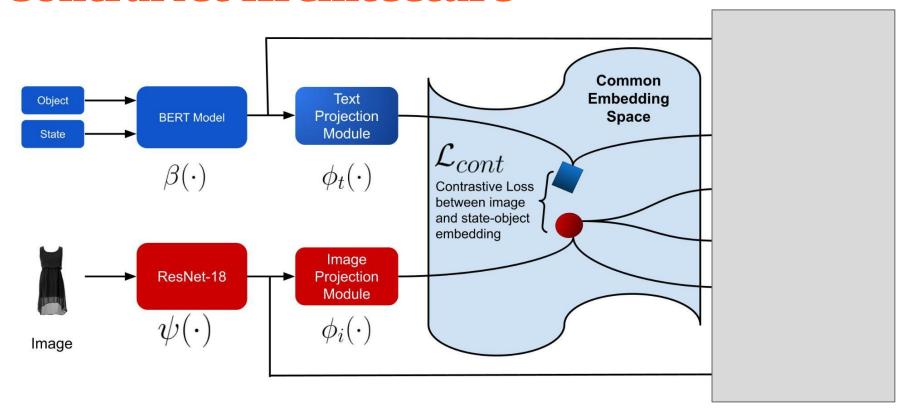
A potential scenario of a smart IR system, "enabling the customer" to express their mind better.

Tasks

Models which learn good state-object representations

- (1) Differentiate between different states of an object and can recognise even unseen combinations of state-object. (CZSL Task)
- (2) Retrieve images based on multi-modal (image-text) query, where the text describes the changes sought by the user in the query image. (Retrieval Task)

ContraNet Architecture



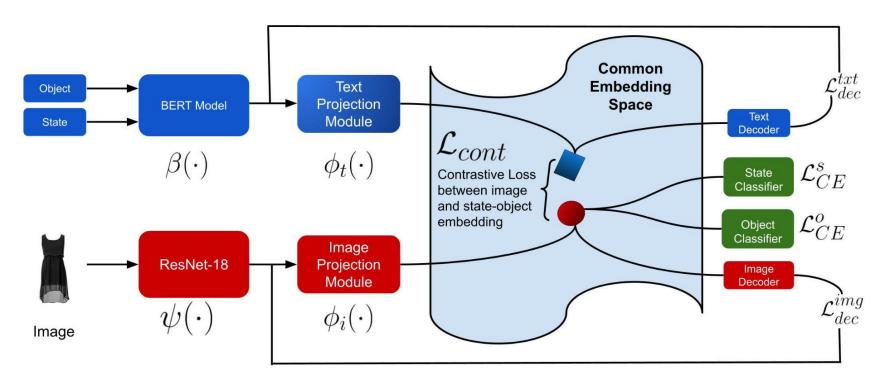
$$\Delta_{t} = \left\| \int_{j=1}^{B} \phi_{t}(\beta(t_{j})), \right\|$$

$$\Delta_{i} = \left\| \int_{j=1}^{B} \phi_{i}(\psi(x_{j})), \right\|$$

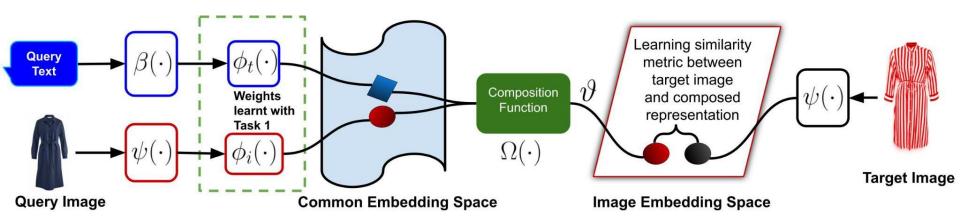
$$\mathcal{E} = \Delta_{t} * \Delta_{i}^{T},$$

$$\mathcal{L}_{cont} = \frac{1}{2B} \sum_{j=1}^{B} -\log \left\{ \frac{\exp\{\mathcal{E}_{jj}\}}{\sum_{p=1}^{B} \exp\{\mathcal{E}_{jp}\}} \right\}$$
$$-\log \left\{ \frac{\exp\{\mathcal{E}_{jj}\}}{\sum_{p=1}^{B} \exp\{\mathcal{E}_{pj}\}} \right\},$$

