		Product Review Domain						
	Bags	KBs	Boots	B/T	TVs	VCs		
β	0.01	0.02	0.01	0.02	0.02	0.02		
I	6	4	6	8	6	4		
$\tau$	1e-3	1e-5	1e-5	1e-4	1e-3	1e-6		
$\lambda$	5	3000	10	5	10	3000		
$d_1$	2	128	16	16	16	128		
$d_2$	128	32	16	32	16	4		
$d_3$	8	64	8	16	16	64		

Table 1: Hyper-parameter settings for the product review.

		Restaurant Review Domain							
	En	Sp	Fr	Ru	Du	Tur			
β	0.01	0.02	0.02	0.02	0.01	0.01			
I	4	4	6	2	4	2			
$\tau$	1e-3	1e-5	1e-6	1e-3	1e-3	1e-5			
$\lambda$	100	100	10	1000	100	1			
$d_1$	16	16	32	16	32	2			
$d_2$	32	64	16	32	32	32			
$d_3$	128	128	64	32	64	32			

Table 2: Hyper-parameter settings for the restaurant review datasets.

# **APPENDIX**

We provide more information on datasets (Section 1.1) and implementation details (Section 1.2). Besides, we report more detailed experimental results (Section 1.3), and parameter sensitivity analysis (Section 1.4).

#### **Datasets**

During training, segment aspect labels (9-class for product reviews and 12-class for restaurant reviews) are not available but provided during validation and test. For instance, in the Laptop Bags domain, the review segments' aspects could be Compartments, Customer Service, Looks, or Price. Note that a general aspect is assigned if the segment doesn't discuss any specific aspects. For each domain, we train our model on a training set only with seed words G via the teacher. For the aspect seed words, we follow Angelidis and Lapata (2018); Karamanolakis et al. (2019) to use the same 30 seed words for two datasets. Besides, for two datasets, we follow Angelidis and Lapata (2018); Karamanolakis et al. (2019) to do data preprocessing, such as removing stop-words.

For the Amazon review dataset, the reviews of each domain are already segmented by (Angelidis and Lapata 2018), where they use a Rhetorical Structure Theory parser (Feng and Hirst 2012) to segment reviews into elementary discourse units (EDUs). Across domains, the average numbers of training, validation, and test segments are around 1 million, 700 segments, respectively. For restaurant reviews, the reviews of each language are already segmented into sentences. Across languages, the average number of training and test segments is around 2500 and 800 segments, respectively.

### **Implementation details**

For HDAE, the details hyper-parameter settings are given in Table 1 and 2, which are determined by optimizing on a validation set. We also provide the parameter sensitivity experiment of latent semantic modeling distance  $d_1$ ,  $d_2$ , and

		Restaurant Review Domain							
Model	En	Sp	Fr	Ru	Du	Tur			
ABAE	35.8	-	-	-	-	-			
SSCL-BT*	47.3	-	-	-	-	-			
LDA-AR	28.5	17.7	13.1	14.8	25.9	27.7			
MATE	41.0	24.9	17.8	18.4	36.1	39.0			
MATE-UW	40.3	18.3	19.2	21.8	31.5	25.2			
TS-Teacher	44.9	41.8	34.1	54.4	40.7	30.2			
TS-ATT	47.8	41.7	32.4	59.0	42.1	42.3			
TS-BT	51.8	42.0	39.2	58.0	43.0	45.0			
HDAE	57.9	65.7	48.6	62.9	57.2	50.8			

Table 3: Micro-averaged F1 reported for 12-class sentencelevel aspect detection in restaurant reviews.

au	1e-1	1e-2	1e-3	1e-4	1e-5	1e-6
Bags	67.1	67.5	68.8	67.8	67.8	68.2
B/T	68.4	69.1	71.1	71.9	70.4	70.3
Boots	61.8	62.4	63.5	63.7	64.0	63.7
TVs	67.1 68.4 61.8 69.8	70.8	71.2	70.4	70.3	70.0

Table 4: Micro-averaged F1 of HDAE given grumbel-softmax temperature  $\tau$ .

