Kaleido-BERT: Vision-Language Pre-training on Fashion Domain

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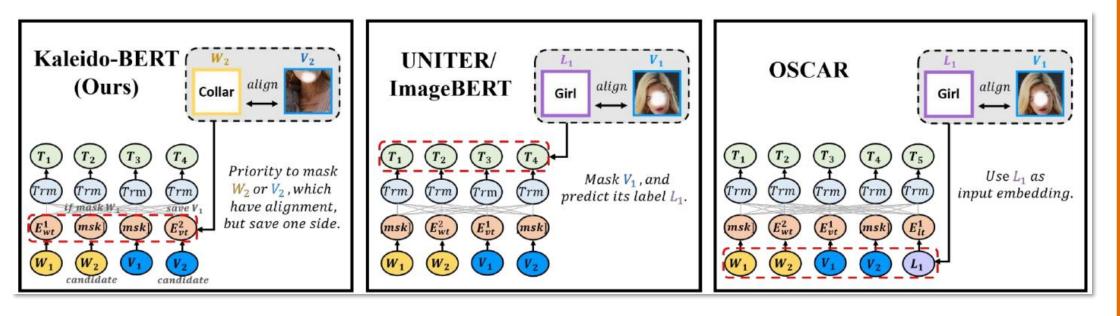






1.Introduction

1. Different utilization of alignment information in VL pre-training architectures.



2. Aligned Kaleido Patch Modeling (AKPM).



Problems:

Existing VL pre-training models:

- Difficult to extend on specific domain.
- Without intelligent masking strategy.
- Use scale-fixed patches or Rols as image inputs.
- Lack of image-level self-supervised pretext tasks.

Contributions:

Based on the above problems, we

- Presented a strong Kaleido-BERT model.
- Design a useful Alignment Guided Masking strategy.
- Rethinking 5 self-supervised pretext tasks in VL Pretraining process.

2. Vision-Languge Pre-training Model (Kaleido-BERT)

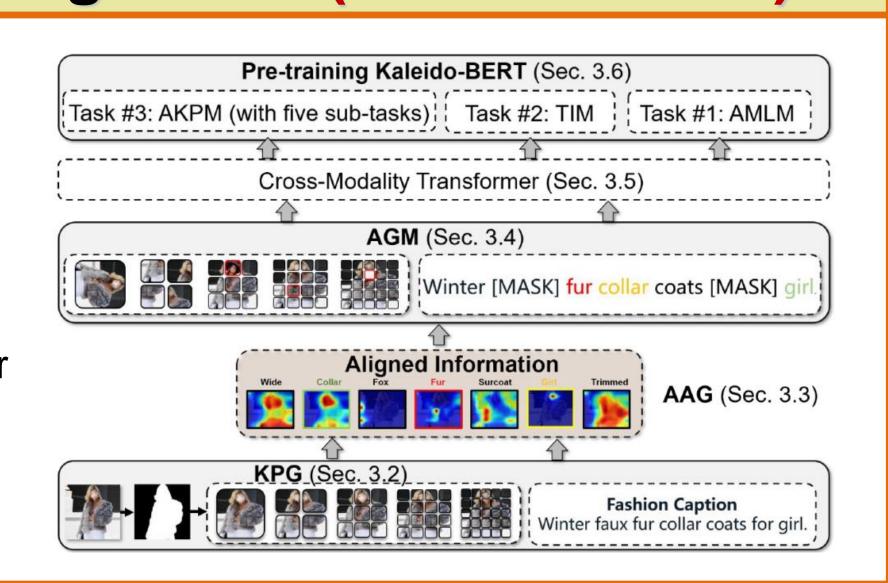
Kaleido-BERT, which consists:

