

Font Shape-to-Impression Translation

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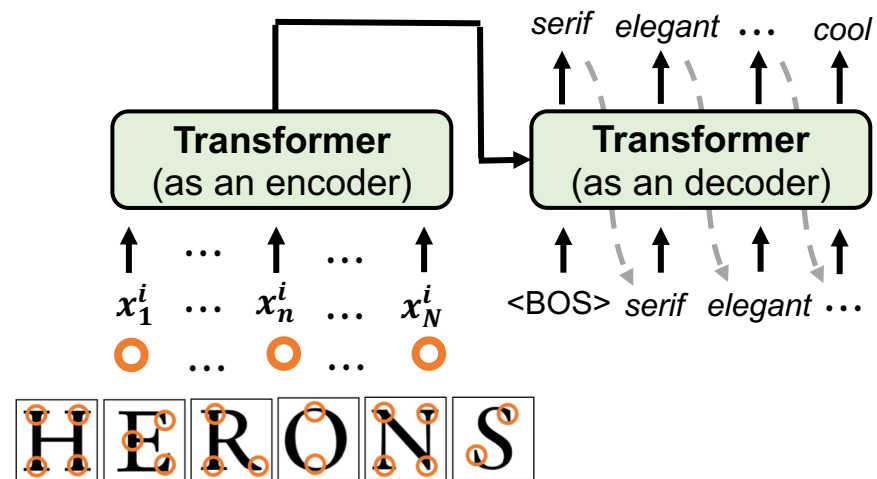
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Key Takeaways

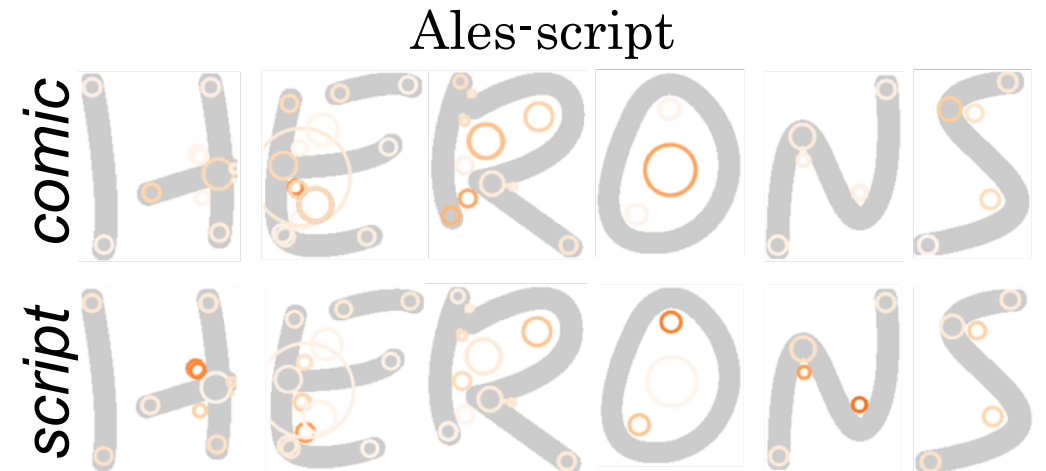
■ Contributions : New part-based font-impression analysis

- Understand the important parts by Transformer and explainable-AI(X-AI) techniques



Font Shape-to-Impression Translation

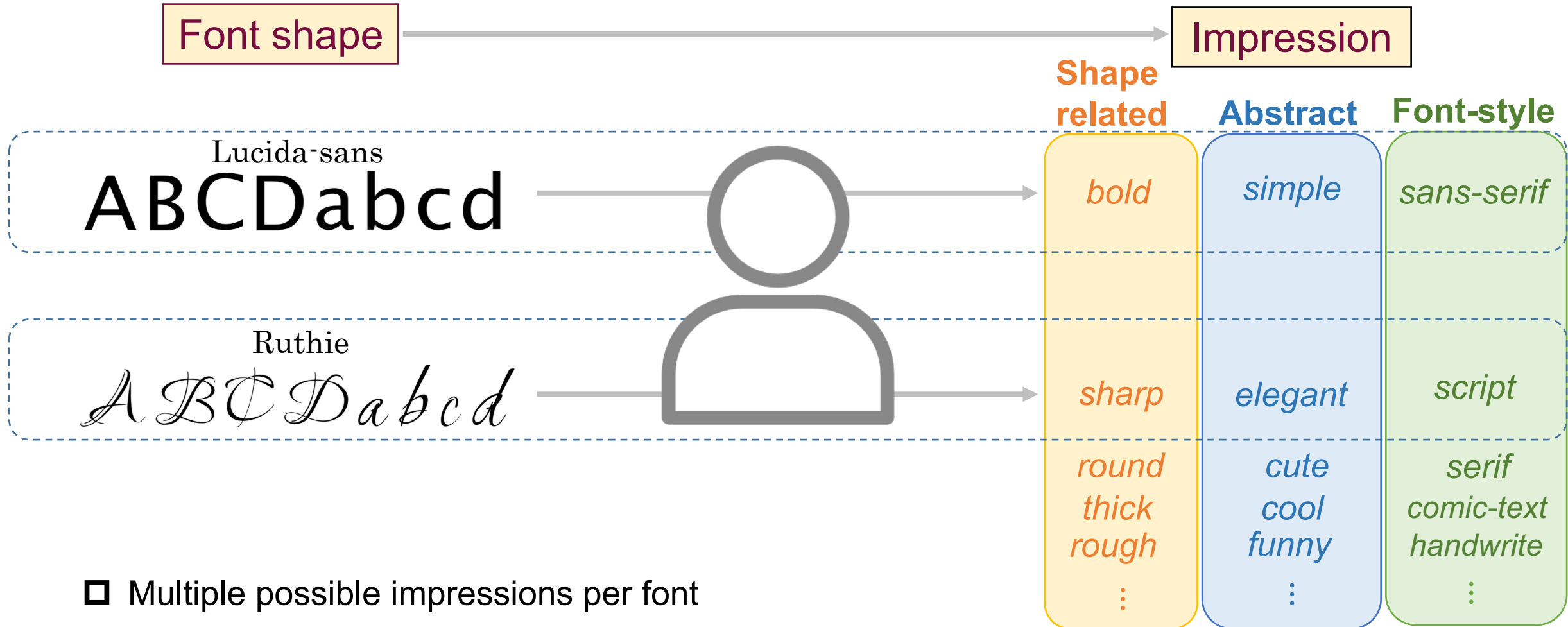
X-AI



Analysis of Important parts for impressions

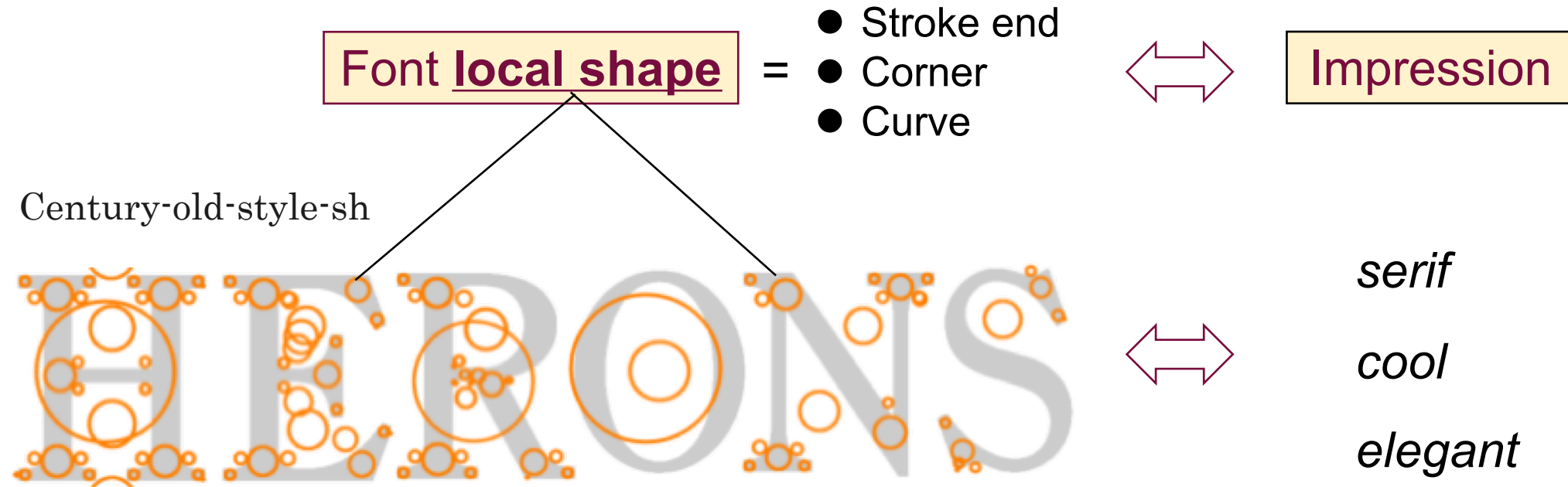
■ Applications : Font recommendation and generation systems for specific impression

Motivation : Different fonts have different impressions



- ❑ Multiple possible impressions per font
- ❑ Relationship between font shape and impression is unrevealed

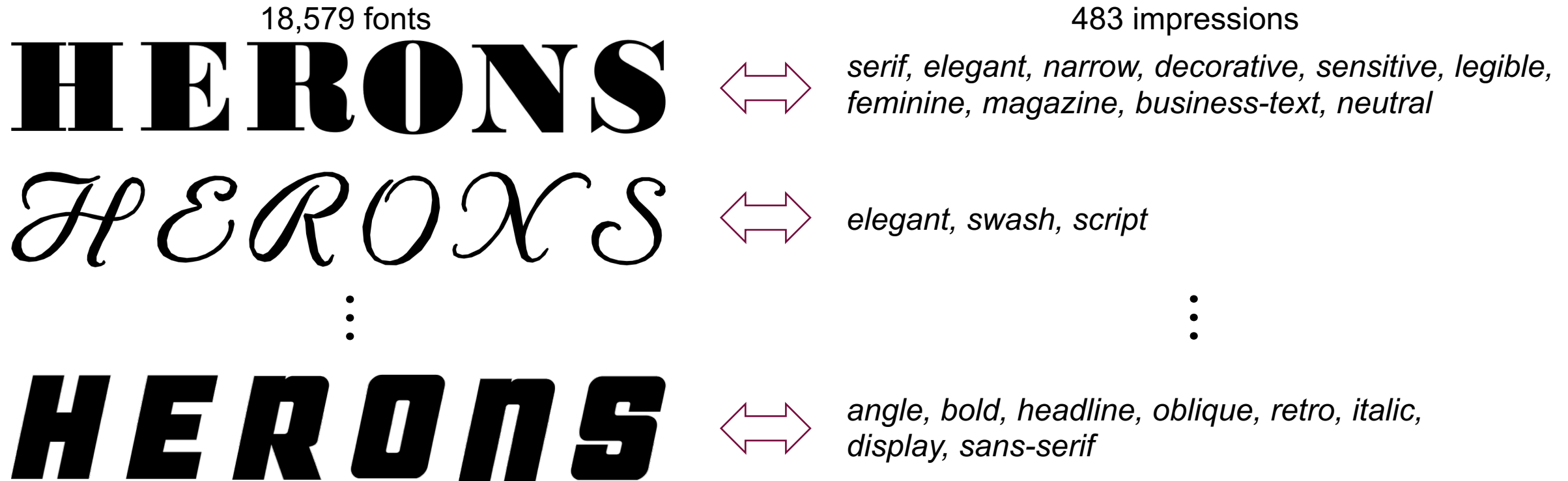
Purpose : Analyze the relationship between font shape and impression



- ❑ Whole shapes → Affects character classes such as "A" and "Z"
- ❑ Local shapes → Not affect character class and the decoration appears

Font-Impression Dataset [*]

- **18,815 fonts** and **1,824 impressions** dataset collected from MyFonts.com
- Use the impressions contained **in more than 100 fonts** and the fonts labeled with them



[*] Chen T. et al., "Large-scale Tag-based Font Retrieval with Generative Feature Learning", ICCV(2019)