 $d_3$ , Grumbel-Softmax temperature  $\tau$ , ratio of distillation objective  $\lambda$ , number of latent semantic I in section . The total number of negative examples  $k_n$  was set to 10. We followed the procedure in Angelidis and Lapata (2018) to set the 200-dimensional word embeddings for the Amazon product reviews and the 300-dimensional multilingual word2vec embeddings from Ruder, Ghaffari, and Breslin (2016) for restaurant reviews. For all models, the same 30 seed words were set per aspect. For HDAE, model parameters are optimized by using the Adam optimizer (Kingma and Ba 2014). For setting distillation objective, the teacher, a bag-of-word classifier, is implement, and we use iterative co-training to update each seed word's predictive quality. For TS-\*, we report the result from iterative co-training, and in each round, we divide the learning rate by 10. For SSCL-BT\*, we use code provided in<sup>1</sup> and conduct aspect mapping after training the teacher model. The smooth factor  $\lambda$  is set to 0.5 and temperature is set to 1. For all models, the learning rate was selected from  $[2 \times 10^{-4}, 1 \times 10^{-6}, 5 \times 10^{-7}, 5 \times 10^{-8},$  $1 \times 10^{-8}$ ]. Other hyperparameters were optimized according to validation results. For each model, we repeat each experiment 5 times and report the average test performance with the parameter configuration that achieves the best validation performance.

#### **Experimental Results**

we provide the performance the unsupervised based method ABAE and SSCL on English Restaurant review, as shown in Table 3

Then, we provide more results for seed word based approaches' inference performance per aspect and their corresponding embedding visualization on Bags, Bluetooth Headsets, Boots, Keyboards, Televisions, and Vacuums (VCs) datasets, shown in Figure 1, 2, 3, 4, 5, 6, respectively.

<sup>&</sup>lt;sup>1</sup>https://github.com/tshi04/AspDecSSCL

$\lambda$	5	10	100	500	1000	3000
Bags	68.8	68.1	67.5	67.0	66.9	66.0
B/T	71.1	70.8	69.0	67.9	67.2	63.3
Boots	62.4	63.7	62.8	62.4	61.2	60.7
TVs	70.1	71.2	70.3	70.1	69.8	69.4

Table 5: Micro-averaged F1 of HDAE given  $\lambda$ 

I	2	4	6	8
Bags	68.0	68.3	68.8	68.3
B/T	68.5	69.1	70.3	71.9
Boots	61.8	62.4	64.0	63.9
TVs	69.3	69.8	71.2	70.5

Table 6: Micro-averaged F1 of HDAE given I.

### **Parameter Sensitivity Analysis**

In this section, we provide more results for parameter sensitivity. Table 4 shows effects of the grumbel-softmax temperature  $\tau$  on the performance of HDAE. We find that our model achieves the best results when small  $\tau$  is set, suggesting that it is important to not to mix the latent semantic when predicting the segment's aspect. Table 5 shows effects of the distillation objective ratio  $\lambda$  on the performance of HDAE. It is found that HDAE achieves the best performance when  $\lambda$  is set to 5 or 10. From Table 6, we see effects of number of latent semantics I and HDAE achieves best result when large I is set. From Table 7, we see effects of  $\beta$  and HDAE achieves the best result when  $\beta$  is set to 0.01 or 0.02.

## Results in ground truth aspect labels ratios

In this section, we provide results for baselines with (HDAE, MATE, TS-ATT, TS-BT) or without (W2V, BERT) leveraging seed word in different ground truth aspect labels ratios. We find the proposed model HDAE can achieve best performance in different ground truth aspect labels ratios, suggesting the effectiveness of purposed hyperbolic disentangled based method. Besides, we notice that in the low aspect labels data ratios, seed word based approaches (MATE, TS-ATT, TS-BT, and HDAE) can achieve better, demonstrating that seed words can give useful guidance and assist models to improve aspect inference ability.

### References

Angelidis, S.; and Lapata, M. 2018. Summarizing opinions: Aspect extraction meets sentiment prediction and they are both weakly supervised. *arXiv* preprint arXiv:1808.08858.

Feng, V. W.; and Hirst, G. 2012. Text-level Discourse Parsing with Rich Linguistic Features. In *Proceedings of the 50th Annual Meeting of the Association for Computational Linguistics (Volume 1: Long Papers)*, 60–68. Jeju Island, Korea: Association for Computational Linguistics.

Karamanolakis, G.; Hsu, D.; Hsu, D.; and Gravano, L. 2019. Leveraging Just a Few Keywords for Fine-Grained Aspect Detection Through Weakly Supervised Co-Training. *CoRR*, abs/1909.00415.

Kingma, D. P.; and Ba, J. 2014. Adam: A Method for Stochastic Optimization. *arXiv e-prints*, arXiv:1412.6980.