KPG: Kaleido Patch Generator

AGM: Alignment Guided Masking

AAG: Attention-based Alignment Generator

AKPM: Aligned Kaleido Patch Modeling

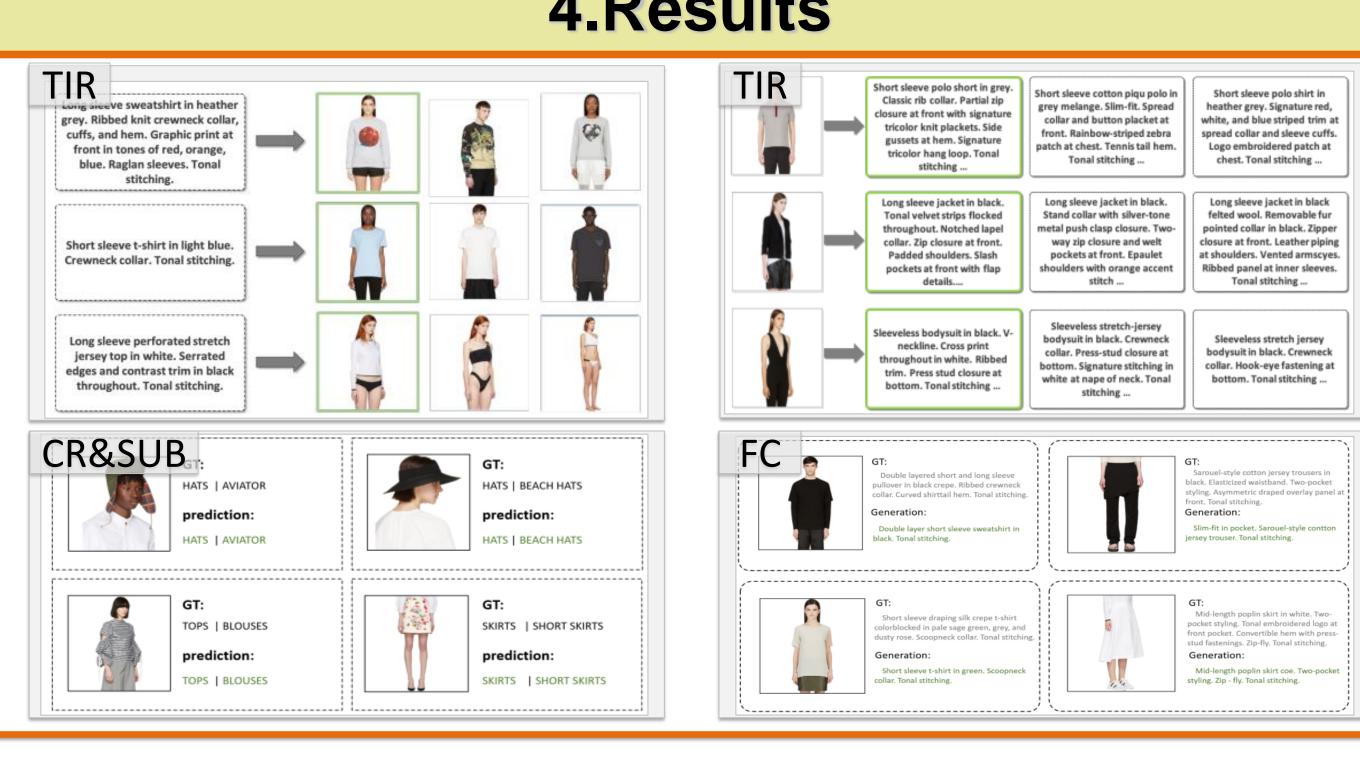


3. Experiments (Tasks & Ablations)

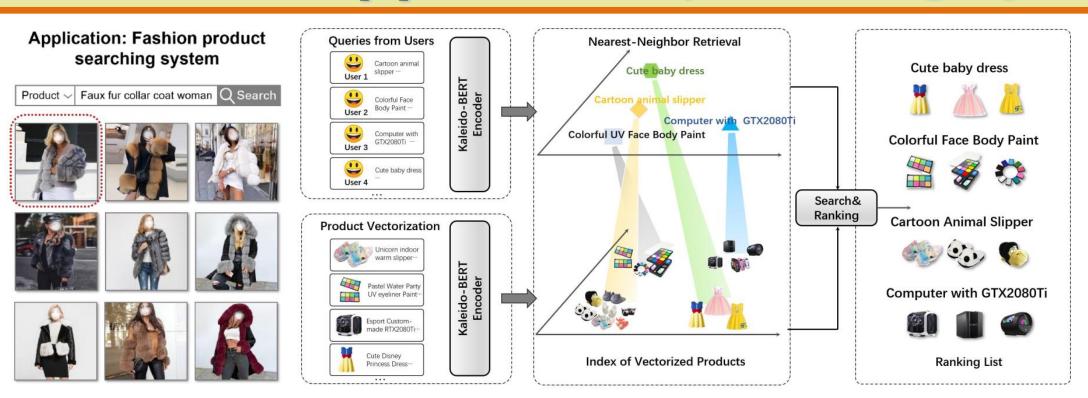
Tasks	ViLBERT [60]	VLBERT [45]	FashionBERT [21]	ImageBERT [55]	OSCAR [42]	Kaleido-BERT Ours	Tasks		FashionBERT [21]	ImageBERT [55]	OSCAR [42]	Kaleido-Bl Ours
Rank@1 ↑ Rank@5 ↑ Rank@10 ↑	20.97% 40.49% 48.21%	19.26% 39.90% 46.05%	23.96% 46.31% 52.12%	22.76% 41.89% 50.77%	23.39% 44.67%	27.99% (+4.030%)	3.CR ACC macro-		$ \begin{array}{c c} 91.25\% \\ 0.705 \\ \underline{85.27\%} \\ \underline{0.620} \end{array} $	$ 90.77\% \\ -0.699 \\ -\overline{80.11\%} \\ 0.575 $	91.79% -0.727 -84.23% 0.591	95.07% (+3 0.714 (-0 88.07% (+2 0.636 (+0
Rank@1 ↑ Rank@5 ↑ Rank@10 ↑	21.12% 37.23% 50.11%	22.63% 36.48% 48.52%	26.75% 46.48% 55.74%	24.78% 45.20% 55.90%	25.10% 49.14%	33.88% _(+7.130%) 60.60% _(+11.46%) 68.59% _(+11.91%)	4.FC Bleu-4 METE ROUG CIDEr	↑ OR ↑ E-L ↑	3.30 9.80 29.7 30.1	298.28 - - - -	307.82 <u>4.50</u> 10.9 30.1 30.7	318.14 5.70 (+1 12.8 (+1 32.9 (+2 32.6 (+1
Sum <i>R</i> ↑	218.13	212.84	251.36	241.30	<u>251.53</u>	319.52	Sum C.		72.9	-	76.2	84.0

