





V2P: Vision-to-Prompt based Multi-Modal Product Summary Generation

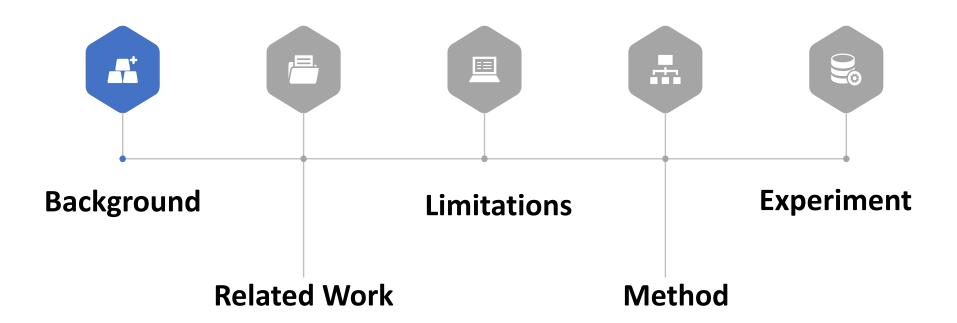
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Background

Online shopping platform always shows various information about the product.



https://www.oberlo.com/blog/online-shopping-statistics



Background

Multi-Modal Product Summarization Generation (MPSG)

Product Title 智能球拳电饭锅

苏泊尔 电饭煲 智能球釜电饭锅 电磁加热电饭煲 (SUPOR electric rice cooker intelligent spherical kettle electric rice cooker electromagnetic heating electric rice cooker)

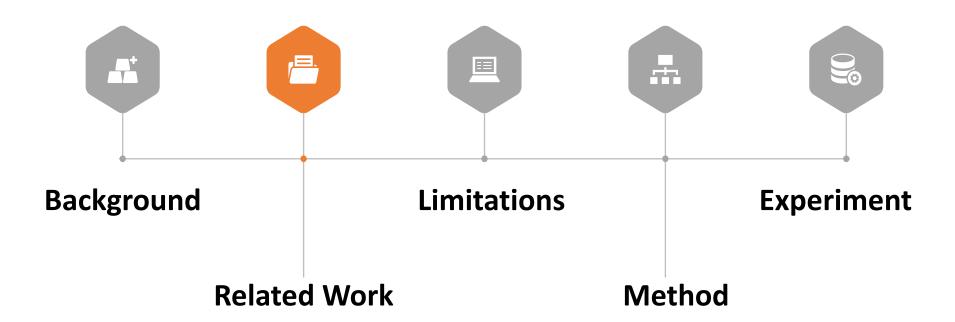
Product Details

Product Image



Product Summary

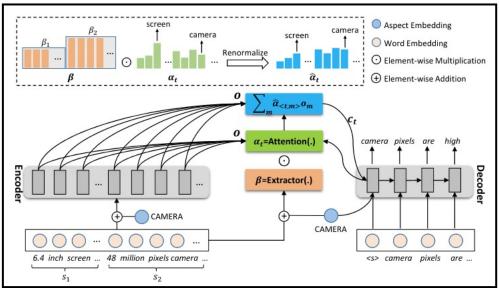
采用IH电磁加热方式,能模拟灶火使火力更大更均匀,实现对食物的360度环绕立体加热。采用钻石纹彩钢外壳,更加美观。拥有8种烹饪功能,能满足您的各种口味需求。顶部还配备了智能感温探针,使米饭更香甜软糯。(IH electromagnetic heating mode can simulate the stove fire, make the fire bigger and more uniform, and realize 360 degree surround three-dimensional heating of food. **Diamond pattern** color steel shell is adopted, which is more beautiful. With 8 cooking functions, it can meet your various taste needs. The top is also equipped with an intelligent temperature probe to make the rice more sweet and soft.)



Related Work

> Traditional Product Summarization





Jiahui Liang et al. CUSTOM: Aspect-Oriented Product Summarization for E-Commerce. NLPCC. 2021.

