

# TEXT MINING

## Text Mining Application to Internet Shopping Mall Customers' Reviews

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### 字例

「如古為柿。」을為屬二。如○

赤春稻。○為大豆。

「如古為流澌。」이為屬二。如○

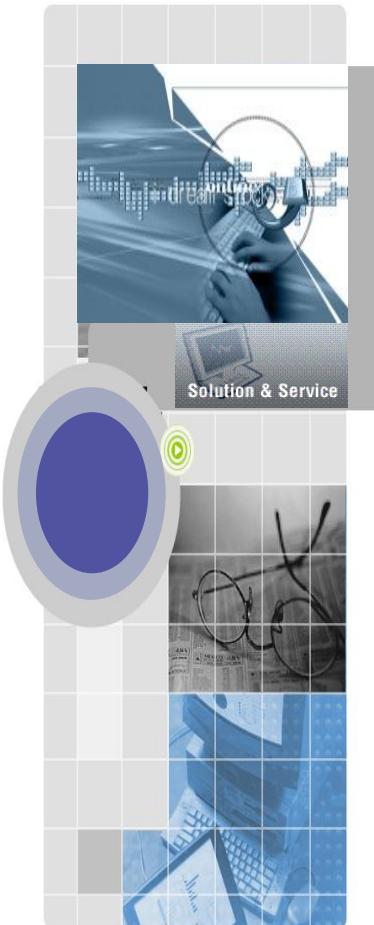
「如古為蘭。」이為屬二。如○

「如古為獐。」남為屬二。如○

「如古為蜂。」비為屬二。如○

「如古為荔。」마為屬二。如○

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## I. Introduction

# TEXT MINING

## I. Introduction

The diagram illustrates the concept of Text Mining by showing four main categories of text data:

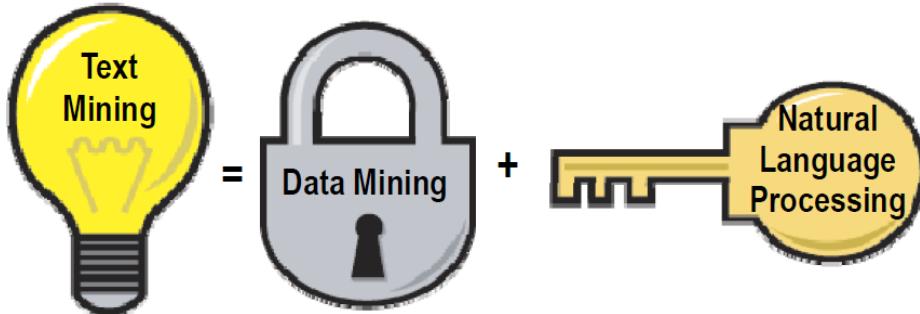
- Research papers**: Represented by a screenshot of the RISS (RISS) website, which contains academic publications and search functions.
- Internet documents**: Represented by a screenshot of the NAVER 지식IN (NAVER Knowledge IN) platform, showing various types of online Q&A and informational posts.
- News**: Represented by a screenshot of KBS NEWS, displaying news articles and media content.
- Reviews of customers**: Represented by a screenshot of a product review section, showing user feedback and ratings.

A large central blue box labeled "Text data" covers the area between the four examples, indicating that all these forms of communication are processed and analyzed by text mining techniques.

## II. Text Mining

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### Text Mining



Statistical technique and machine learning algorithms are used for finding information from big sized text data

TM  
Application of

- Information Retrieval
- Natural Language Processing
- Information Extraction

### Document classification

- Decision Tree
- Bayesian Classification
- Nearest Neighbor Classifier
- Support Vector Machine

### Document clustering

- Hierarchical Cluster Analysis
- K-means Clustering
- SOM(Self Organizing Maps)
- Expectation-Maximization Clustering

## 1. structurizing

### Term-document frequency matrix

	Doc 1	...	Doc j	...	Doc n
Term1	$tf_{11}$	...	$tf_{1j}$	...	$tf_{1n}$
:	:	:	:		:
Term i	$tf_{i1}$	...	$tf_{ij}$	...	$tf_{in}$
:	:		:	:	:
Term m	$tf_{m1}$	...	$tf_{mj}$	...	$tf_{mn}$