β	0.005	0.01	0.02	0.05
Bags	67.8	68.0	67.7	66.6
B/T	70.5	71.2	71.9	71.3
Boots	63.1	64.0	63.7	63.2
TVs	68.1	69.8	70.5	70.0

Table 7: Micro-averaged F1 of HDAE given  $\beta$ 

	Re	Restaurant review domain (En)					
ratio r	10%	30%	50%	70%	100%		
W2V-Gold	16.3	33.0	38.6	46.7	58.8		
BERT-Gold	24.8	36.5	48.5	55.9	63.1		
MATE	43.8	46.7	50.4	54.3	60.1		
TS-ATT	48.5	50.6	53.2	57.7	61.1		
TS-BT	53.6	56.1	58.4	61.2	64.2		
HDAE	58.6	62.2	64.1	66.9	70.5		

Table 8: Micro-averaged F1 for 12-class sentence-level aspect detection in restaurant reviews in English with different ratios of training set r.

Ruder, S.; Ghaffari, P.; and Breslin, J. G. 2016. A Hierarchical Model of Reviews for Aspect-based Sentiment Analysis. *CoRR*, abs/1609.02745.

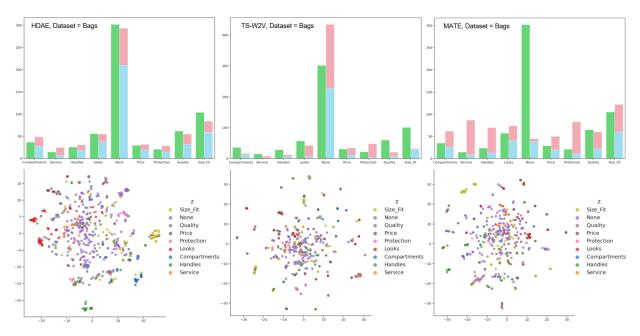


Figure 1: Inference performance per aspect of HDAE, TS-W2V, and MATE on Bags dataset. The following figure is segment vector t-SNE visualization of each model, where the different color of point represent different aspect.

	Restaurant review domain (Sp)						
ratio r	10%	30%	50%	70%	100%		
W2V-Gold	21.3	32.2	35.5	40.1	50.4		
BERT-Gold	29.2	39.1	42.2	46.0	51.6		
MATE	32.5	39.4	45.1	48.5	52.4		
TS-ATT	42.3	46.8	54.4	58.2	62.3		
TS-BT	44.0	46.5	52.6	60.7	66.6		
HDAE	66.8	68.1	70.0	72.1	72.5		

Table 9: Micro-averaged F1 for 12-class sentence-level aspect detection in restaurant reviews in Spanish with different ratios of training set r.

	Restaurant review domain (Ru)						
ratio r	10%	30%	50%	70%	100%		
W2V-Gold	28.5	32.5	43.5	50.0	55.7		
BERT-Gold	23.5	31.5	41.3	47.9	55.3		
MATE	22.6	30.3	36.8	44.3	51.2		
TS-ATT	58.8	59.1	59.8	62.1	65.5		
TS-BT	59.5	61.3	62.1	65.5	67.4		
HDAE	61.3	65.0	67.8	71.5	76.8		

Table 11: Micro-averaged F1 for 12-class sentence-level aspect detection in restaurant reviews in Russian with different ratios of training set r.

	Re	Restaurant review domain (Fr)						
ratio r	10%	30%	50%	70%	100%			
W2V-Gold	21.2	29.9	37.1	43.1	50.4			
BERT-Gold	20.2	24.8	33.0	40.9	50.6			
MATE	28.7	34.8	40.5	45.2	48.1			
TS-ATT	32.8	38.1	44.1	46.6	50.1			
TS-BT	43.0	44.9	46.5	48.0	53.0			
HDAE	48.7	51.8	54.9	60.8	65.4			

Table 10: Micro-averaged F1 for 12-class sentence-level aspect detection in restaurant reviews in French with different ratios of training set r.

	Restaurant review domain (Du)						
ratio r	10%	30%	50%	70%	100%		
W2V-Gold	24.4	32.8	42.4	47.3	51.4		
BERT-Gold	28.1	40.8	47.5	51.6	53.5		
MATE	38.8	40.9	46.2	48.4	52.0		
TS-ATT	43.9	45.8	51.8	53.5	55.4		
TS-BT	45.4	47.1	50.4	53.6	57.6		
HDAE	58.5	62.5	68.3	72.1	73.8		

Table 12: Micro-averaged F1 for 12-class sentence-level aspect detection in restaurant reviews in Dutch with different ratios of training set r.