Related Work

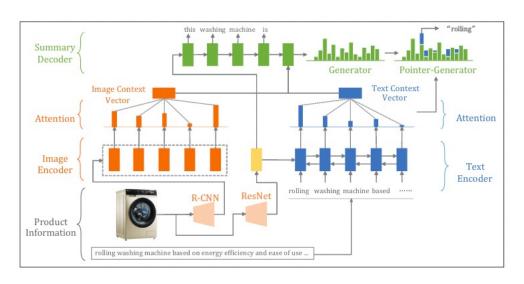
Multi-Modal Product Summarization



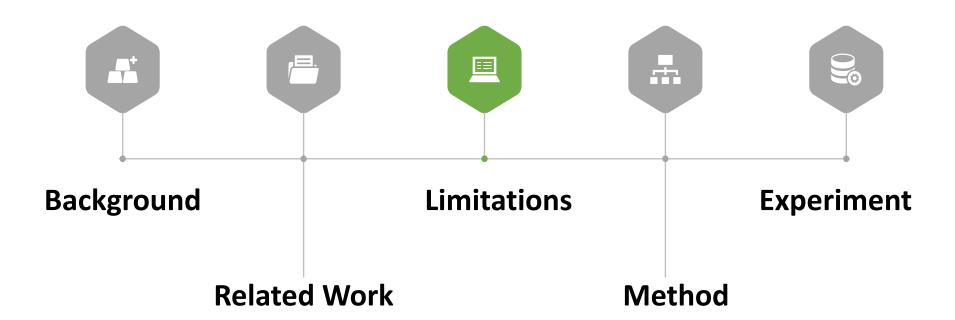
Product Summary

美的金色两门冰箱,搭配玻璃面板,外观时尚。立体风冷无霜技术,使冷气均匀分布。配备大冷冻空间,快速冻透食品,满足全家人需求。

(Midea golden double-door refrigerator with glass panel is fashionable. The technology of stereo air-cooled frost-free makes cold air disperse evenly. The refrigerator freezes food quickly, and the space is large enough to meet the requirement of the whole family.)



Haoran Li et al. Aspect-Aware Multimodal Summarization for Chinese E-Commerce Products. AAAI. 2020.

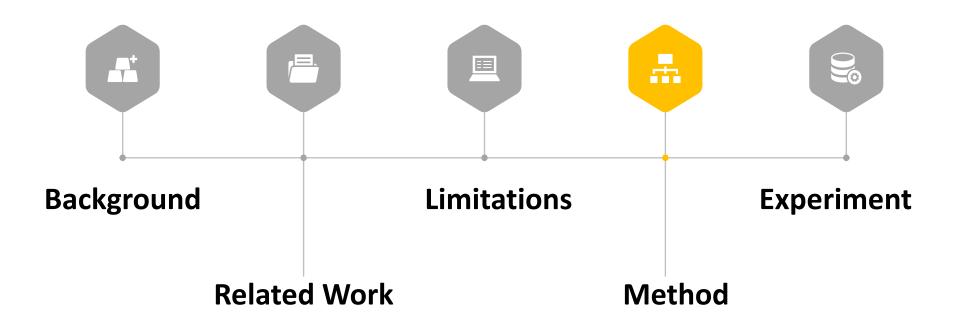


Limitations

> L1: Overlook the benefit of pre-training.

➤ L2: Lack the representation-level supervision.

> L3: Ignore the diversity of the seller-generated data.