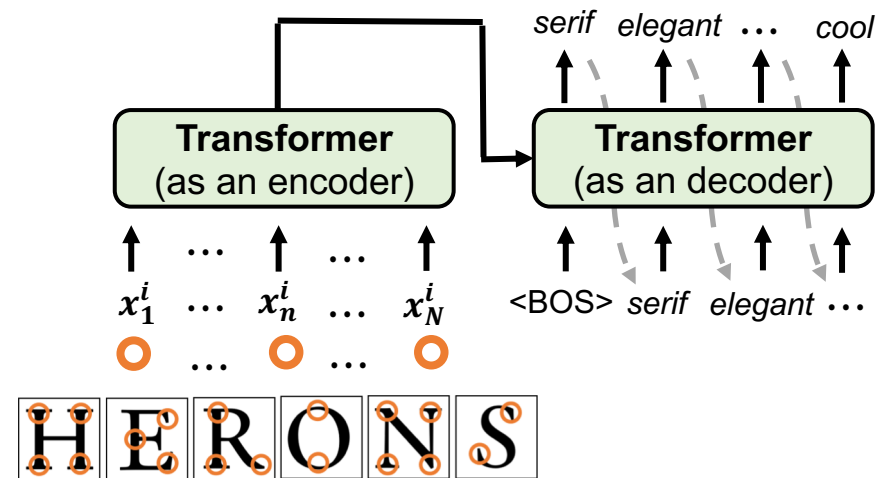
Methodology:
Shape-to-impression translation

Strategy : Local Shape \times Transformer [*] \times X-AI

- ① Local shape \rightarrow Feature (SIFT[**]) vector (extracted from “HERONS” per a font)



- ② Output an arbitrary number of impression from a set of SIFT vectors by Transformer

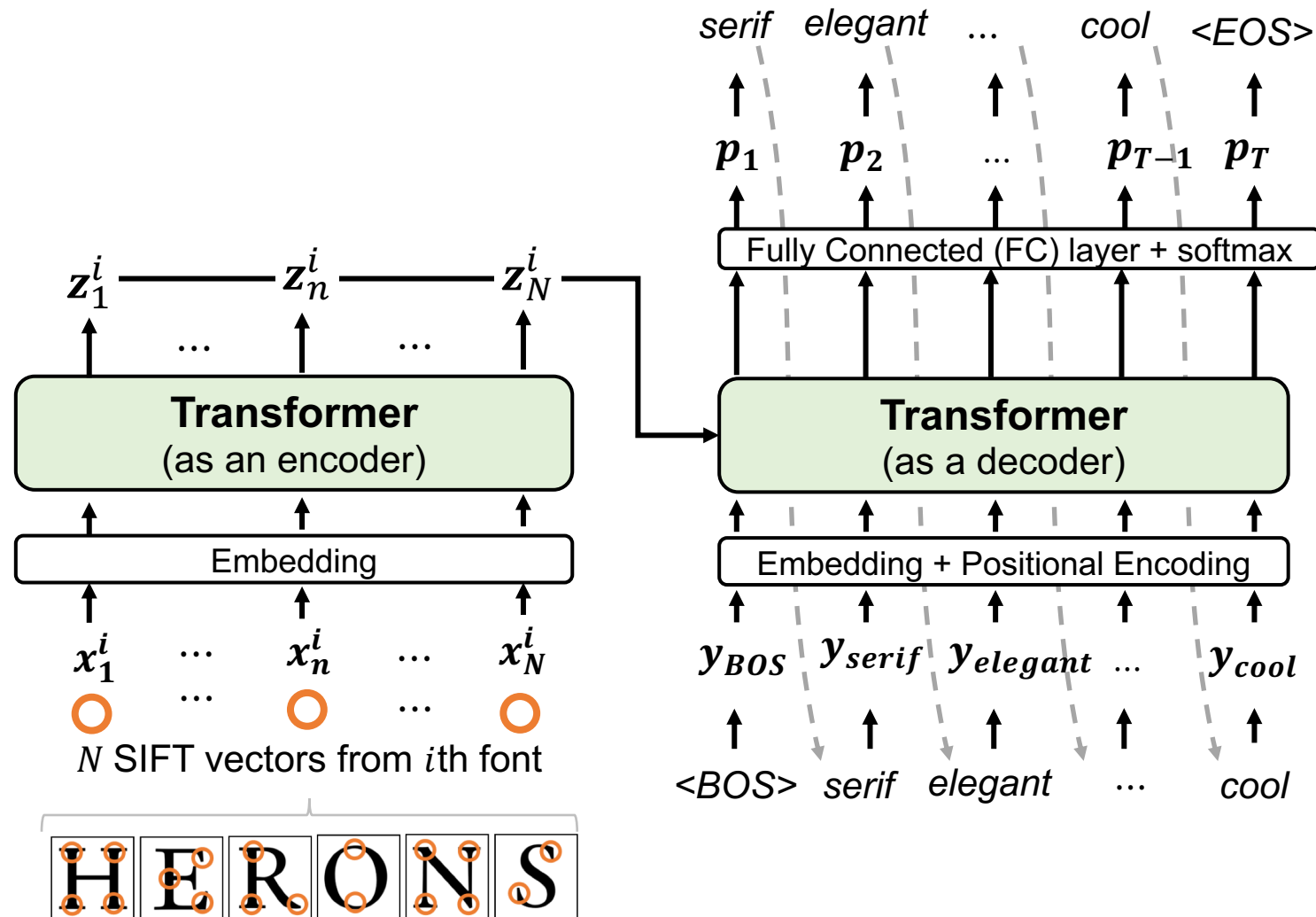


- ③ Analyze the important parts for specific impressions by **explainable-AI (X-AI) techniques**

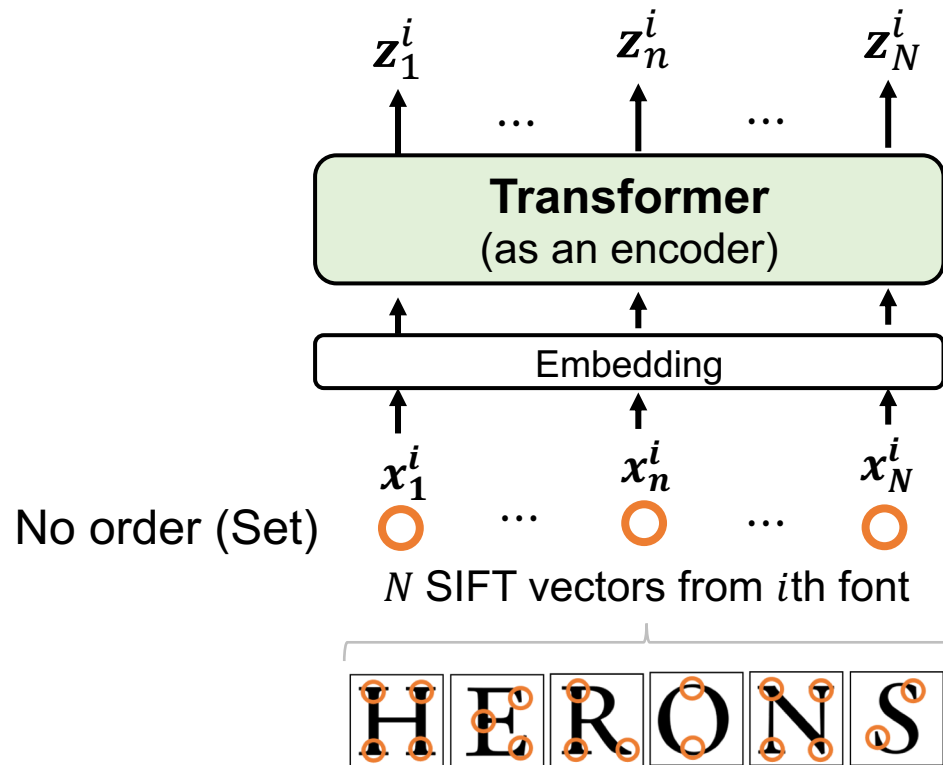
[*] Vaswani et al, "Attention is all you need", NeurIPS (2017)

[**] Lowe, D.G., "Distinctive Image Features from Scale-invariant Keypoints", Int. J.Comp. Vis.60(2), 91–110 (2004)

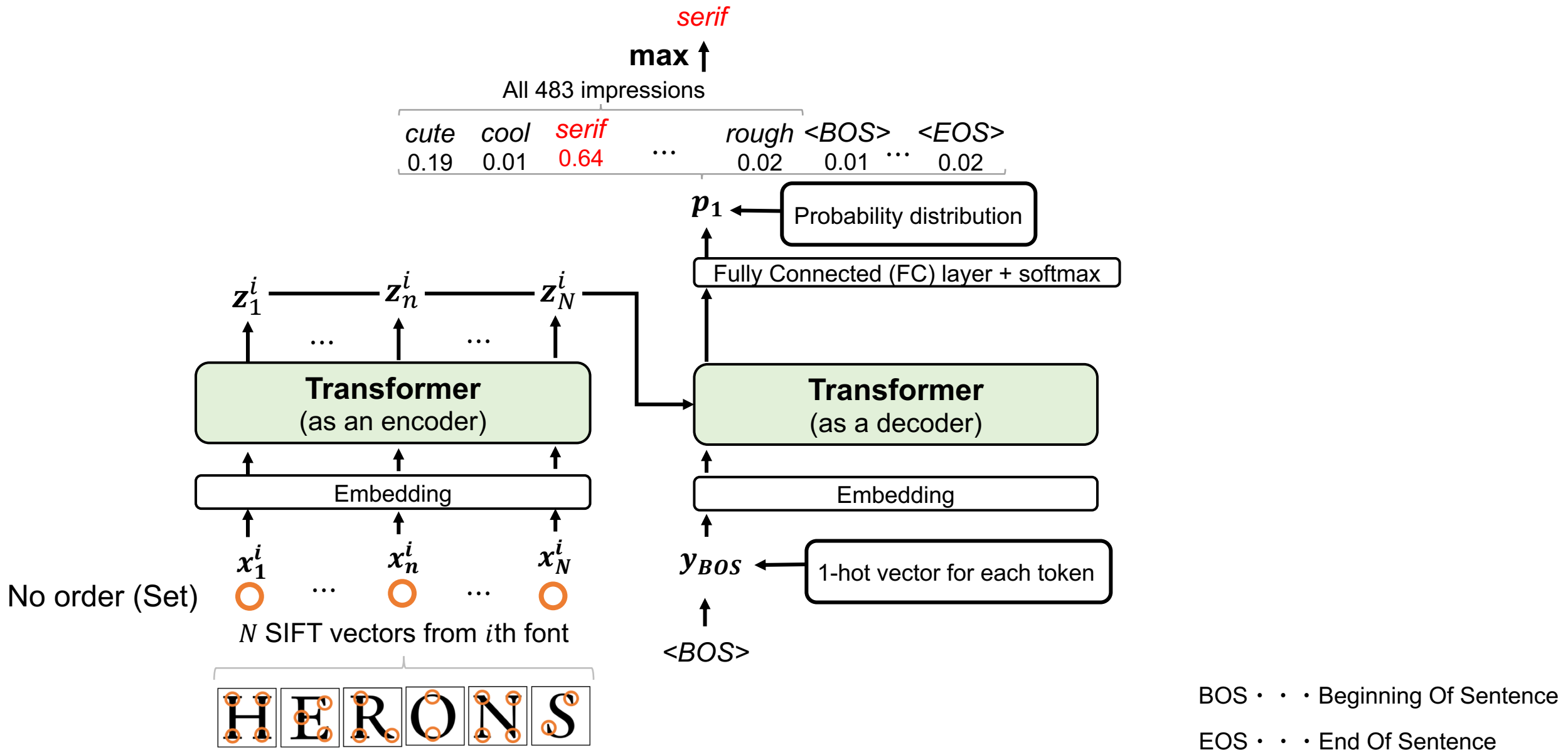
Our Model : Shape-to-Impression Translator