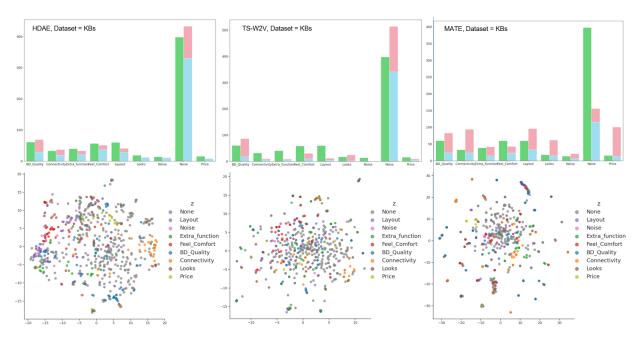


Figure 2: Inference performance per aspect of HDAE, TS-W2V, and MATE on Keyboards dataset. The following figure is segment vector t-SNE visualization of each model, where the different color of point represent different aspect.

	Restaurant review domain (Tur)				
ratio r	10%	30%	50%	70%	100%
W2V-Gold	28.6	37.2	42.3	50.8	55.7
BERT-Gold	31.5	39.0	45.6	52.3	56.5
MATE	41.3	44.9	47.1	49.9	53.0
TS-ATT	45.9	47.5	49.7	52.8	55.5
TS-BT	45.5	48.7	52.6	54.3	57.6
HDAE	49.8	52.4	56.9	60.1	65.4

Table 13: Micro-averaged F1 for 12-class sentence-level aspect detection in restaurant reviews in Turkish with different ratios of training set r.

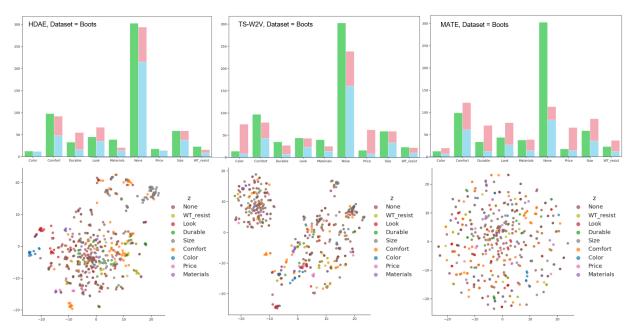


Figure 3: Inference performance per aspect of HDAE, TS-W2V, and MATE on Boots dataset. The following figure is segment vector t-SNE visualization of each model, where the different color of point represent different aspect.

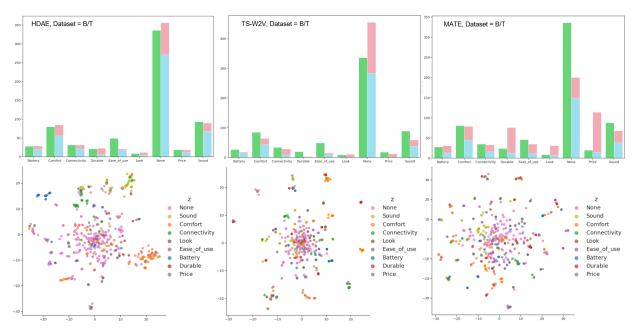


Figure 4: Inference performance per aspect of HDAE, TS-W2V, and MATE on Bluetooth Headsets dataset. The following figure is segment vector t-SNE visualization of each model, where the different color of point represent different aspect.

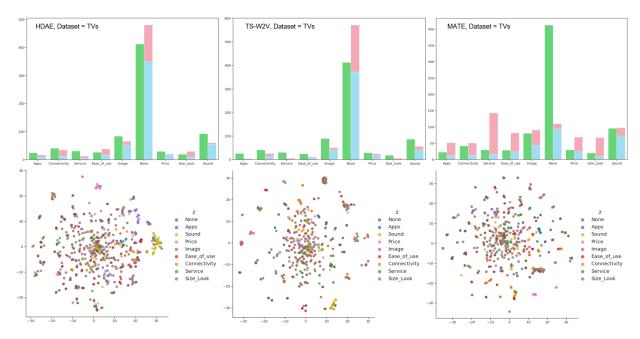


Figure 5: Inference performance per aspect of HDAE, TS-W2V, and MATE on Televisions dataset. The following figure is segment vector t-SNE visualization of each model, where the different color of point represent different aspect.

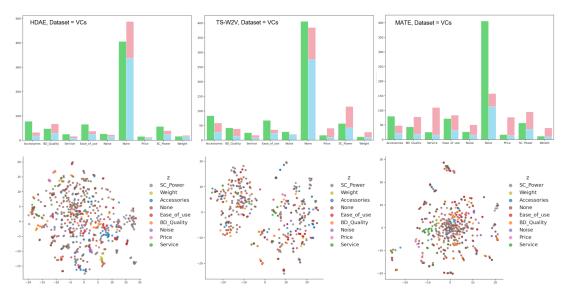


Figure 6: Inference performance per aspect of HDAE, TS-W2V, and MATE on Vacuums dataset. The following figure is segment vector t-SNE visualization of each model, where the different color of point represent different aspect.