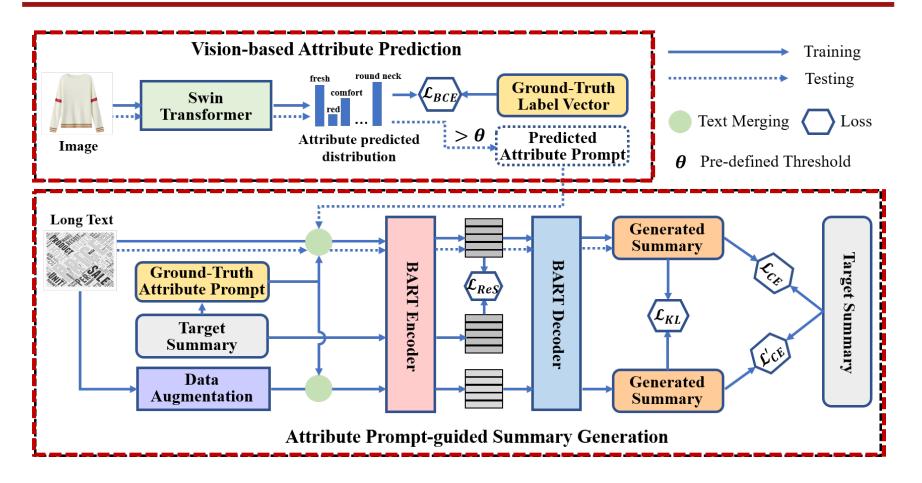
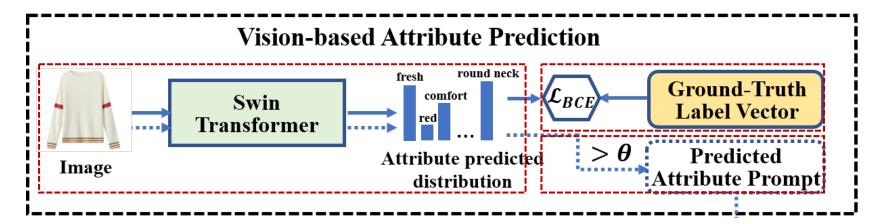


Fig.1: The proposed V2P scheme.

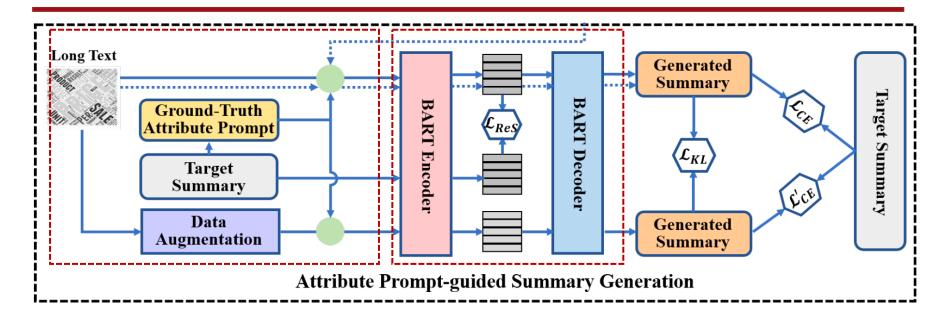


Attribute Prediction

$$\hat{\mathbf{s}} = Swin(V),$$

$$\widehat{\mathcal{A}} = \{a^k | \hat{s}^k \ge \theta, k = 1, \cdots, S\},$$

$$L_{BCE} = \min_{\Theta_a} - [\mathbf{s} \ln (\sigma(\hat{\mathbf{s}})) + (\mathbf{1} - \mathbf{s}) \ln (\mathbf{1} - \sigma(\hat{\mathbf{s}}))],$$



Text and Attribute Prompt Embedding

$$\widehat{X} = [X, a^{k_1}, a^{k_2}, \cdots, a^{k_Q}]$$

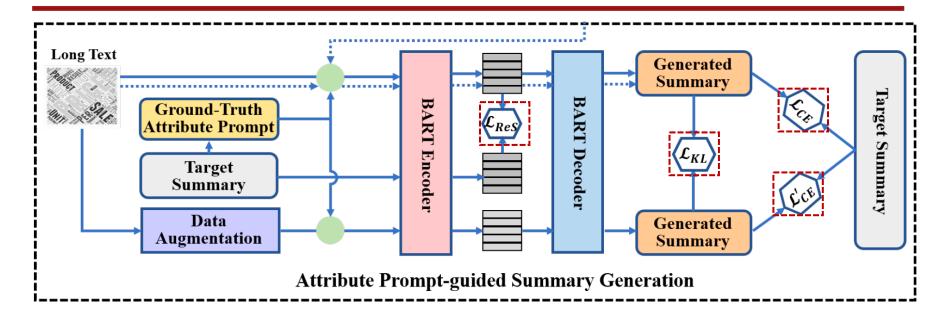
$$\mathbf{e}^{j} = \mathbf{W}^{T} \mathbf{g}^{j}, j = 1, \dots, U,$$

$$\mathbf{E} = [\mathbf{e}^1; \mathbf{e}^2; \cdots; \mathbf{e}^U] + \mathbf{E}_{pos},$$