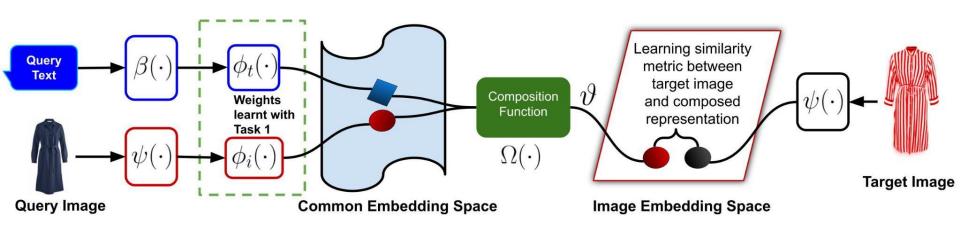
Task # 1 (CZSL Task)



Task#2 (Retrieval Task)



Task#2 (Retrieval Task)



$$\mathcal{L}_{trip} = \frac{1}{MB} \sum_{j=1}^{B} \sum_{m=1}^{M} \log \left\{ 1 + \exp\{\kappa(\vartheta_j, \psi(\tilde{y}_{j,m})) - \kappa(\vartheta_j, \psi(y_j))\} \right\}$$

Results

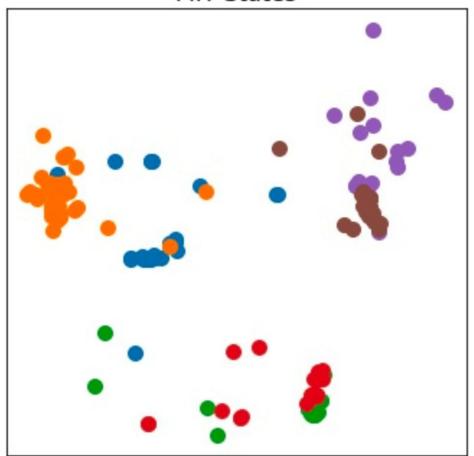
Task # 1 [Traditional CZSL Split]

Method	N	/IIT-State	es .	UT-Zappos			
	Top-1	Top-2	Top-3	Top-1	Top-2	Top-3	
AnalogousAttr	1.4	-	-	18.3	-	-	
LabelEmbed	13.4	17.6	22.4	25.8	39.8	52.4	
Red Wine	13.1	21.2	27.6	40.3	52.8	67.1	
AoP	14.2	19.6	25.1	46.2	56.6	69.2	
TAFE-Net	16.4	26.4	33.0	33.2	45.8	57.3	
SymNet	19.9	28.2	33.8	52.1	67.8	76.0	
ContraNet	22.1	33.7	38.2	54.6	73.1	80.4	
- without \mathcal{L}_{cont}	14.7	18.8	24.6	39.9	50.2	62.4	
- without \mathcal{L}_{dec}	<u>20.4</u>	<u>31.2</u>	<u>35.9</u>	<u>53.3</u>	<u>69.7</u>	<u>78.6</u>	

Task # 1 [GCZSL Split]