Term weighting : emphasize term-document relation

$$a_{ij} = L(i, j) \times G(i)$$

weight = local weight X global weight

### frequency weight $L(i, j)$

Term-frequency

$$L(i, j) = tf(i, j)$$

Log transformation

$$L(i, j) = \log_2(tf(i, j) + 1)$$

Binary

$$\begin{aligned} L(i, j) &= 1, & tf(i, j) &\geq 1 \\ L(i, j) &= 0, & tf(i, j) &= 0 \end{aligned}$$

### term weight $G(i)$

Entropy

$$G(i) = 1 + \sum_j \frac{(tf_{ij}/g_i) \log_2 (tf_{ij}/g_i)}{\log_2(N)}$$

GfIdf

Global frequency Inverse document frequency

$$G(i) = \frac{g_i}{d_i}$$

Idf

Inverse document frequency

$$G(i) = \log_2\left(\frac{N}{d_i}\right) + 1$$

Normal

$$G(i) = \sqrt{\frac{1}{\sum_j L(i, j)^2}}$$

## 2. Dimension reduction by singular value decomposition

$$A = U\Lambda V'$$

$A : m \times n$  matrix

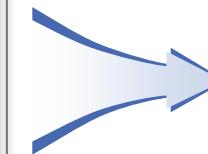
$U : m \times m$  matrix, rank =  $r$ , orthogonal

$V : n \times n$  matrix, orthogonal

$\Lambda : m \times n$  diagonal matrix

$$\lambda_1 \geq \lambda_2 \geq \dots \geq \lambda_r \geq \lambda_{r+1} = \dots = \lambda_{\min(m,n)} = 0$$

$U : m \times m$   
 $\Lambda : m \times n$   
 $V : n \times n$



$U : m \times r$   
 $\Lambda : r \times r$   
 $V : r \times n$

$$\Lambda = \begin{bmatrix} \lambda_1 & 0 & \cdots & 0 & \cdots & 0 \\ 0 & \lambda_2 & 0 & 0 & \cdots & 0 \\ \vdots & 0 & \ddots & \vdots & \cdots & \vdots \\ 0 & 0 & \cdots & \lambda_m & \cdots & 0 \end{bmatrix} \quad \Lambda = \begin{bmatrix} \lambda_1 & 0 & \cdots & 0 \\ 0 & \lambda_2 & 0 & 0 \\ \vdots & 0 & \ddots & \vdots \\ 0 & 0 & \cdots & \lambda_n \\ \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & \cdots & 0 \end{bmatrix}$$