BART-based Summary Generation

$$Z = \mathcal{E}(E)$$
,

$$\hat{\mathbf{p}}_{\mathbf{j}} = \mathcal{D}(\mathbf{Z}, \hat{y}_1, \hat{y}_2, \cdots, \hat{y}_{j-1}),$$



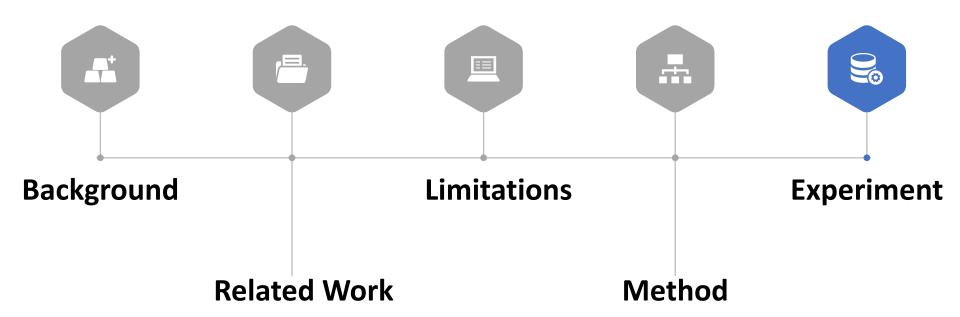
Output-level and Representation-level Supervision

$$\begin{split} \mathcal{L}_{CE} &= -\frac{1}{L} \sum_{j=1}^{L} log(\hat{\mathbf{p}}_{j}[t*]), \\ \mathcal{L}_{ReS} &= \mathbb{E}_{p(z_{1:B})p(\tilde{z}|z_{1})} \Big[-log \frac{e^{f(z^{y},z_{1})}}{\sum_{i=1}^{B} e^{f(z^{y},z_{i})}} \Big], \end{split}$$

Data Augmentation-based Robustness Regularization

$$\mathcal{L}'_{CE} = -\frac{1}{L} \sum_{i=1}^{L} log(\hat{\mathbf{p}}'_{j}[t*]),$$

$$\mathcal{L}_{KL} = \sum_{j=1}^{L} D_{KL}(\hat{\mathbf{p}}_{j} || \hat{\mathbf{p}}'_{j}) = \sum_{j=1}^{L} \sum_{k=1}^{L} \hat{p}_{jk} log(\frac{\hat{p}_{jk}}{\hat{p}'_{jk}}).$$



Dataset

Table 1: Dataset statistics of CEPSUM.

| Category | Home Appliances | Clothing | Cases&Bags | |
|----------------|-----------------|----------|------------|--|
| #Train Sample | 437,646 | 790,297 | 97,510 | |
| #Valid Sample | 10,000 | 10,000 | 5,000 | |
| #Test Sample | 10,000 | 10,000 | 5,000 | |
| Avg Input Len | 335 | 286 | 299 | |
| Avg Output Len | 79 | 78 | 79 | |

> Evaluation Metric

Rouge-1, Rouge-2, Rouge-L.

On Model Comparison

Table 2: Performance comparison among different methods.

| Dataset/ | Home Appliances | | | Clothing | | | Cases&Bags | | |
|-----------------|-----------------|--------------|---------|----------|---------|--------------|------------|--------------|--------------|
| Model | Rouge-1 | Rouge-2 | Rouge-L | Rouge-1 | Rouge-2 | Rouge-L | Rouge-1 | Rouge-2 | Rouge-L |
| Lead | 21.97 | 9.54 | 12.79 | 19.83 | 8.39 | 13.56 | 21.49 | 9.37 | 14.19 |
| LexBank | 24.06 | 10.01 | 18.19 | 26.87 | 9.01 | 17.76 | 27.09 | 9.87 | 18.03 |
| Seq2Seq | 21.57 | 7.18 | 17.61 | 23.05 | 6.84 | 16.82 | 23.18 | 6.94 | 17.29 |
| MASS | 28.19 | 8.02 | 18.73 | 26.73 | 8.03 | 17.72 | 27.19 | 9.03 | 18.17 |
| PG | 31.31 | 10.93 | 21.11 | 29.11 | 9.24 | 19.92 | 31.11 | 10.27 | 21.79 |
| MMPG | 32.88 | <u>11.88</u> | 21.96 | 30.73 | 10.29 | 21.25 | 32.69 | <u>11.78</u> | 22.27 |
| VG-Bart-1 | 32.71 | 11.46 | 22.87 | 31.36 | 9.94 | 21.34 | 32.73 | 10.26 | 22.44 |
| VG-Bart-2 | 32.73 | 11.74 | 23.61 | 31.63 | 10.08 | <u>21.64</u> | 33.30 | 11.31 | <u>23.13</u> |
| V2P | 34.47* | 12.63* | 25.09* | 35.05* | 11.98* | 22.62* | 34.65* | 11.89* | 24.53* |
| Improvement . ↑ | 4.84% | 6.32% | 6.27% | 10.81% | 16.42% | 5.06% | 4.05% | 0.93% | 6.05% |