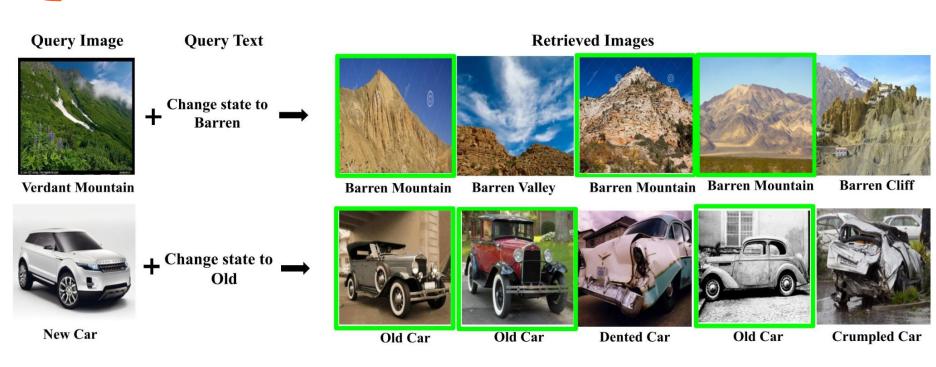
Method	MIT-States					UT-Zappos						
	Attribute	Object	Seen	Unseen	HM	AUC	Attribute	Object	Seen	Unseen	HM	AUC
AoP	21.1	23.6	14.3	17.4	9.9	1.6	38.9	69.9	59.8	54.2	40.8	25.9
Red Wine	22.7	25.1	20.7	17.9	11.6	2.4	40.6	69.1	53.6	52.1	41.3	26.1
LabelEmbed	23.5	26.3	15.0	20.1	10.7	2.0	41.2	<u>69.3</u>	53.0	<u>61.9</u>	41.0	25.7
ComposeAE	23.8	26.4	21.4	22.6	14.9	2.7	41.9	68.8	57.2	58.9	44.2	29.2
TMN	23.3	26.5	20.2	20.1	13.0	2.9	40.8	69.2	58.7	60.0	45.0	29.3
SymNet	24.3	27.3	24.2	25.2	16.1	3.0	41.3	68.6	49.8	57.4	40.4	23.4
ContraNet	28.9	26.7	28.1	27.4	17.4	4.7	52.7	68.1	60.7	62.5	48.9	34.7
- without \mathcal{L}_{cont}	22.9	<u>27.1</u>	14.8	18.6	9.7	1.9	42.4	69.2	54.9	52.6	42.9	27.4
- without \mathcal{L}_{dec}	28.2	26.5	<u>27.5</u>	<u>26.8</u>	<u>17.2</u>	<u>3.9</u>	<u>51.4</u>	66.7	58.4	60.8	<u>47.2</u>	<u>33.1</u>

MIT States



- dark_room
- filled_room
- straight_highway
- winding_highway
- diced_apple
- peeled_apple

Qualitative Results: MIT-States



Qualitative Results: Fashion 200k





Gray Slub Knit Skirt

Replace with **Blue Vintage** 1980s Denim



Blue Vintage 1980s **Denim Skirt**



Barbara Denim Mini Skirt



Blue Denim Pencil Skirt



Blue Vintage 1980s **Denim Skirt**



Gray Floral Print Blouse



Blue Denim Debra Skirt

Task # 2: MIT-States and F200k

Method	R@1	R@5	R@10	R@1	R@10	R@50
Raw Image features only	3.3	12.8	20.9	3.5	22.7	43.7
Raw Text features only	7.4	21.5	32.7	1.0	12.3	21.8
Concatenation [Image,Text]	11.8	30.8	42.1	11.9	39.7	62.6
Show and Tell	11.9	31.0	42.0	12.3	40.2	61.8
AoP	8.8	27.3	39.1	12.2	40.0	61.7
Relationship	12.3	31.9	42.9	13.0	40.5	62.4
FiLM	10.1	27.7	38.3	12.9	39.5	61.9
SymNet	11.2	29.5	41.4	11.7	38.6	60.4
TIRG	12.2	31.9	43.1	14.1	42.5	63.8
ComposeAE	<u>13.9</u>	<u>35.3</u>	<u>47.9</u>	22.8	<u>55.3</u>	<u>73.4</u>
ContraNet- Ω_{res}	14.5	40.7	51.4	24.0	58.4	79.2
ContraNet- Ω_{rot}	13.9	39.1	50.8	18.5	54.8	76.3
ContraNet- Ω_{mlp}	13.7	36.9	48.8	22.9	56.7	77.5
ContraNet- Ω_{res} without						
Common Embedding Space	12.0	31.2	42.9	17.8	50.6	71.1

Conclusion

- Novel ContraNet Approach for both CZSL and Retrieval tasks
- Text-aware image representations via Contrastive Learning
- ContraNet outperforms the SOTA methods by a huge margin on benchmark datasets

Thanks!