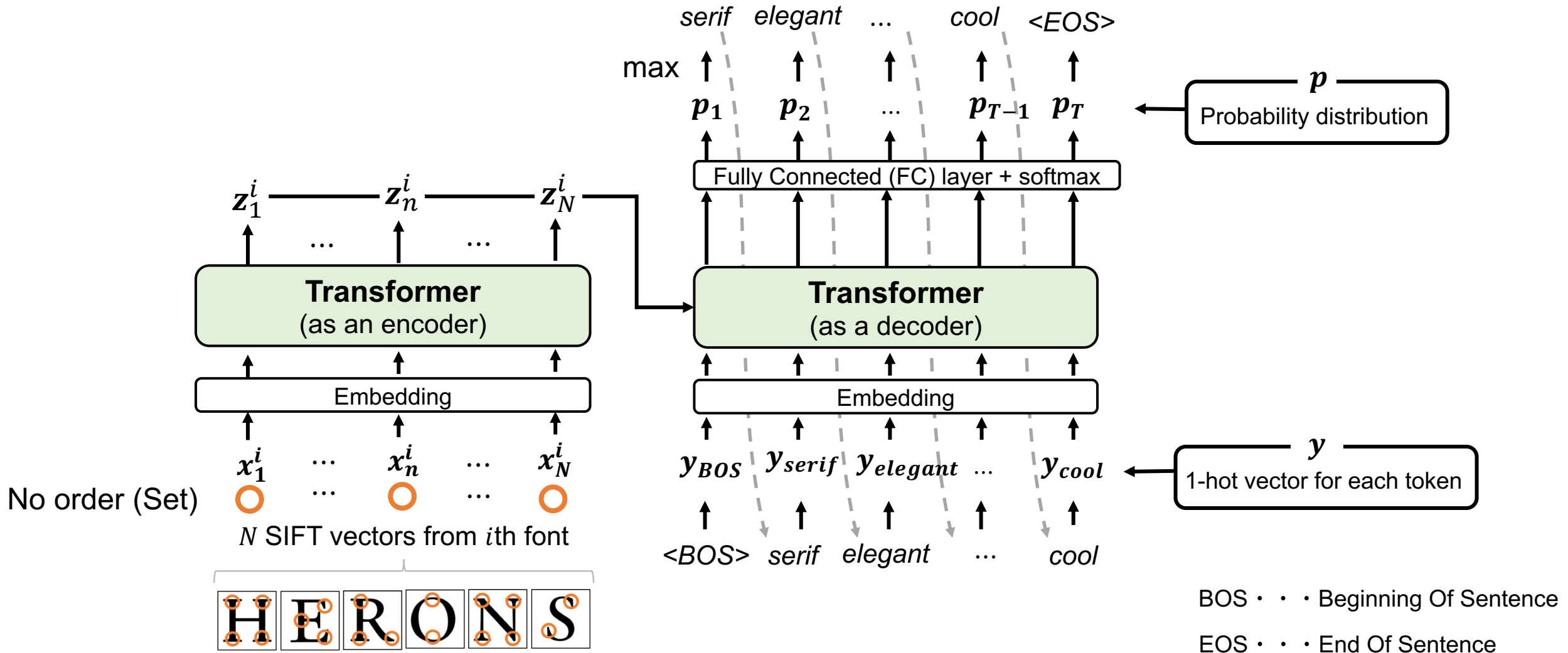
Our Model : Shape-to-Impression Translator



Our Model : Shape-to-Impression Translator



Our Model : Shape-to-Impression Translator



Advantages of Translation Framework

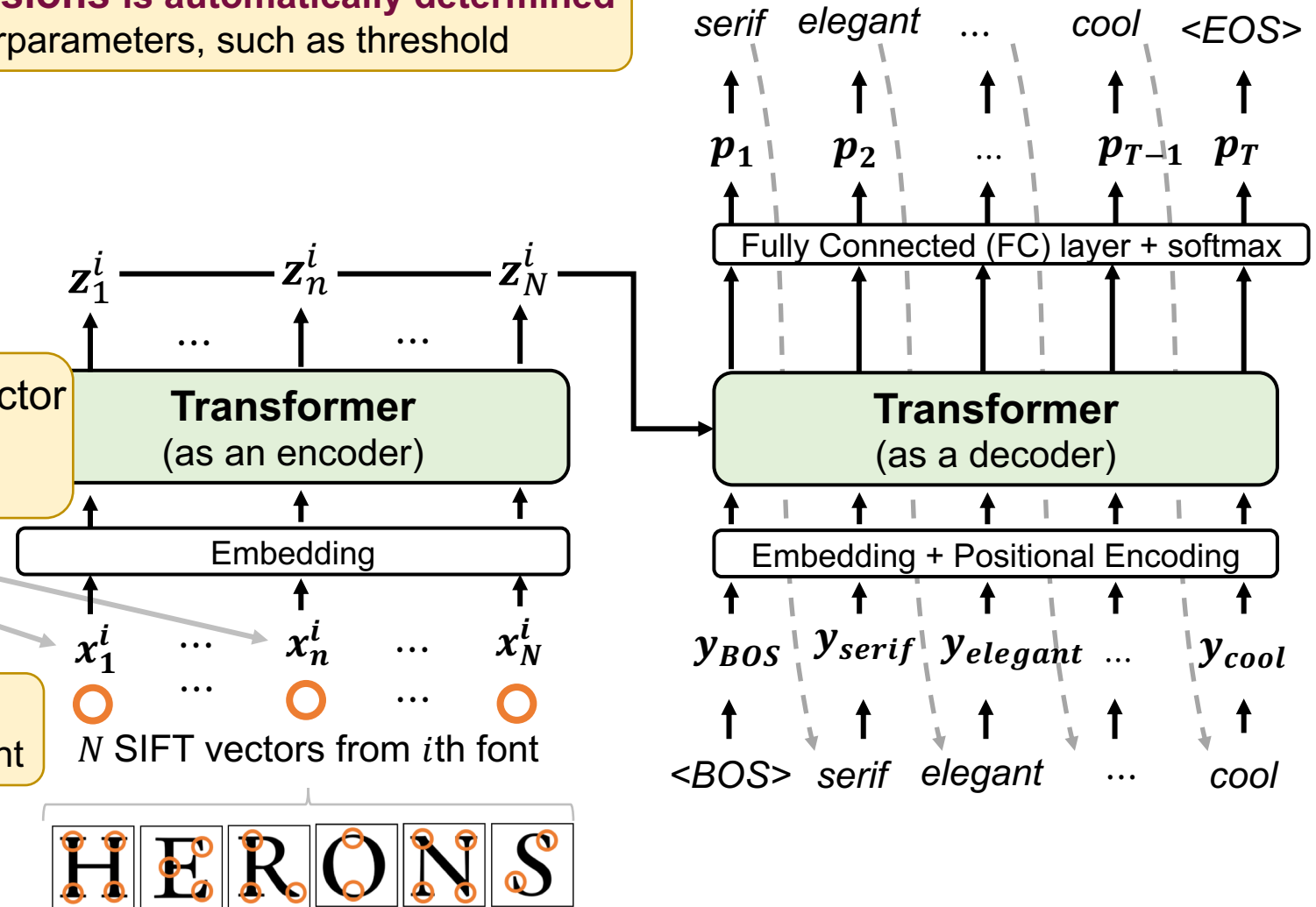
Output impressions is automatically determined without any hyperparameters, such as threshold

Self-attention determines weights for every vector

- **Boost the correlation** among the local shapes that are important for estimation

Accept **a variable number** of SIFT vectors

- The number of local shapes (N) is **variable** per font

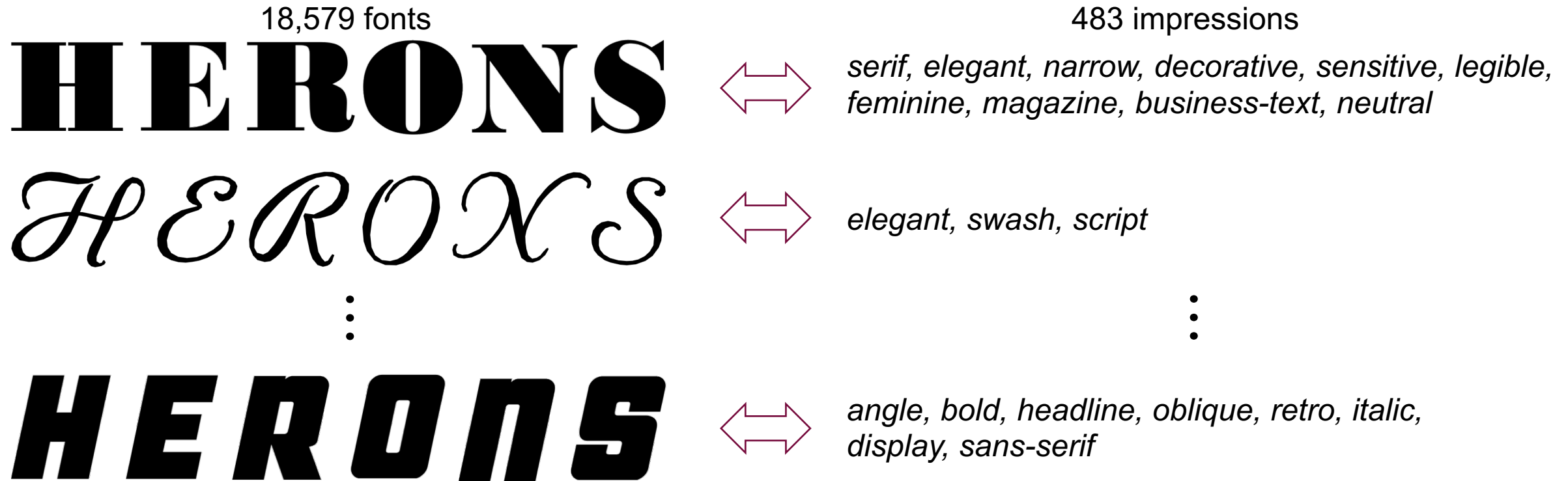


Experimental result (1/2)

Accuracy of the translation results

Font-Impression Dataset [*]

- 18,815 fonts and 1,824 impressions dataset collected from MyFonts.com
- Use the impressions contained in more than 100 fonts and the fonts labeled with them
- Train : Validation : Test = 14,876 : 1,856 : 1,847 fonts



Translation Examples

Clothe

HERONS

Original image



SIFT vectors

Ground Truth

*decorative, display, headline, poster, magazine, **handwrite**, retro, **script**, **elegant**, informal, logo, vintage, alternate, brand, **round**, ligature, invitation, advertise, **calligraphy**, **handmade**, hand, **hand-drawn**, **brush**, **calligraphic**, cursive, signage, connect, logotype, **soft**, contextual-alternates*

Century-old-style-sh

HERONS

Original image



SIFT vectors

Ground Truth

*magazine, **legible**, **serif**, **news**, technical, american, masculine, 1900s, transitional, **news-headline**, **news-text***

Translation Examples

Clothe



Ground Truth

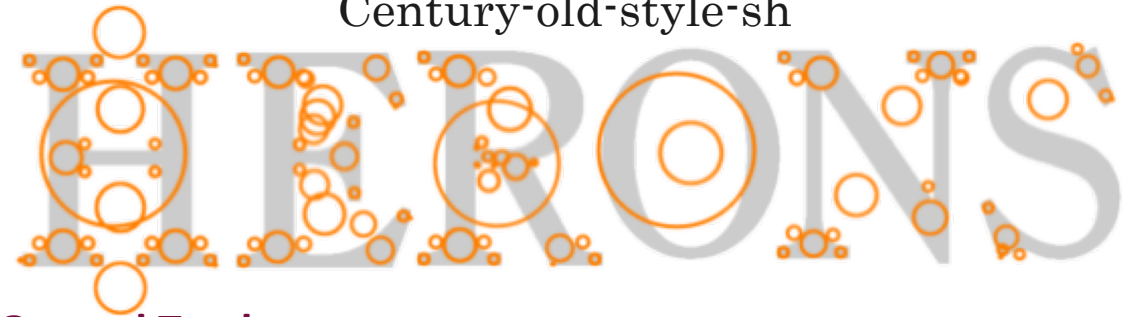
*decorative, display, headline, poster, magazine, **handwrite**, retro, **script**, **elegant**, informal, logo, vintage, alternate, brand, **round**, ligature, invitation, advertise, **calligraphy**, **handmade**, hand, **hand-drawn**, **brush**, **calligraphic**, cursive, signage, connect, logotype, **soft**, contextual-alternates*



Translation Result

*decorative, display, headline, poster, magazine, **handwrite**, retro, **script**, **elegant**, informal, logo, vintage, alternate, brand, **round**, ligature, invitation, advertise, **calligraphy**, **handmade**, hand, **hand-drawn**, **brush**, **calligraphic**, cursive, signage, connect, logotype, **soft**, contextual-alternates*