Do not purchase.	quiet, action, sound, quieter, know, make	GT: Noise	
HDAE: General X	MATE: General	TS-W2V: General X	
Which died.	GT: General		
	, purchase, using, unit, star, microsoft, mouse	G1. General	
HDAE: Build Quality 🗡	MATE: Build Quality X	TS-W2V: Build Quality 🗡	
Except the keyboard was one of	f those high keyed, clackety-clunkety types.	GT: General	
Seed Words: think, recommend	, purchase, using, unit, star, microsoft, mouse		
HDAE: General ✓	MATE: General ✓	TS-W2V: Comfort X	
I really liked the look of it.		GT: Looks	
	l, appearance, little, attractive, beautiful		
HDAE: Looks ✓	MATE: General X	TS-W2V: Looks ✓	
I liked the feel of the keys.		GT: Comfort	
	mushy, key, like, keyboard, good, perfect		
HDAE: Comfort ✓	MATE: General 🗡	TS-W2V: Looks 🗡	
But it has all the buttons to inter		GT: Extra functionality	
Seed Words: buttons, light, pen-	cil, volume, power, feature, bright, mute, handy, low, o	lark	
HDAE: Extra functionality 🗸	MATE: Extra functionality ✓	TS-W2V: Extra functionality 🗸	
It is quiet.		GT: Noise	
Seed Words: loud, noise, noisy,	red, action, sound, quieter, know, make		
HDAE: Noise ✓	MATE: Noise ✓	TS-W2V: Noise ✓	
	difficult for me to use, with keys like the backspace.	GT: Layout	
Seed Words: key, delete, backsp	pace, size, layout, end, insert, home, bar, perfect, space		
HDAE: Layout √	MATE: General 🗡	TS-W2V: Layout ✓	
And doesn't depress at times	And doesn't depress at times GT: Build Quality		
Seed Words: working, build, sto	opped, quality, month, spacebar, stuck, left, plastic, kir	nd, died	
HDAE: Build Quality 🗸	MATE: Build Quality √	TS-W2V: General X	
And the key for the "t" is alrea		GT: Build Quality	
Seed Words: working, build, sto	opped, quality, month, spacebar, stuck, left, plastic, kir	nd, died	
HDAE: Build Quality √	MATE: Layout 🗡	TS-W2V: Comfort 🗡	
Has a top row of quick link	-	GT: Extra functionality	
Seed Words: button, light, penc	il, volume, power, feature, bright, mute, handy, low, da	ark	
HDAE: Extra functionality 🗸	MATE: Extra functionality ✓	TS-W2V: Comfort X	
The keyboard is sleek and visua	ally appealing.	GT: Looks	
Seed Words: look, slim, origina	l, appearance, little, attractive, beautiful		
HDAE: Looks √	MATE: General 🗡	TS-W2V: General X	
That had popped off.	at had popped off.  GT: Build Quality		
Seed Words: working, build, stopped, quality, month, spacebar, stuck, left, plastic, kind, died			
HDAE: Build Quality ✓	MATE: Build Quality √	TS-W2V: General X	
It is very responsive. GT: Comfort			
Seed Words: feel, comfortable, mushy, key, like, keyboard, good, perfect, press, wrist, action, shallow, smooth			
HDAE: Comfort ✓	MATE: Connectivity X	TS-W2V: General X	

Table 14: Comparison of predictions on sample Keyboards product review segments between HDAE, MATE, and TS-W2V. For each review segment, the ground true (GT) aspect and its corresponding seed words are provided.