Metrics			KPG	AGM		AKPM							
		Scale-fixed	Kaleido.	Kaleido.+SOD	Random	AGM	В	B+I	B+I∼II	B+I∼III	B+I∼IV	B+I∼V	B+V
1. Rank@1	↑	24.71	26.73(+8.2%)	27.99(+13.3%)	26.55	27.99(+5.4%)	25.37	25.07(-1.2%)	26.03(+2.6%)	26.88(+6.0%)	26.20(+3.3%)	27.99(+10.3%)	24.62(-2.9%)
1. Rank@5	\uparrow	50.05	54.55(+9.0%)	60.09(+20.1%)	55.13	60.09(+8.9%)	54.97	55.14(+0.3%)	56.31(+2.4%)	58.34(+6.1%)	59.13(+7.6%)	60.09(+9.3%)	53.78(-2.2%)
1. Rank@10	\uparrow	58.93	65.44(+11.0%)	68.37(+16.0%)	64.92	68.37(+5.3%)	62.13	62.90(+1.2%)	63.37(+2.0%)	67.79(+9.1%)	67.99(+9.4%)	68.37(+10.0%)	60.88(-2.0)%
2. Rank@1	\uparrow	30.17	32.19(+6.7%)	33.88(+12.0%)	32.14	33.88(+5.4%)	31.09	30.98(-0.4%)	32.22(+3.6%)	33.17(+6.7%)	33.80(+8.7%)	33.88(+9.0%)	30.77(-1.0%)
2. Rank@5	\uparrow	52.29	58.40(+11.7%)	60.60(+15.9%)	56.99	60.60(+6.3%)	57.35	57.44(+0.2%)	58.73(+2.4%)	58.55(+2.1%)	60.57(+5.6%)	60.60(+5.7%)	55.95(-2.4%)
2. Rank@10	↑	60.82	66.49(+9.3%)	68.59(+12.8%)	63.77	68.59(+7.6%)	64.79	65.65(+1.3%)	64.16(-1.0%)	67.92(+4.8%)	68.41(+5.6%)	68.09(+5.1%)	61.70(-4.8%)
Sum $\mathcal R$	↑	276.97	303.80(+9.7%)	319.52(+16.2%)	299.50	319.52(+6.7%)	295.70	297.18(+0.5%)	300.82(+1.7%)	312.65(+5.7%)	316.10(+6.9%)	319.02(+7.9%)	287.70(-2.7%)
3. ACC	1	93.44%	93.45%(+0.0%)	95.07%(+1.7%)	92.71%	95.07%(+2.5%)	90.94%	90.82%(-0.1%)	91.40%(+0.5%)	93.91%(+3.3%)	94.05%(+3.4%)	95.07%(+4.5%)	88.87(-2.3%)
3. macro- \mathcal{F}	↑	0.701	0.705(+0.6%)	0.714(+1.9%)	0.711	0.714(+0.4%)	0.690	0.692(+0.3%)	0.721(+4.5%)	0.713(+3.3%)	0.710(+2.9%)	0.714(+3.5%)	0.701(+1.4%)
4. ACC	\uparrow	86.89%	87.61%(+0.8%)	88.07%(+1.4%)	87.20%	88.07(+1.0%)	81.66%	81.25%(-0.5%)	84.44%(+3.4%)	86.49%(+5.9%)	88.53%(+8.4%)	88.07%(+7.9%)	81.64(+0.0%)
4. macro- \mathcal{F}	↑	0.630	0.634(+0.6%)	0.636(+1.0%)	0.633	0.636(+0.5%)	0.558	0.575(+3.0%)	0.596(+6.8%)	0.636(+14.0%)	0.633(+13.4%)	0.636(+14.0%)	0.596(+8.4%)
$\mathbf{Sum}~\mathcal{CLS}$	↑	313.43	314.96(+0.5%)	318.14(+1.5%)	314.31	318.14(+1.2%)	297.40	298.77(+0.4%)	307.54(+3.4%)	315.30(+6.0%)	316.88(+6.5%)	318.14(+7.0%)	300.21(+0.9%)
5. Bleu-4	1	4.9	5.2(+6.1%)	5.7(+16.3%)	5.3	5.7(+7.5%)	4.9	5.2(+6.1%)	5.2(+6.1%)	5.1(+4.1%)	5.6(+14.3%)	5.7(+16.3%)	5.3(+8.2%)
5. METEOR	↑	11.0	11.7(+6.4%)	12.8(+16.4%)	11.3	12.8(+13.3%)	11.6	11.6(+0.0%)	11.8(+1.7%)	12.6(+8.6%)	12.8(+10.3%)	12.8(+10.3%)	11.4(-1.7%)
5. ROUGE-L	\uparrow	29.8	31.5(+5.7%)	32.9(+10.4%)	30.3	32.9(+8.6%)	30.4	30.7(+1.0%)	30.8(+1.3%)	31.9(+4.9%)	32.7(+7.6%)	32.9(+8.2%)	30.6(+0.7%)
5. CIDEr	↑	30.9	31.3(+1.3%)	32.6(+5.5%)	31.7	32.6(+2.8%)	31.0	31.5(+1.6%)	31.4(+1.3%)	32.0(+3.2%)	32.3(+4.2%)	32.6(+5.2%)	31.3(+1.0%)
Sum \mathcal{CAP}	↑	76.6	79.7(+4.0%)	84.0(+9.7%)	78.6	84.0(+6.9%)	77.9	79.0(+1.4%)	79.2(+1.7%)	81.6(+4.7%)	83.4(+7.1%)	84.0(+7.8%)	78.6(+0.9%)

4.Results







Cross-modality Semantic Searching at Alibaba.com.

6.More Information

https://dpfan.net/Kaleido-BERT

https://github.com/mczhuge/Kaleido-BERT

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