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Ground Truth

magazine**, **legible**, **serif**, **news**, technical, american, masculine, 1900s, **transitional**, news-headline, **news-text



Translation Result

magazine**, **legible**, **serif**, text, book, **news**, newspaper, workhorse, **transitional**, book-text, **news-text

Translation Examples (Common impression of GT and result is green highlights)

Clothe



Ground Truth

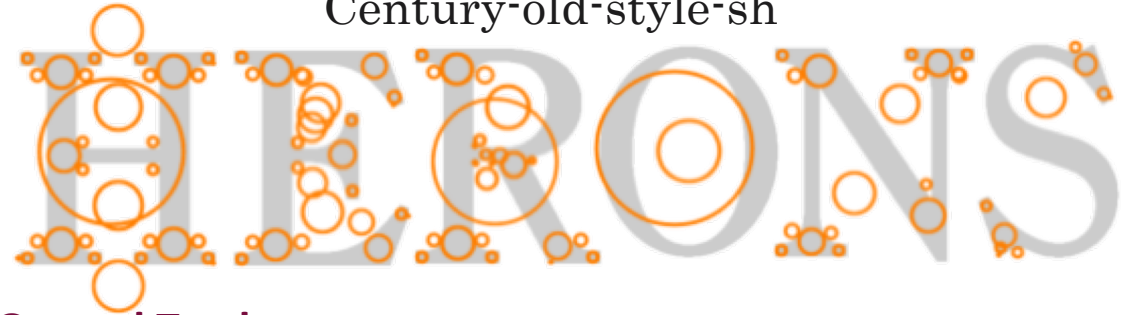
decorative, display, headline, poster, magazine, handwrite, retro,
script, elegant, informal, logo, vintage, alternate, brand, round,
ligature, invitation, advertise, calligraphy, handmade, hand,
hand-drawn, brush, calligraphic, cursive, signage, connect,
logotype, soft, contextual-alternates

Translation Result

decorative, display, headline, poster, magazine, handwrite, retro,
script, elegant, informal, logo, vintage, alternate, brand, round,
ligature, invitation, advertise, calligraphy, handmade, hand,
hand-drawn, brush, calligraphic, cursive, signage, connect,
logotype, soft, contextual-alternates



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Ground Truth

magazine, legible, serif, news, technical, american,
masculine, 1900s, transitional, news-headline, news-text

Translation Result

magazine, legible, serif, text, book, news, newspaper,
workhorse, transitional, book-text, news-text



Translation Examples (Common impression of GT and result is green highlights)

Clothe



Ground Truth

decorative, display, headline, poster, magazine, handwrite, retro,
script, elegant, informal, logo, vintage, alternate, brand, round,
ligature, invitation, advertise, calligraphy, handmade, hand,
hand-drawn, brush, calligraphic, cursive, signage, connect,
logotype, soft, contextual-alternates

Translation Result

decorative, display, headline, poster, magazine, handwrite, retro,
script, elegant, informal, logo, vintage, alternate, brand, round,
ligature, invitation, advertise, calligraphy, handmade, hand,
hand-drawn, brush, calligraphic, cursive, signage, connect,
logotype, soft, contextual-alternates

Prediction Perfectly !

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Ground Truth

magazine, legible, serif, news, technical, american,
masculine, 1900s, transitional, news-headline, news-text

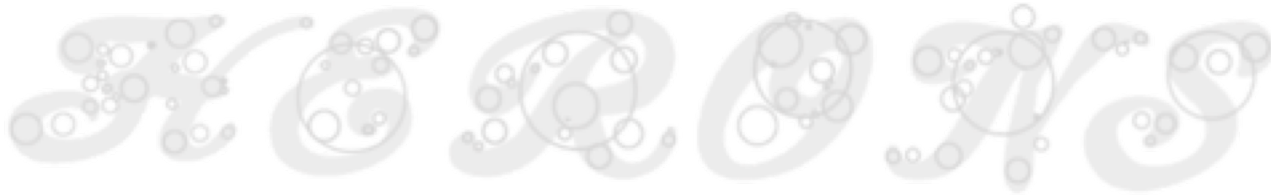


Translation Result

magazine, legible, serif, text, book, news, newspaper,
workhorse, transitional, book-text, news-text

Translation Examples (Common impression of GT and result is green highlights)

Clothe



Ground Truth

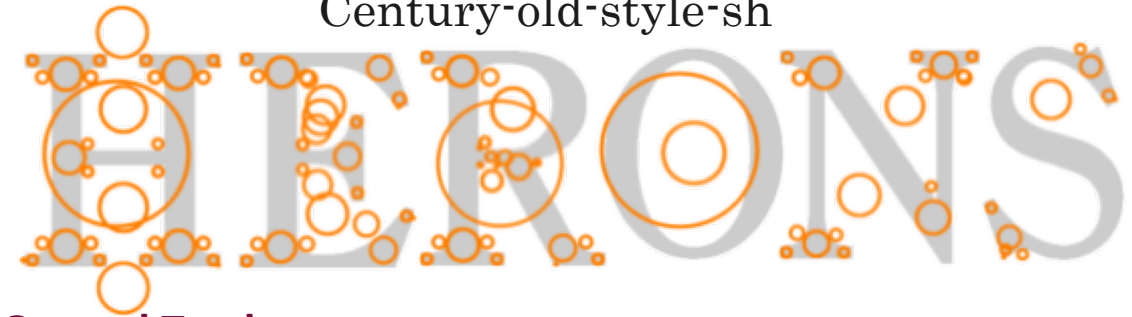
decorative, display, headline, poster, magazine, handwrite, retro,
script, elegant, informal, logo, vintage, alternate, brand, round,
ligature, invitation, advertise, calligraphy, handmade, hand,
hand-drawn, brush, calligraphic, cursive, signage, connect,
logotype, soft, contextual-alternates



Translation Result

decorative, display, headline, poster, magazine, handwrite, retro,
script, elegant, informal, logo, vintage, alternate, brand, round,
ligature, invitation, advertise, calligraphy, handmade, hand,
hand-drawn, brush, calligraphic, cursive, signage, connect,
logotype, soft, contextual-alternates

Century-old-style-sh



Ground Truth

magazine, legible, serif, news, *technical*, *american*,
masculine, *1900s*, transitional, *news-headline*, news-text



Translation Result

magazine, legible, serif, *text*, *book*, news, *newspaper*,
workhorse, transitional, *book-text*, news-text

text, *book*, *book-text*, *newspaper*

- They are similar to GT impressions
- They might be included in the ground-truth

Experimental result (2/2)

Font Shape-Impression

Relationship Analysis

Contribution Visualization of Inputs by Integrated Gradients [*]

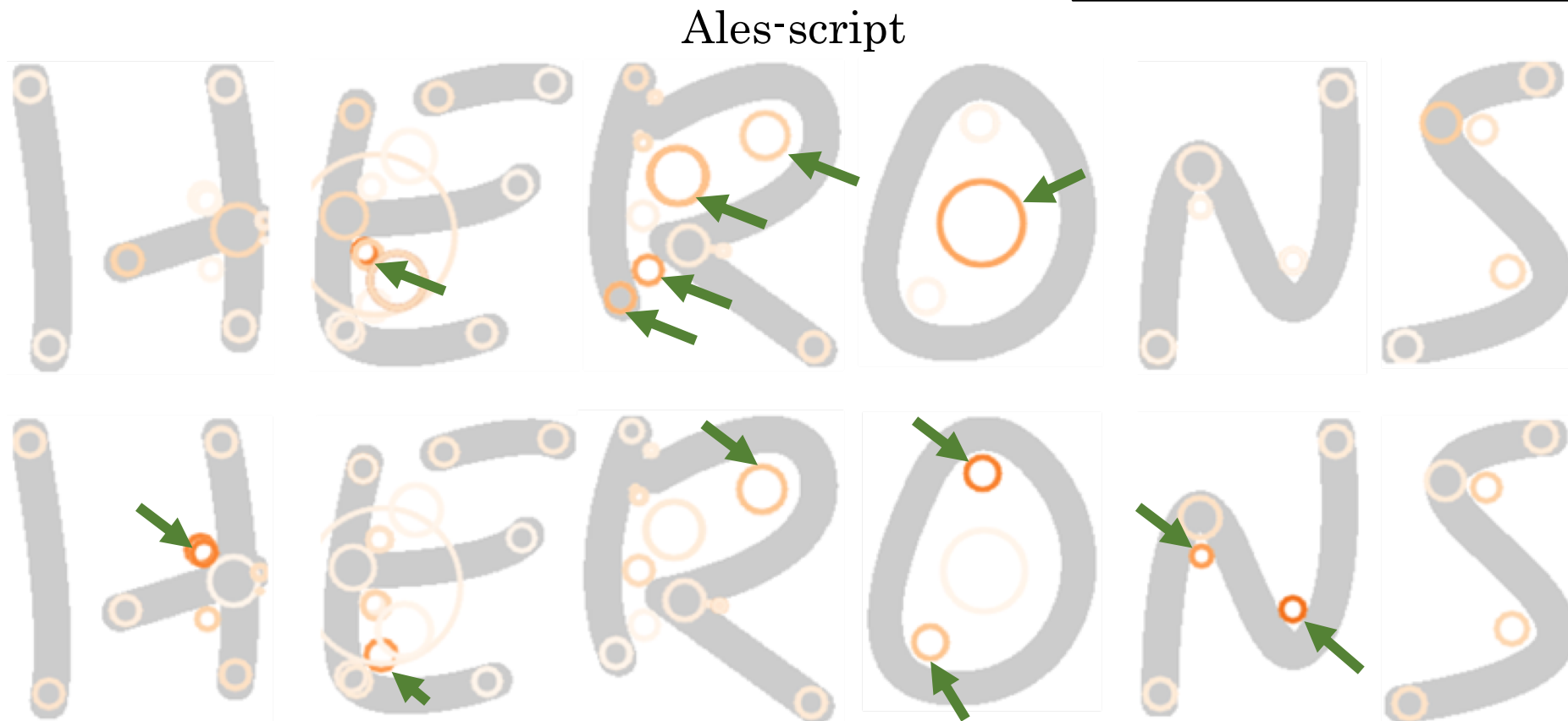
Thicker orange color shape are more important

The shapes of **large space** have large contribution for *comic-text*

Translation Result

comic-text

script



The shapes of **sharp corner** have large contribution for *script*

Contribution Visualization of Inputs by Integrated Gradients [*]

Thicker orange color shape are more important

The shapes of **serif-part** have large contribution for *comic-text*

Linoletter

Translation Result



The shapes of **rounded space** have large contribution for *script*

Concluding remarks

Conclusion

- ❑ Propose the part-based font-impression analysis using Transformer
- ❑ Confirm that Translator can correctly output impressions from a set of local shapes
- ❑ Analyze important local shapes for specific impressions by using the trained Transformer with explainable-AI techniques

Future Work

- ❑ Practical applications of the proposed methods
- ❑ Collaboration with experts of cognitive psychology

Another Simpler Model : Multi-label Classifier

Probability for all impressions (0~1)

<i>serif</i>	<i>cool</i>	<i>beautiful</i>	...	<i>rough</i>
0.89	0.23	0.54	...	0.12