But may be unsightly.		GT: Look	
	great, fringe, great, fringe, style, color	r, love, design, going, unattractive, attractive	
HDAE: Durability 🗡	MATE: Size X	TS-W2V: General X	
However, I bought them in February.  GT: General			
Seed Words: return, pair, though	nt, year, boot, high, kind, merrell, excit	ted, wore, disappointed, buy, problem	
HDAE: General ✓	MATE: Price 🗡	TS-W2V: General ✓	
I ordered these a half-size up	GT: Size		
Seed Words: size, ordered, half,	order, big, little, bigger, usually, narro	ow, normally, width	
HDAE: Size ✓	MATE: General 🗡	TS-W2V: Size ✓	
Nice looking shoes.		GT: Look	
Seed Words: cute, look, looked,	great, fringe, great, fringe, style, color	r, love, design, going, unattractive, attractive	
HDAE: Look √	MATE: Look ✓	TS-W2V: Color 🗡	
Hurt after 30 minutes		GT: Comfort	
		t, break, wear, able, extreme, walking, sock	
HDAE: Comfort √	MATE: Comfort ✓	TS-W2V: Comfort ✓	
The material is very thin and fli		GT: Materials	
		buckle, suede, true, rubber, advertised	
HDAE: Materials ✓	MATE: Materials 🗸	TS-W2V: Materials ✓	
Which helps in the Maine winter		GT: Weather resistance	
		d, puddle, warm, got, chicago, protect, suppose	
HDAE: Weather resistance ✓	MATE: General 🗡	TS-W2V: General X	
It was a washed purplish but no	t a true purple	GT: Color	
Seed Words: color, love, style, u	inbelievably, gorgeous, blue, favorite,		
HDAE: Color ✓	MATE: Size 🗡	TS-W2V: General 🗡	
I wanted a true purple boot.		GT: Color	
Seed Words: color, love, style, u	inbelievably, gorgeous, blue, favorite,		
HDAE: Color ✓	MATE: General 🗡	TS-W2V: Price 🗡	
That I can easily slip on and not		GT: Comfort	
	•	, break, wear, able, extreme, walking, sock	
HDAE: Comfort ✓	MATE: Comfort ✓	TS-W2V: General X	
seem pretty solid to me. GT: Durability			
	e, buckle, appear, heel, long, time, brea		
HDAE: Durability ✓	MATE: Durability 🗸	TS-W2V: General 🗡	
	ting your foot into the things though.	GT: Comfort	
Seed Words: comfortable, fit, fo		, break, wear, able, extreme, walking, sock	
HDAE: Comfort ✓	MATE: General 🗡	TS-W2V: Weather Resistance X	
It would stay tied as well.  GT: Durability			
Seed Words: rubber, quality, use, buckle, appear, heel, long, time, breaking, enjoy, provide, protection, clearly, cheaply			
HDAE: Durability ✓	MATE: Look 🗡	TS-W2V: General 🗡	
You need to design these with a zipper.  GT: Materials  Seed Words: leather, inside, fringe, material, heel, cheaper, soft, sole, buckle, suede, true, rubber, advertised			
,	<del>-</del>		
HDAE: Materials ✓	MATE: Materials 🗸	TS-W2V: General X	

Table 15: Comparison of predictions on sample Boots product review segments between HDAE, MATE, and TS-W2V. For each review segment, the ground true (GT) aspect and its corresponding seed words are provided.

Plus, a lot of the more expensi	ve cases aren't as bright as this one.	GT: Looks	
	stylish, looked, pretty, lime, green, fashion	nable, picture, awesome, good, great	
HDAE: General X	MATE: Price X	TS-W2V: General X	
And we bought two of this sleeve.		GT: General	
Seed Words: recommend, bou		say, backpack, weight, case, new, purchased	
HDAE: General ✓	MATE: Size fit 🗡	TS-W2V: General ✓	
My only complaint is the interior color		GT: Looks	
Seed Words: look, color, pink,	stylish, looked, pretty, lime, green, fashio	nable, picture, awesome, good, great	
HDAE: Looks ✓	MATE: Looks ✓	TS-W2V: Looks ✓	
About what size laptop fits in		GT: Size fit	
Seed Words: fit, size, macbook		, inch, laptop, perfectly, 13, cable, folder, notebook	
HDAE: Size fit ✓	MATE: Looks 🗡	TS-W2V: Size fit ✓	
I can barely sling it over my sl		GT: Handles	
Seed Words: strap, handle, sho	oulder, broke, later, month, hand, comfortal	ble, plastic, wear, tear, ripped	
HDAE: Handles ✓	MATE: Handles ✓	TS-W2V: Handles ✓	
It just looks kinda funny. GT: Looks			
Seed Words: look, color, pink,	stylish, looked, pretty, lime, green, fashion	nable, picture, awesome, good, great, design, fre, cell, styling	
HDAE: Looks ✓	MATE: Looks ✓	TS-W2V: Looks ✓	
Starting to fall apart.		GT: Quality	
Seed Words: quality, material,		, wear, tear, zipper, chemical, started, excellent, terrible	
HDAE: Quality ✓	MATE: Looks 🗡	TS-W2V: General 🗡	
And part of it chipped off		GT: Quality	
		, wear, tear, zipper, chemical, started, excellent, terrible	
HDAE: Quality ✓	MATE: Handles X	TS-W2V: General 🗡	
Very fast shipping.		GT: Customer service	
Seed Words: hassle, return, un		rrived, promptly, gift, easy, getting, returned, week, amazon	
HDAE: Customer service ✓	MATE: General service X	TS-W2V: General 🗡	
The case is so small		GT: Size fit	
Seed Words: fit, size, macbook	k, big, space, air, slightly, lot, bulk, perfect,	, inch, laptop, perfectly, 13, cable, folder	
HDAE: Size fit ✓	MATE: Protection X	TS-W2V: General X	
Which is a good deal		GT: Price	
Seed Words: price, worth, pay, good, great, requested, believe, send, 100, suppose, buck, based, reasonable, inexpensive			
HDAE: Price ✓	MATE: Price ✓	TS-W2V: Looks 🗡	
On the plus side it does have a lot of nooks and crannies to put stuff in.  GT: Compartments			
Seed Words: pocket, cable, compartment, outside, lot, wish, wallet, connector, space, power, pen, folder, charger, flap, mouse, nice			
HDAE: Compartments ✓	MATE: Size Fit 🗡	TS-W2V: Size Fit 🗡	