V2P consistently surpasses all the baselines, exhibiting the effectiveness of the proposed scheme.

On Ablation Study

Table 3: Performance comparison between V2P and its derivatives.

| Dataset/ | Home Appliances | | Clothing | | | Cases&Bags | | | |
|----------------|-----------------|---------|----------|---------|---------|------------|---------|---------|---------|
| Model | Rouge-1 | Rouge-2 | Rouge-L | Rouge-1 | Rouge-2 | Rouge-L | Rouge-1 | Rouge-2 | Rouge-L |
| V2P | 34.47 | 12.63 | 25.09 | 35.05 | 11.98 | 22.62 | 34.65 | 11.89 | 24.53 |
| V2P-w/o-Image | 32.78 | 11.38 | 24.26 | 30.71 | 9.67 | 20.96 | 32.95 | 10.53 | 23.17 |
| V2P-w/o-ReS | 33.74 | 11.73 | 24.64 | 34.81 | 11.67 | 21.76 | 33.77 | 11.11 | 23.48 |
| V2P-w/o-Robust | 34.22 | 12.39 | 24.76 | 34.62 | 11.47 | 21.10 | 33.19 | 10.02 | 22.56 |
| V2P-w-VGG | 32.90 | 11.67 | 23.92 | 34.31 | 11.60 | 21.07 | 33.46 | 10.99 | 23.72 |
| V2P-w-Res | 33.15 | 11.90 | 24.12 | 34.49 | 11.67 | 21.13 | 19.44 | 4.58 | 13.34 |

V2P obtains the best performance, which verifies these components are significant in our model.

On Case Study

Product Long Text Description:

Off season women's classic Pima cotton cardigan, navy blue, cotton, slim fit, long sleeve, round neck, suitable for different seasons. This cardigan is made of ring spun Pima cotton with high softness, slim scissors, slim cutting, long sleeve, rib knitted cuffs. Rib knit hem. Button placket.

Product Image:



GT Summary: This cardigan is made of ring spun Pima cotton with high softness. The slim fit is suitable for different seasons. Thread knitted hem effectively modifies the waist line, which is quite thin.

V2P: This cardigan is made of ring spun leather horse cotton with high softness. It is exquisite and fashionable. The fabric is neat and stylish. The slim version shows more temperament. The cardigan design modifies the facial lines, and the button cardigan is easy to wear and take off.

(Rouge-1: 56.20%, Rouge-2: 42.02%, Rouge-L: 54.00%)

V2P-w/o-Image: Made of high-quality Pima cotton fabric, it feels soft and delicate, has good air permeability and brings a comfortable wearing experience. The classic round neck design naturally fits the neck and is beautiful.

(Rouge-1: 21.90%, Rouge-2: 4.44%, Rouge-L: 15.93%)

Figure 2: Comparison between the summaries generated by our model and its variant V2P-w/o-image for a clothing product.

Conclusion

- ➤ We design a **vision-to-prompt** based multi-modal product summary generation scheme, where the **heterogeneous multi-modal data** are unified in the same space by converting the vision modality into **semantic attribute prompts**.
- ➤ We are the first to perform both output-level and representation-level supervision for MPSG simultaneously and introduce the data augmentation-based robustness regularization.
- ➤ We are the first to adapt the **generative pre-trained language model** to solve the MPSG task and achieve superior performance on a real-world dataset.



Thanks for your listening.



Codes are available!