Fully Connected (FC) layer + sigmoid

z_{CLS} z_1^i ... z_n^i ... z_N^i

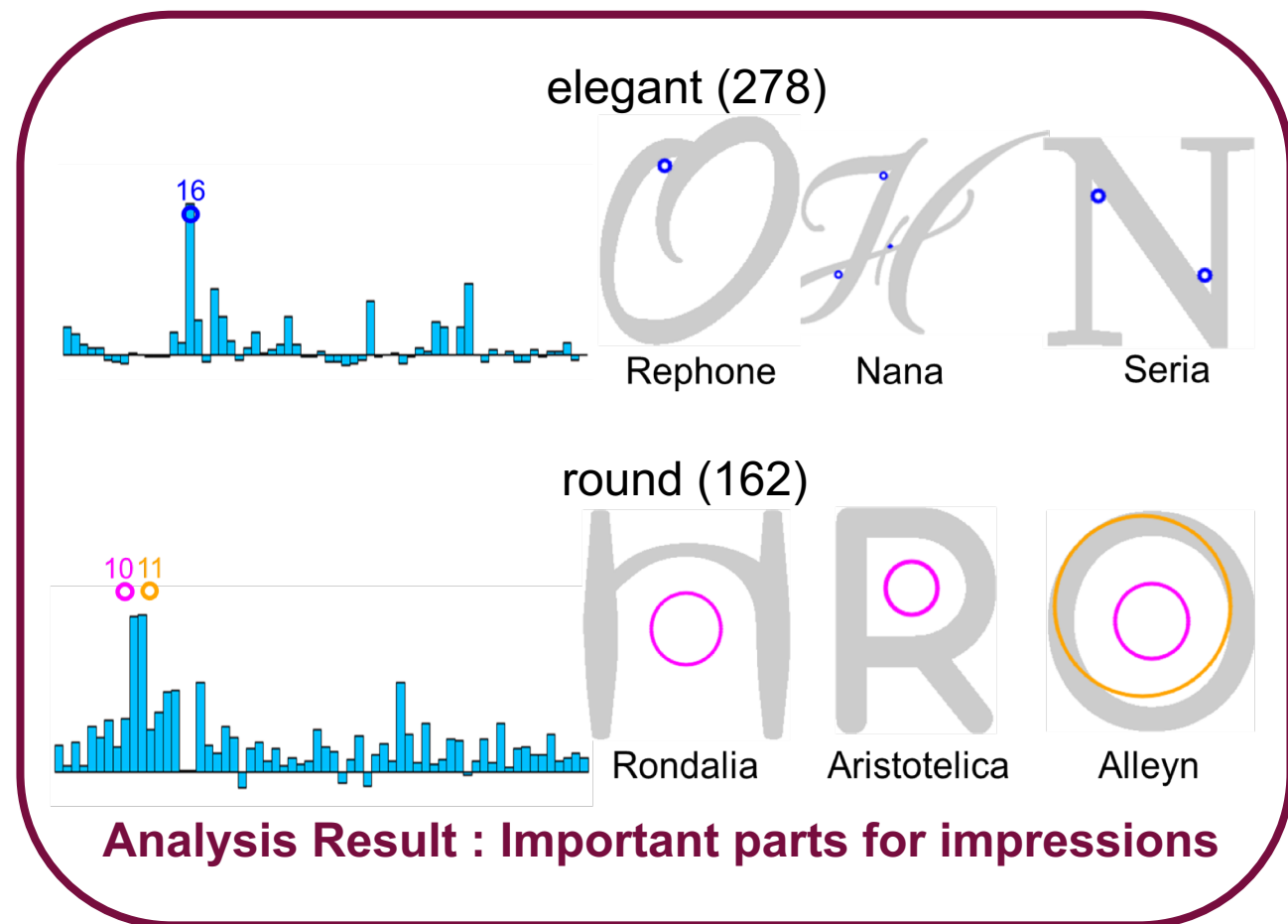
Transformer
(only an encoder)

Embedding

$\langle CLS \rangle$ x_1^i ... x_n^i ... x_N^i

○ ... ○ ... ○

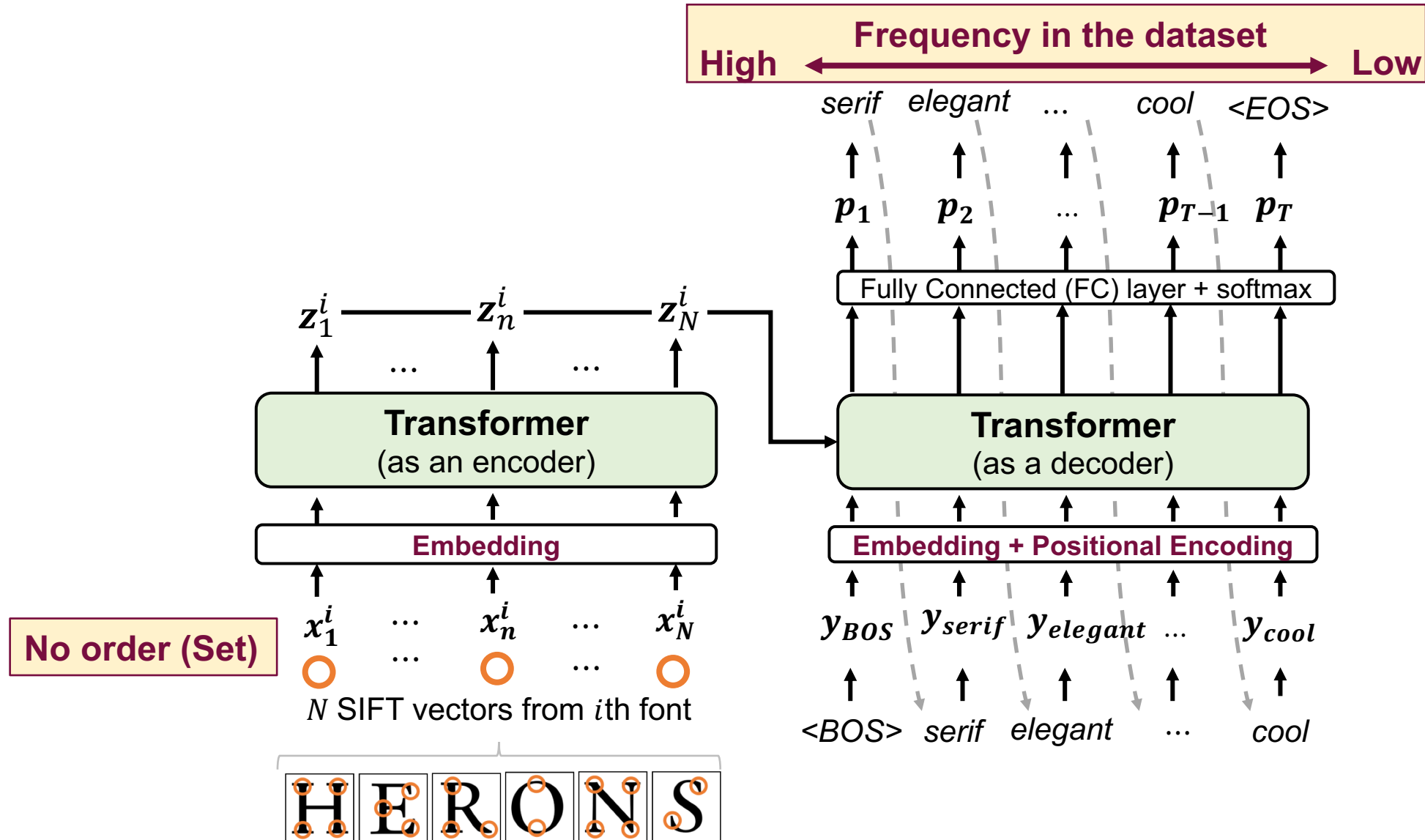
N SIFT vectors from i th font



For its details,
please refer to our paper

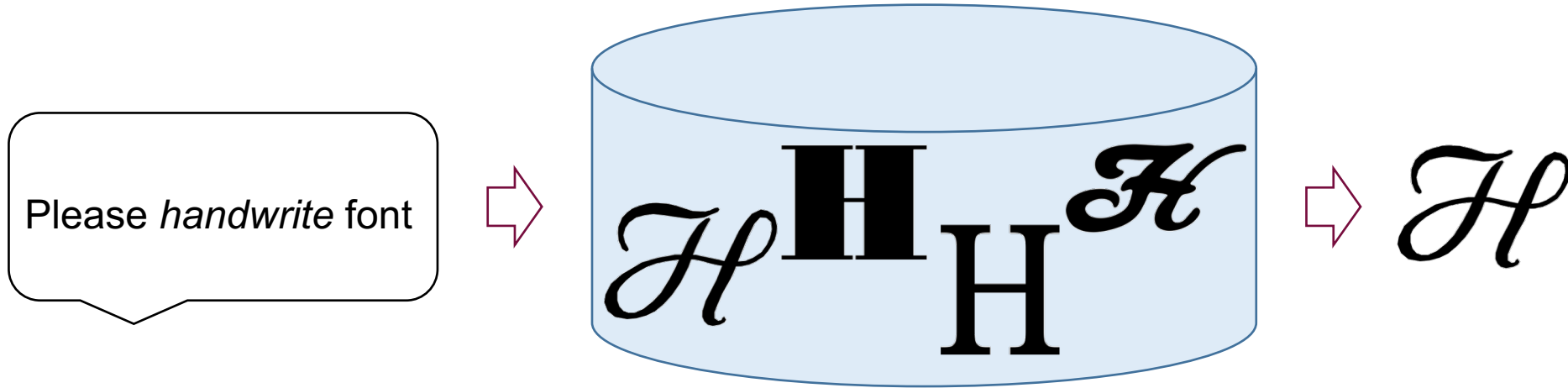
ここまで

The Order of Inputs and Outputs

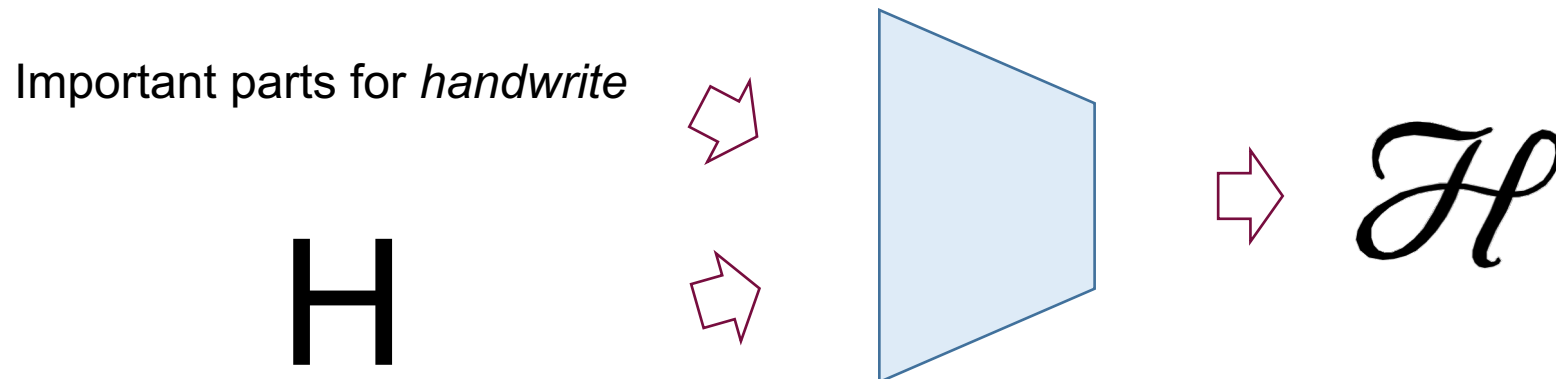


Applications

Font recommendation system



Font generation system



Why Did We Use SIFT?