Table 16: Comparison of predictions on sample Bags product review segments between HDAE, MATE, and TS-W2V. For each review segment, the ground true (GT) aspect and its corresponding seed words are provided.

, trace, beautiful			
, trace, beautiful			
, trace, beautiful			
dio, play, stopped			
uts, cycle, signal			
Seed Words: picture, color, quality, back, bright, nice, clear, look, excellent, crisp, screen, right, dead, pixel, trace, beautiful			
Seed Words: netflix, user, file, hulu, apps, watch, flash, internet, smart, video  HDAE: Apps Interface ✓ MATE: Ease of Use ✓ TS-W2V: Sound ✓ TS-W2V: Sound ✓			
, trace, beautiful			

Table 17: Comparison of predictions on sample Televisions product review segments between HDAE, MATE, and TS-W2V. For each review segment, the ground true (GT) aspect and its corresponding seed words are provided.

To roll the windows up.		GT: Sound	
Seed Words: sound, qualit	y, hear, noise, volume, end, audio, good, voice, conversation, people, t	told, clearly, background	
HDAE: General X	MATE: General X	TS-W2V: General X	
The volume is GREAT!		GT: Sound	
Seed Words: sound, qualit	y, hear, noise, volume, end, audio, good, voice, conversation, people, t	told, clearly, background	
HDAE: Sound ✓	MATE: Price 🗡	TS-W2V: Look 🗡	
Very easy to put on.		GT: Ease Of Use	
Seed Words: easy, control	, button, simple, setup, mic, phone, colored, flashing, confusing, make	, convenient	
HDAE: Ease of use ✓	MATE: Durability 🗡	TS-W2V: Ease Of Use ✓	
I have had mine for about		GT: Durability	
	month, durable, sweat, charging, turn, second, expected, holding, stand		
HDAE: Durability √	MATE: General 🗡	TS-W2V: Durability ✓	
They hurts your ears.		GT: Comfort	
Seed Words: ear, fit, comf	ortable, bud, feel, jaybird, comfortably, lightweight, wear, readjusting,	hard	
HDAE: Comfort ✓	MATE: Comfort √	TS-W2V: Comfort ✓	
	h phone, a Razr, and a Blackberry with excelent results in every case	GT: Connectivity	
Seed Words: phone, range	, easy, easily, galaxy, foot, connection, optimal, zone, cut, paired, link	, connects	
HDAE: Connectivity ✓	MATE: Durability 🗡	TS-W2V: Connectivity ✓	
As for the ear pad.		GT: Comfort	
Seed Words: ear, fit, comf	ortable, bud, feel, jaybird, comfortably, lightweight, wear, readjusting,	hard	
HDAE: Comfort ✓	MATE: Comfort √	TS-W2V: Comfort ✓	
Strengths: Sweat and wat		GT: Durability	
	month, durable, sweat, charging, turn, second, expected, holding, stand	lard, comment	
HDAE: Durability √	MATE: Durability ✓	TS-W2V: Durability ✓	
It is portable and sweat pr		GT: Durability	
	month, durable, sweat, charging, turn, second, expected, holding, stand	lard, comment	
HDAE: Durability ✓	MATE: Durability 🗸	TS-W2V: Durability ✓	
I do like the design of it ar		GT: Look	
Seed Words: look, appeara	ance, hs850, nice, earpiece, h700, smaller, feel, fine, bluetooth, great, g	good, headset	
HDAE: Look ✓	MATE: Durability 🗡	TS-W2V: General 🗡	
To roll the windows up		GT: Sound	
	y, hear, noise, volume, end, audio, good, voice, conversation, people, t		
HDAE: Sound ✓	MATE: Comfort X	TS-W2V: General 🗡	
The style is very cool and		GT: Look	
Seed Words: look, appeara	ance, hs850, nice, earpiece, h700, smaller, feel, fine, bluetooth, great, g	good, headset	
HDAE: Look ✓	MATE: General 🗡	TS-W2V: Look ✓	
On the plus side, the call quality is good.  GT: Sound			
Seed Words: sound, quality, hear, noise, volume, end, audio, good, voice, conversation, people, told, clearly, background			
HDAE: Sound ✓	MATE: Price X	TS-W2V: Look 🗡	
It starting to hurt me. GT: Comfort			
Seed Words: ear, fit, comfortable, bud, feel, jaybird, comfortably, lightweight, wear, readjusting, hard			
HDAE: Comfort ✓	MATE: Comfort ✓	TS-W2V: General X	