- ❑ SIFT is so convenient because it gives us more explainability
- ❑ There is no difference between SIFT and SURF

Quantitative Evaluation by **Multi-label Classifier**

	SIFT	SURF
F1@all	0.145	0.140
mAP	0.135	0.151

Strong and Weak Impression Words

Comparison among Methods

- **Multi-label classifier** outperforms the other models on all measures
- **Translator** does not require a threshold

	Multi-label Classifier	Translator	ViT [*]	DeepSets [**]
Inputs	SIFT	SIFT	Patch	SIFT
F1@100↑	0.301	0.264	0.264	0.279
F1@200↑	0.221	0.186	0.185	0.194
F1@all↑	0.145	0.117	0.109	0.110
mAP↑	0.135	N/A	0.115	0.115
Threshold	✓		✓	✓

F1@ n · · · F1 score with frequency top n impressions in the dataset

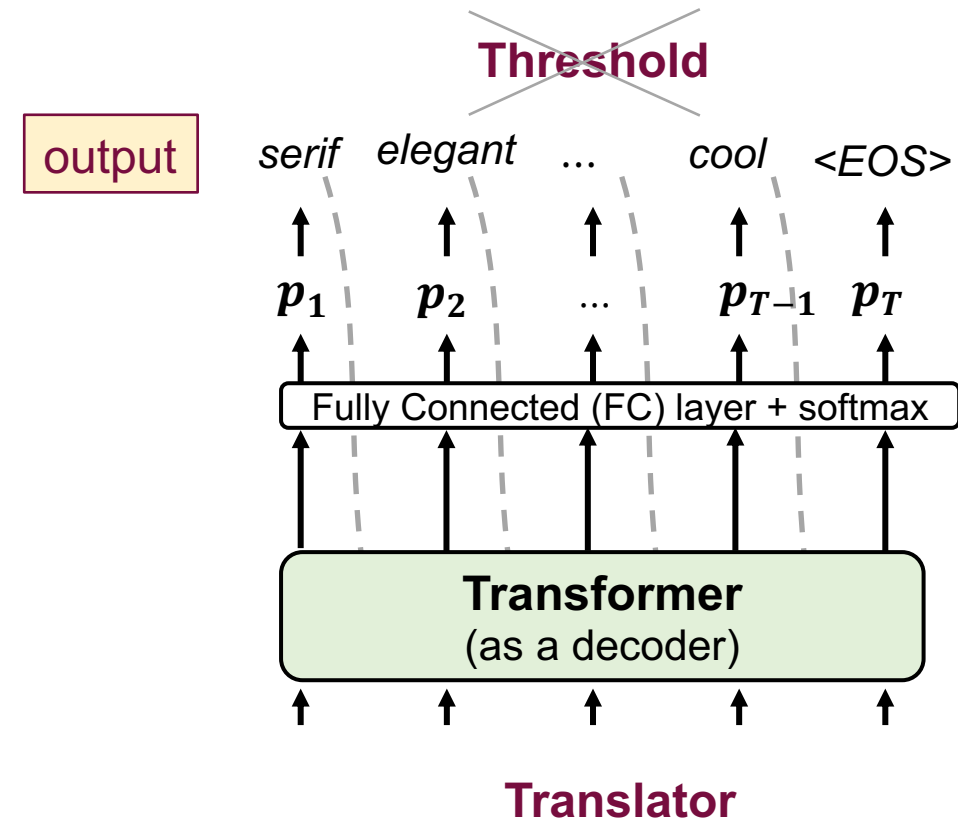
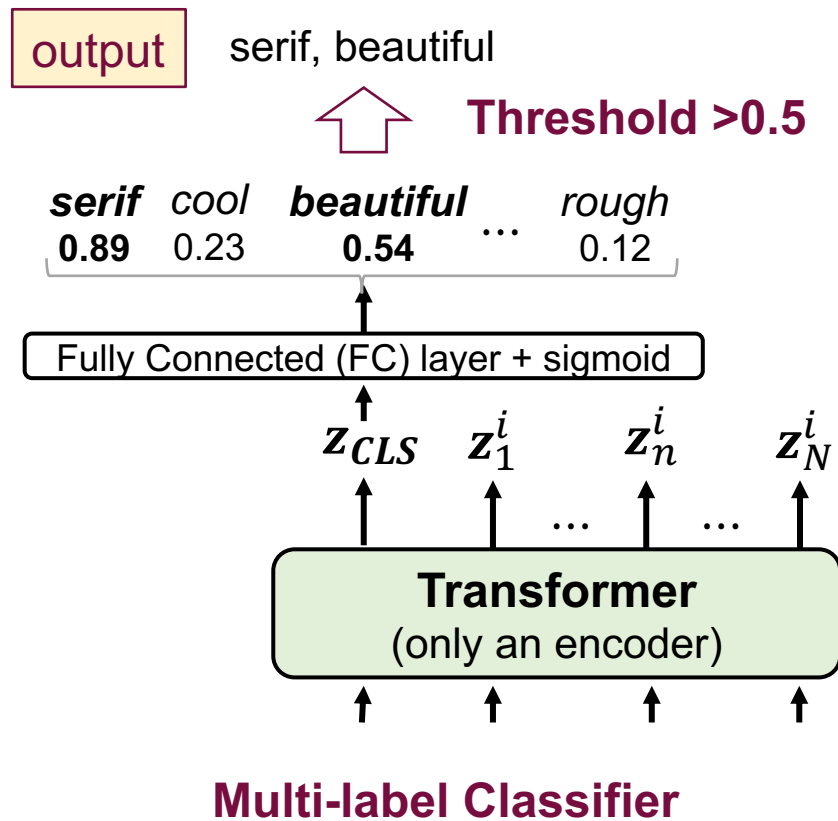
[*] Dosovitskiy, A et al., “An Image is Worth 16x16 Words: Transformers for Image Recognition at Scale”, ICLR(2021)

[**] Ueda, M, et al., “Which Parts Determine the Impression of the Fonts?”, ICDAR (2021)

Translator Does Not Require a Hyperparameter

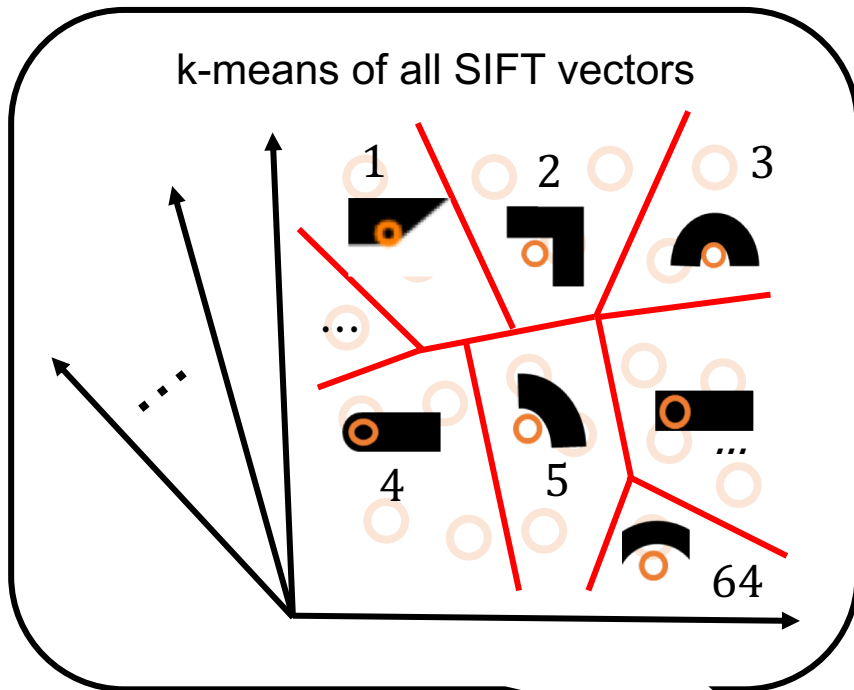
EOS token **determines** the number of output impressions

- In multi-label classifier, we **have to determine hyperparameter (threshold)**

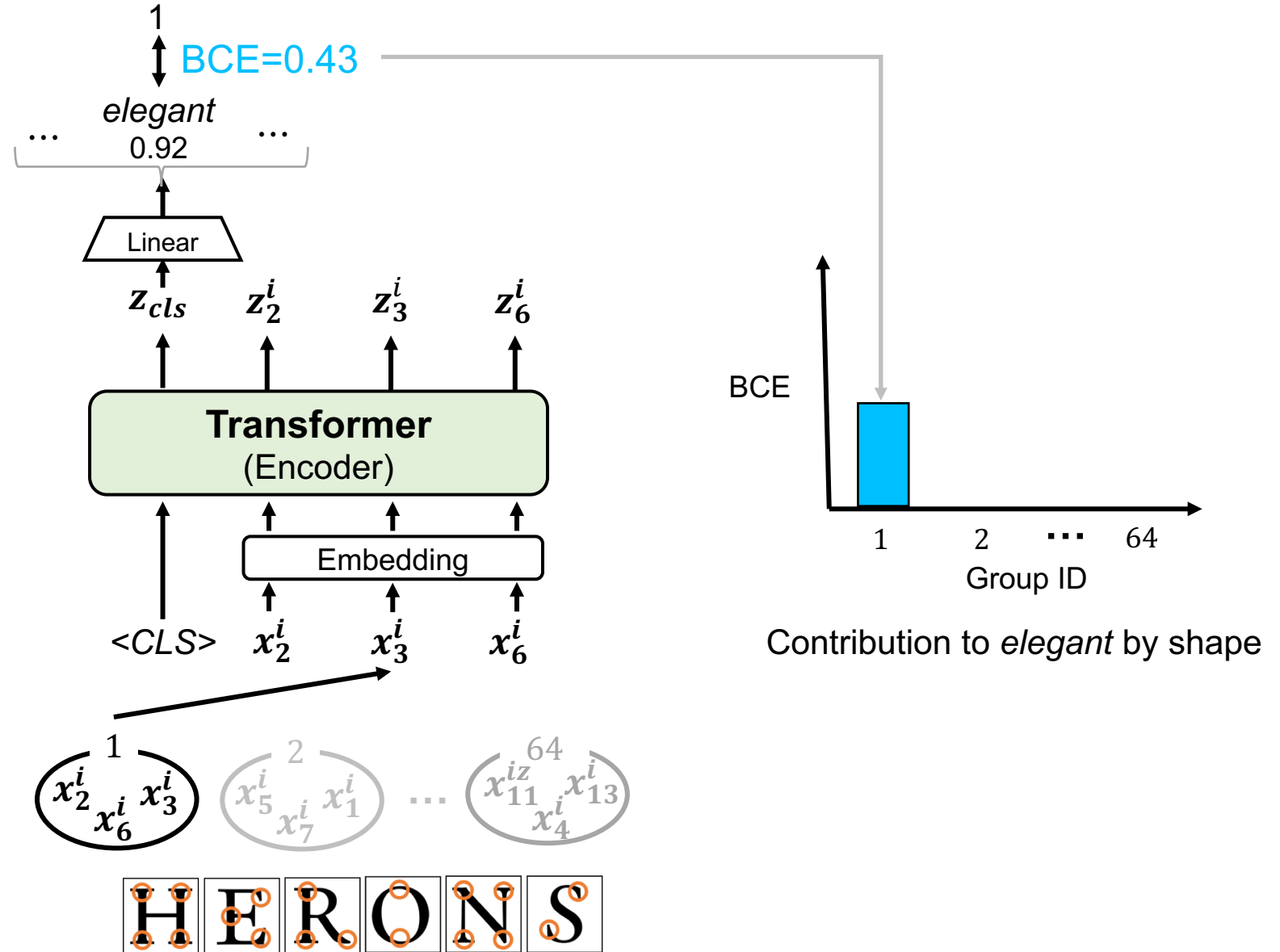


Group-Based Occlusion Sensitivity

Example : *elegant*

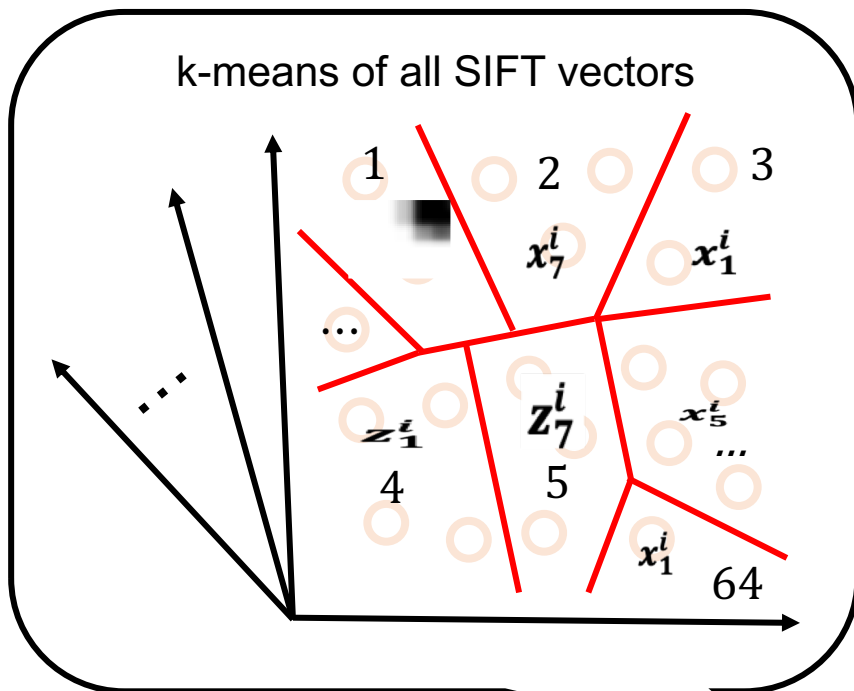


Grouped by clusters to which the SIFT vector belongs

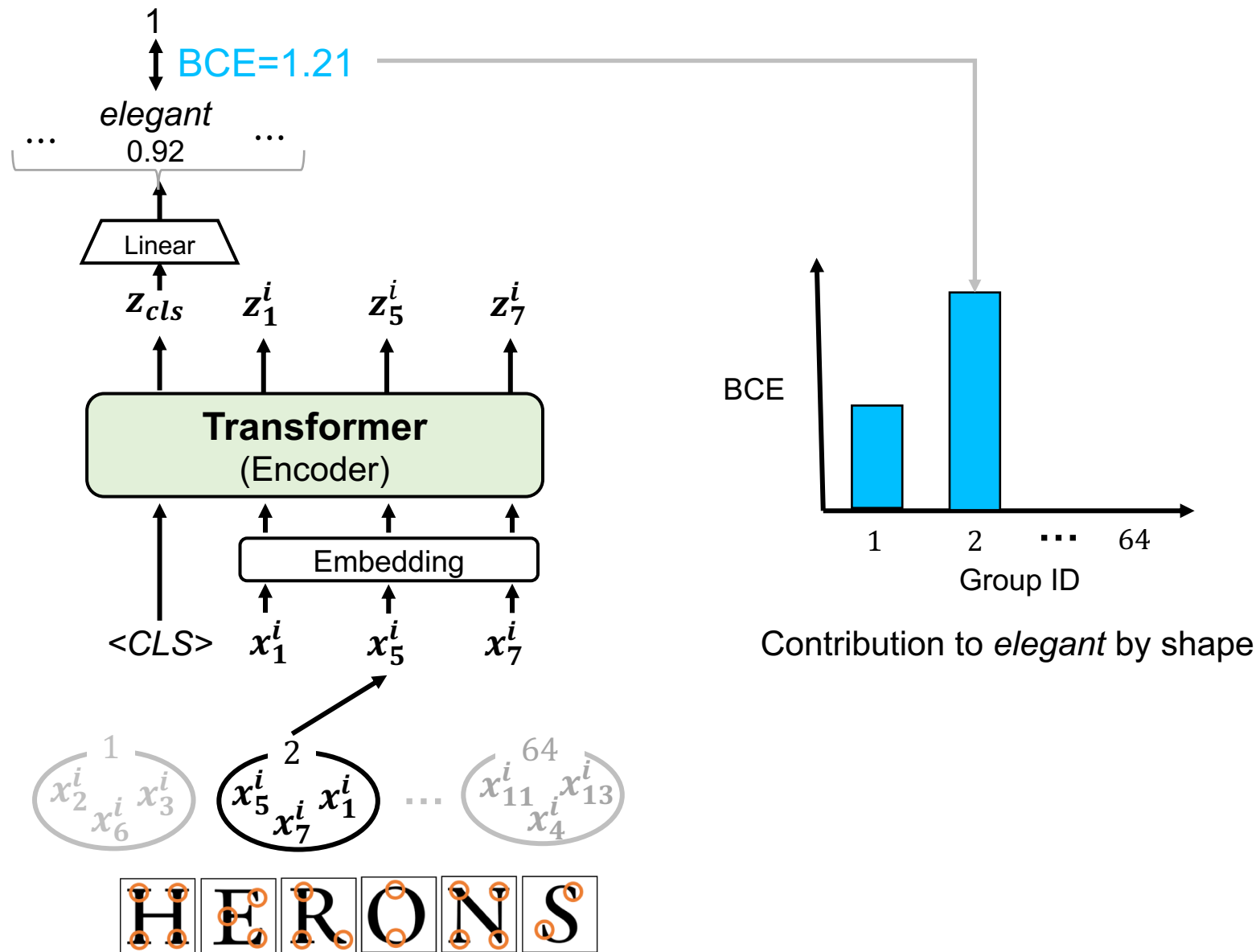


Group-Based Occlusion Sensitivity

Example : *elegant*

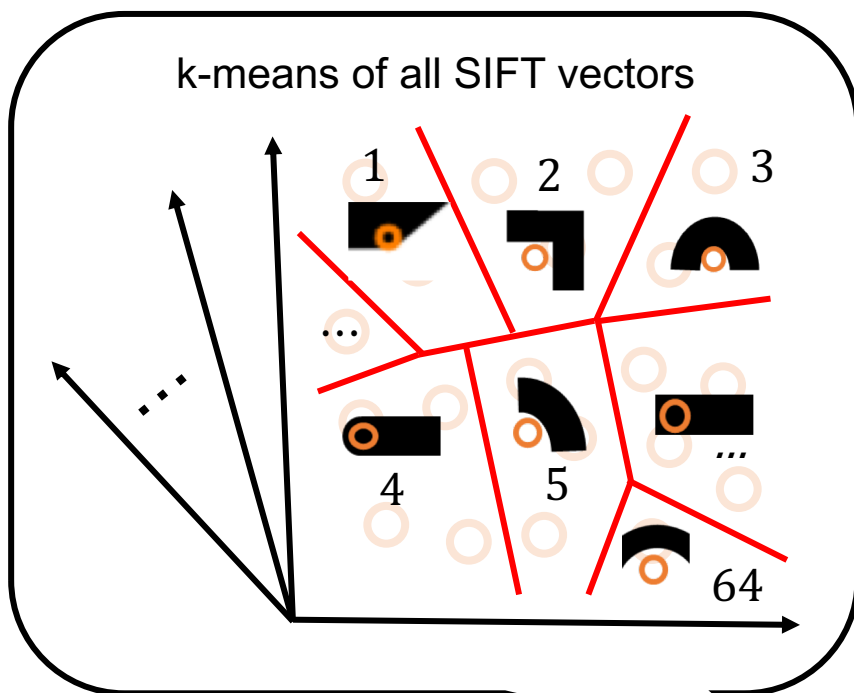


Grouped by clusters to which the SIFT vector belongs

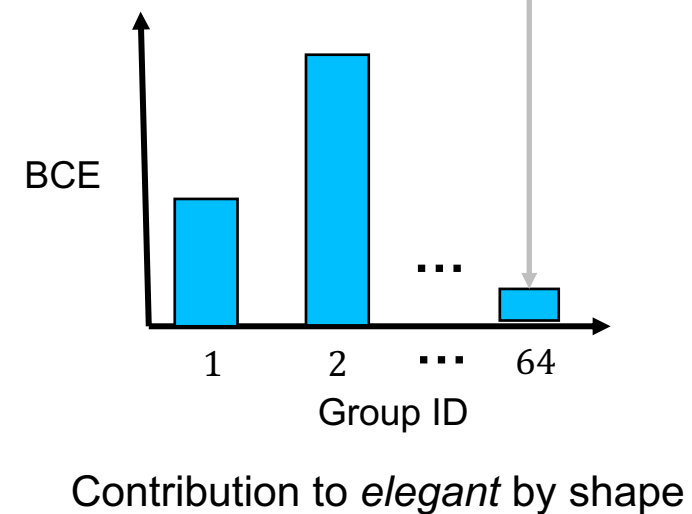
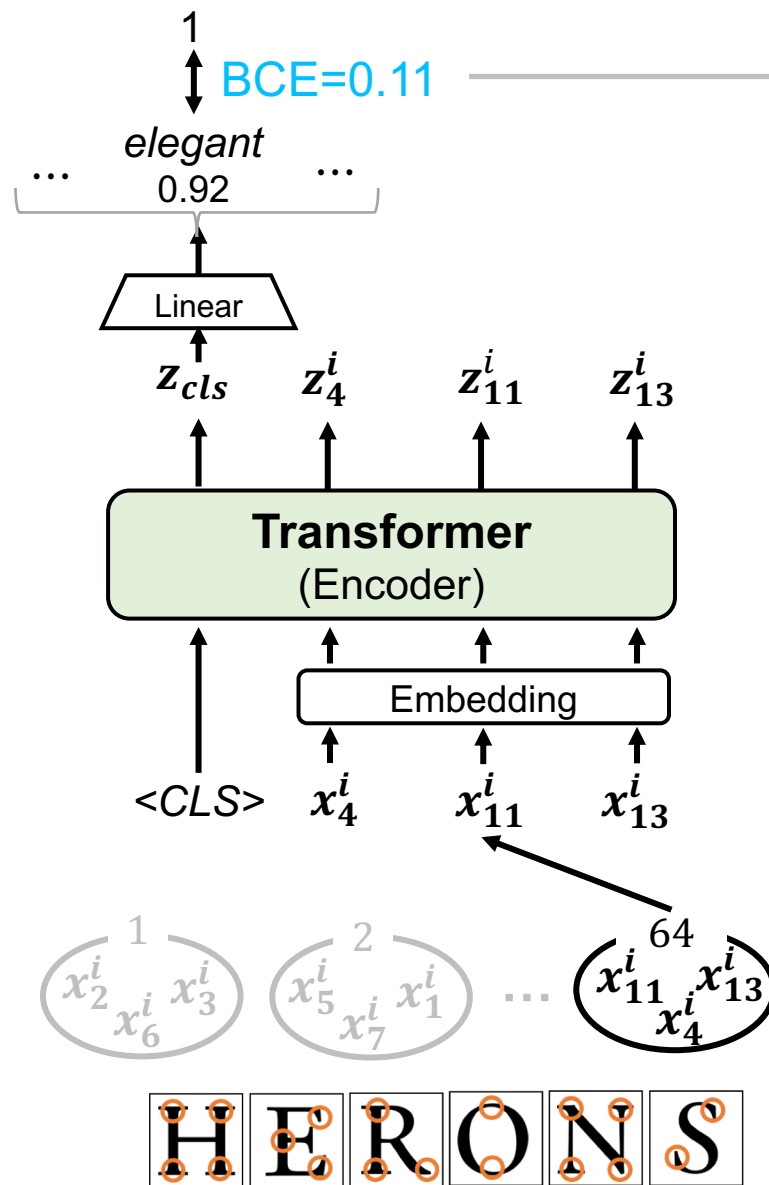