Table 18: Comparison of predictions on sample Bluetooth Headsets product review segments between HDAE, MATE, and TS-W2V. For each review segment, the ground true (GT) aspect and its corresponding seed words are provided.

Eventually I started listenin		GT: General		
Seed Words: vac, cleaner, vacuum, buy, bought, new, better, year, recommend, product, owned, review, gave, away, kenmore, dyson				
HDAE: Build Quality 🗡	MATE: Build Quality 🗡	TS-W2V: General ✓		
	It is easy to move because of the adjustable wheels on the side of the brush  GT: Ease of use			
Seed Words: easy, cord, pus		vkward, crevice, constantly, bog, impossible, short		
HDAE: Ease of use ✓	MATE: Accessories X	TS-W2V: Ease of use ✓		
Too bulky, cord is unusually	stiff and tangles are impossible to remove.	GT: Ease of use		
		wkward, crevice, constantly, bog, impossible, short		
HDAE: Ease of use ✓	MATE: Build Quality 🗡	TS-W2V: Ease of use ✓		
It is so easy to maneuver.		GT: Ease of Use		
Seed Words: easy, cord, pus		vkward, crevice, constantly, bog, impossible, short		
HDAE: Ease of use ✓	MATE: Weight 🗡	TS-W2V: Ease of Use ✓		
Because it was so loud.		GT: Noise		
Seed Words: quiet, noisy, lo	oud, powerful, noise, louder, ear, loudest, light, in	credibly, deafening, seriously, actually		
HDAE: Noise ✓	MATE: Noise ✓	TS-W2V: Noise ✓		
	And have powerful suction.  GT: Suction Power			
	powerful, power, good, hair, carpet, such, quiet, r	really, performs, dirt, tile, ok		
HDAE: Suction Power ✓	MATE: Suction Power ✓	TS-W2V: Suction Power ✓		
While the suction is very go		GT: Suction Power		
Seed Words: suction, pick,	powerful, power, good, hair, carpet, such, quiet, r	really, performs, dirt, tile, ok		
HDAE: Suction Power ✓	MATE: Noise 🗡	TS-W2V: Suction Power ✓		
	d tearing the plastic parts on the brush	GT: Build Quality		
Seed Words: belt, broke, tur	rn, working, burning, electrical, built, stop, month	n, roller, time, minute		
HDAE: Build Quality ✓	MATE: Accessories X	TS-W2V: Build Quality ✓		
The sides wrapped with pro		GT: Build Quality		
Seed Words: belt, broke, turn, working, burning, electrical, built, stop, month, roller, time, minute, problem				
HDAE: Build Quality ✓	MATE: Noise 🗡	TS-W2V: General X		
Then the engine completely stopped vacuuming.  GT: Build Quality				
Seed Words: belt, broke, turn, working, burning, electrical, built, stop, month, roller, time, minute, problem, brush, design				
HDAE: Build Quality 🗸	MATE: Suction Power X	TS-W2V: Ease of Use X		
A small, light-weight applia	nnce that can do a big job.	GT: Weight		
Seed Words: light, weight, lightweight, heavy, size, compact, maneuver, guess, quiet, quite, probably				
HDAE: Weight √	MATE: Price 🗡	TS-W2V: Suction Power X		
You have to hold it at a very uncomfortable angle  GT: Ease of Use				
Seed Words: easy, cord, push, corner, vacuuming, pile, maneuver, nozzle, awkward, crevice, constantly, bog, impossible, short				
HDAE: Ease of Use ✓	MATE: Ease of Use ✓	TS-W2V: General X		
The tools were previously stored inside the canister GT: Accessories				
Seed Words: filter, brush, attachment, roll, turbo, easily, expensive, wide, turn, bag, replacing, typical, hepa				
HDAE: Accessories ✓	MATE: Build Quality 🗡	TS-W2V: General X		

Table 19: Comparison of predictions on sample Vacuums product review segments between HDAE, MATE, and TS-W2V. For each review segment, the ground true (GT) aspect and its corresponding seed words are provided.