Group-Based Occlusion Sensitivity

Example : *elegant*

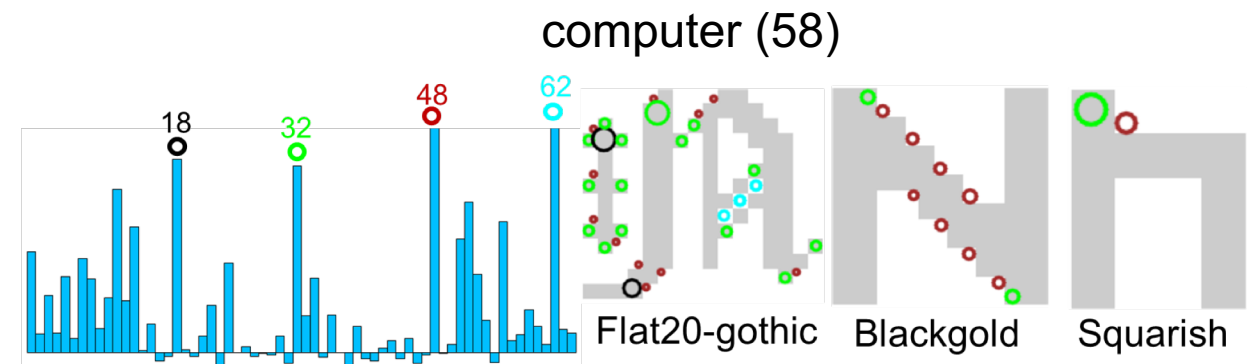
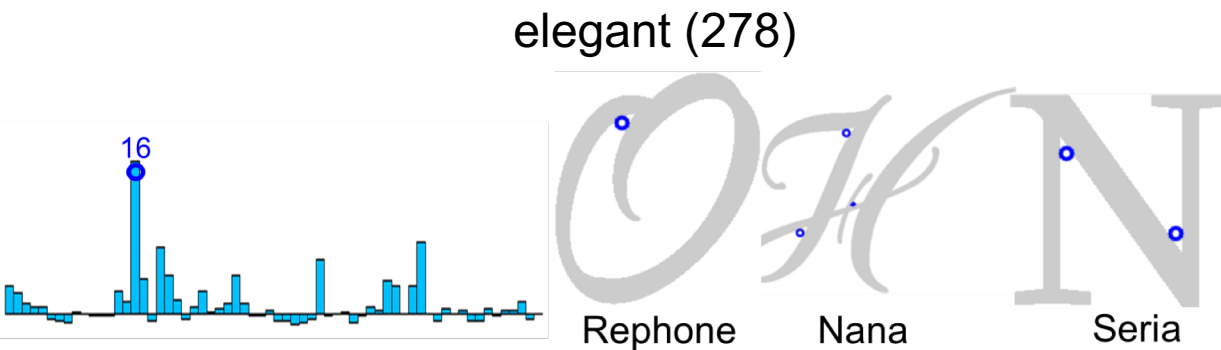
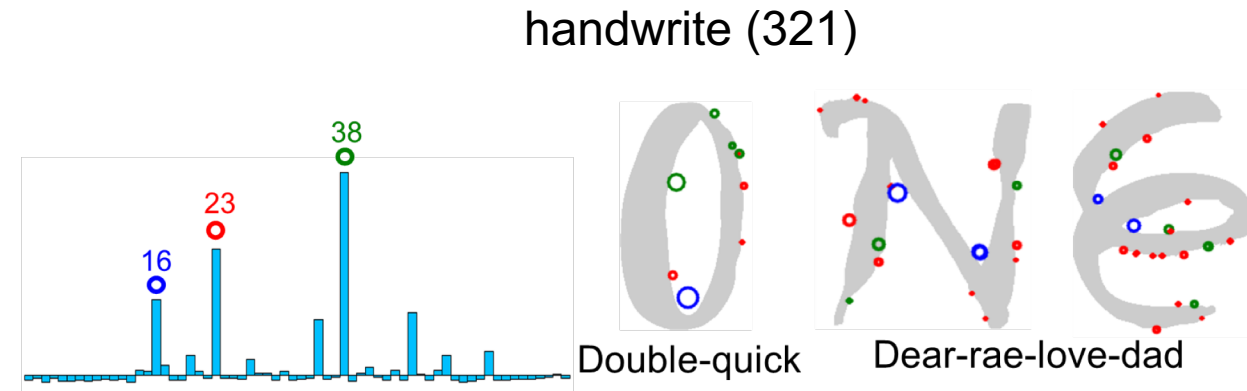
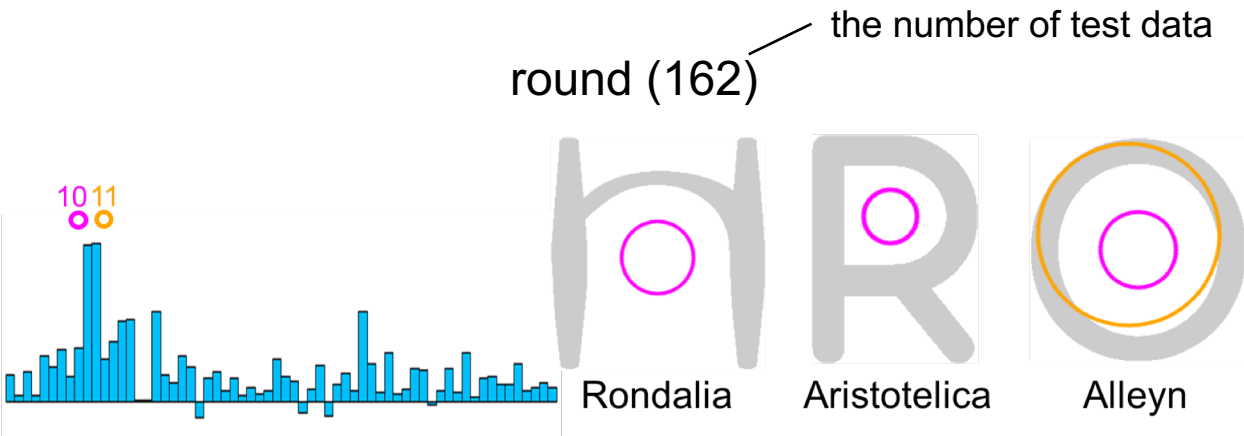


Grouped by clusters to which the SIFT vector belongs



Important Local Shapes for Specific Impressions

Create weighted histogram of the shapes that are important for based on **specific impressions**



Contribution Visualization of Inputs by Integrated Gradients [*]

$$\text{Integrated Gradients(IG)} = (\mathbf{x} - \mathbf{x}') \int_0^1 \nabla F(\mathbf{x}' + \alpha(\mathbf{x} - \mathbf{x}')) d\alpha$$

$\mathbf{x} \in \mathbf{R}^d$: input vector (SIFT vector)

$\mathbf{x}' \in \mathbf{R}^d$: baseline (0 vector)

$F : \mathbf{R}^d \rightarrow [0, 1]$ that represents a deep network (Transformer)

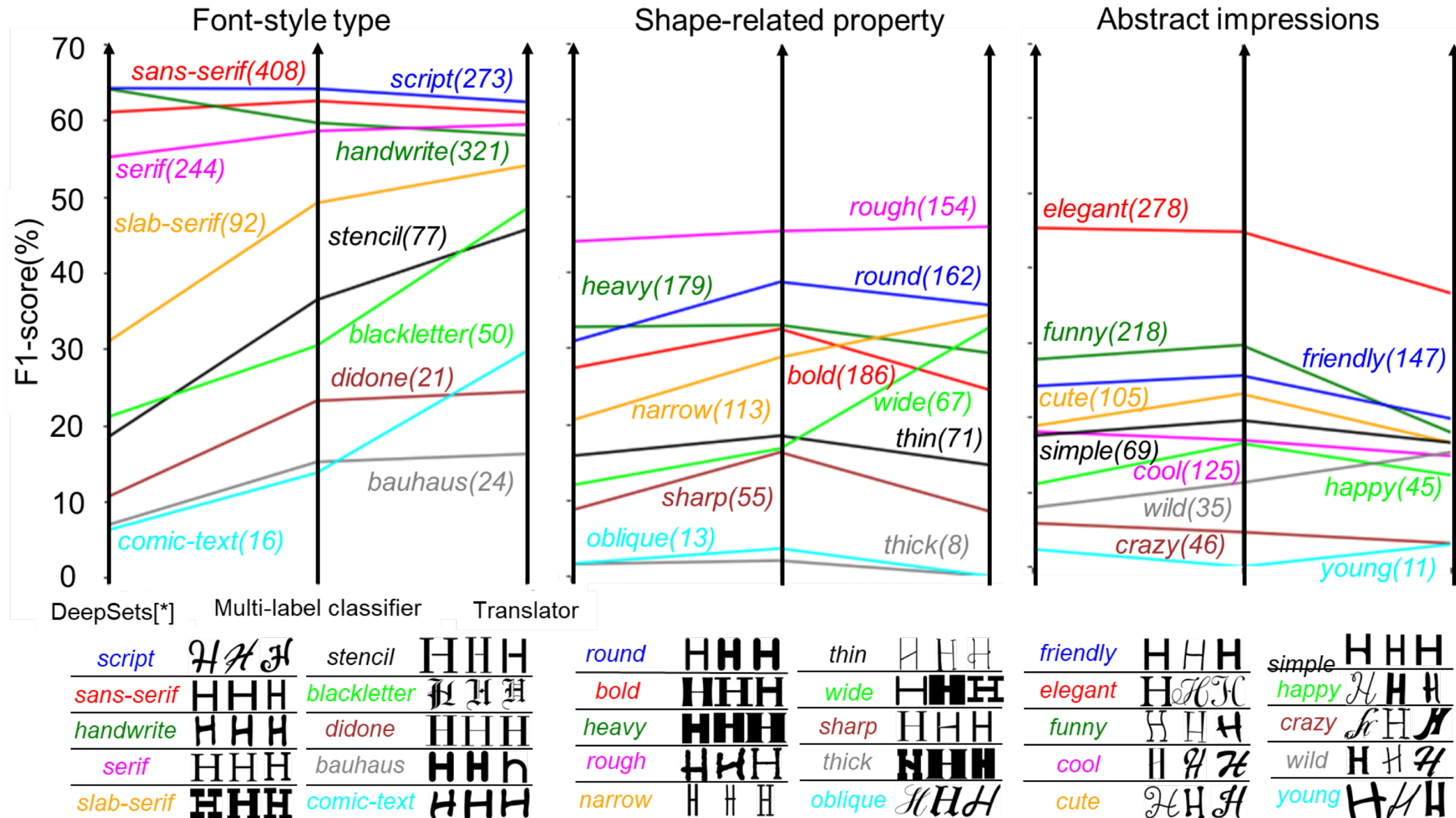


Contribution Map (by IG)

Translation Result

serif	0.32	0.12	...	0.15
elegant	-0.23	0.4	...	0.02
script	-0.11	-0.11	...	0.32
<EOS>	0.01	-0.24	...	-0.12
	\mathbf{x}_1	\mathbf{x}_2	...	\mathbf{x}_N

Comparison of F1-score by 3 methods



[*] Zaheer, M, et al., "Deep Sets", NeurIPS (2017)

Metrics : F1-score

- Binary classification is performed for each impression, and the F1-score is calculated for each impression.
- Average F1-score for each impression

	Font1	Font2	Font3	Font4
Label	[0, 1, 1, 1]	[1, 0, 1, 0]	[0, 1, 1, 0]	[1, 1, 1, 1]
Estimation	[1, 0, 1, 1]	[1, 1, 1, 1]	[1, 0, 0, 1]	[1, 0, 1, 0]

Impression1

Impression1

	EP	EN
GTP	2	0
GTN	2	0

Impression2

	EP	EN
GTP	0	3
GTN	1	0

Impression3

	EP	EN
GTP	3	1
GTN	0	0

Impression\$

	EP	EN
GTP	1	1
GTN	2	0

Metrics : Average Precision (AP)

Ranking of all fonts with impression X as label Prediction probability $\{r_1, r_2 \dots r_H\}$
 H is the number of fonts with impression X as a label

$$AP = \frac{1}{H} \sum_{h=1}^H \frac{h}{r_h}$$

Ex. : decorative

Rank	1	2	3	4	5	6	...	n
Probability	0.99	0.91	0.83	0.74	0.71	0.7		0.001
Label or Not	TRUE	FALSE	TRUE	FALSE	FALSE	TRUE		FALSE

$$AP_{decorative} = \frac{1}{\text{Sum of True}} \left(\frac{1}{1} + \frac{2}{3} + \frac{3}{6} + \dots \right)$$

Contribution Visualization of Inputs by Integrated Gradients [*]

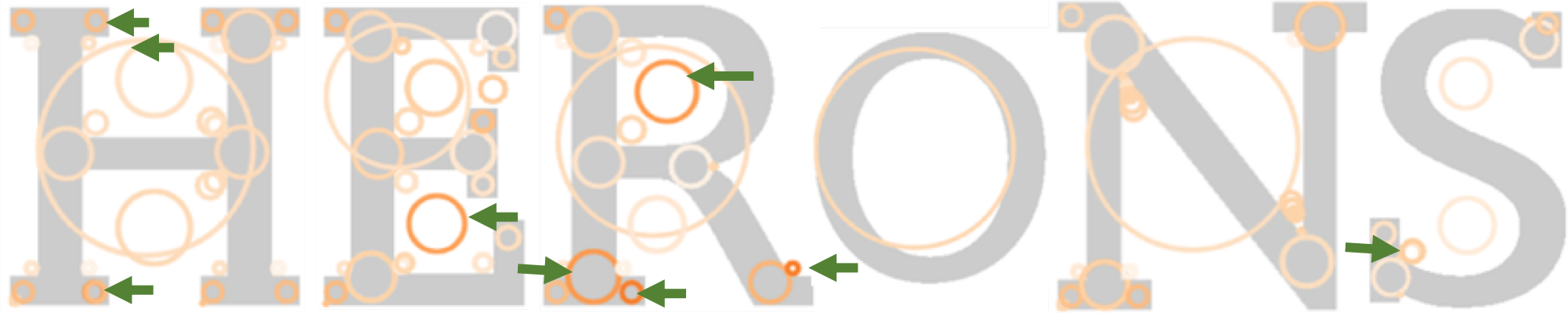
Thicker orange color shape are more important

The shapes of **serif-part** are deeply visualized

Translation Result

serif

Slam-normal



slab-serif



The shapes of **right-angle part** are deeply visualized



Questions

- ~~入出力の順序は？~~
- ~~アプリケーションは？~~
- ~~最後にチラッと saying いたマルチラベル認識の結果は~~
- comparative studyは？
 - そもそも類例が少ないのだが自分たちがやった範囲では
- ~~SIFTじゃないとだめなのか？~~
 - ~~SURFのやつを出す~~
- ~~どれぐらいの性能ができれば使えるのか？~~
- ~~得意な印象語と不得意な印象語は？~~