$$\lambda_i (i = 1, 2, \dots, \min(m, n))$$

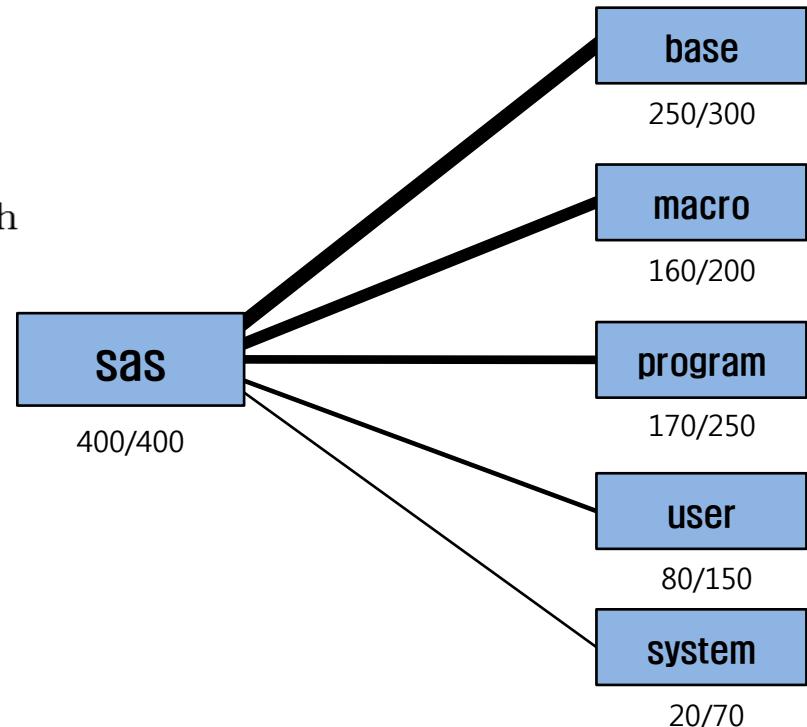
**$D_k = U\Lambda_k V'$**

### 3. Clustering

#### Expectation-Maximization Clustering

- Step 1.
  - Obtain initial parameter estimates
- Step 2.
  - Compute the membership probability of  $x$  in each cluster  $k$ .
  - $P(G_j|x), j = 1, \dots, K$
  - Assign  $x$  to its maximum  $P(G_j|x)$
- Step 3.
  - Update the mixture model parameters for each cluster
- Step 4.
  - Repeat Step 2 and Step 3 until converging

### 4. Network Analysis(Concept Link)



## III. Practical application

## 1. data

1

### Parsing and filtering

2

### Singular value decomposition

3

### Determine the number of clusters

4

### Clustering and cluster profiling

5

### Concept link

상품정보	상품리뷰 (1376) / 구매후기 (10242)	상품 Q&A (371)	반품/교환
<p><b>상품리뷰</b> 구매 후 상품리뷰를 작성하시면 고객님의 구매등급(작성 시점)에 따라 <b>별로최 최대 57점</b> 드립니다. 자세히보기</p> <p>포트&amp;동영상 리뷰</p>			
<p><b>상품리뷰</b></p> <p>[옵션]●사은품선택▶ 세타필 클렌저 237ml[구입 시 안티박테리얼비누증정,●상품필수선택_01대용량_모델:568g] 다른 세티필 세티필 허길래 저도 한번 구매해 봤어요 꿀 테어날을 이용한데 더없이 좋은 상품 이길 비례요.. 저국하고 엄마들이 서로 좋은 그런제품.. 체보고 좋으면 더 구매해야지... [더보기]</p> <p>• 조회수 : 137건 • 댓글수 : 0건 uds*** 2011-05-04 구매</p>			
<p>[옵션]●사은품선택▶ 크랩구매시 세타필후대용기증정,●상품필수선택_03대용량_모이스춰라이징크림568g 제가 이토록는 아니지만 세티필이 건강에도 좋다기에 구매했어요 할은 조금 있구요 묵직한 풀감이 맘에 드네요 양이 굉장히 많기 때문에 함께 보내주신 헤미용기는 훤나 유용하네요 ~ 귀 [더보기]</p> <p>• 조회수 : 435건 • 댓글수 : 0건 jius*** 2011-05-29 구매</p>			
<p>[옵션]●사은품선택▶ 크랩구매시 세타필후대용기증정,●상품필수선택_03대용량_모이스춰라이징크림568g 일단 배송도 잘나들고 포장도 안전하게 잘 되어있어서 망해드립니다. 아.. 후기쓰기가 잘 설레 이네요. 너무 좋아서 그래요. 사진보시면 정말 눈으로 확인할 수 있어요. 일부러 디카페 날짜도 [더보기]</p> <p>• 조회수 : 447건 • 댓글수 : 0건 mics**** 2011-05-27 구매</p>			
<p><b>일반리뷰</b></p> <p><b>상품리뷰</b></p> <p>[옵션]●상품필수선택▶ 세타필후대용기증정,●상품필수선택_01대용량_모이스춰라이징크림568g 저렴하면서 흡도 많고 사용법을 잘해 주어서 계속 구매하게되네요. 건선피부라서 항상 로션을 발라줘야하는데 바른후부터 엄청 좋아졌어요. 정말 좋은것같아요. 얼굴에 바른는 로션도 따 [더보기]</p> <p>• 조회수 : 9건 • 댓글수 : 0건 dyna***** 2011-06-11 구매</p>			
<p>[옵션]●상품선택▶ 세타필 클렌저 237ml[구입 시 안티박테리얼비누증정,●상품필수선택_01대용량_모델:568g] 차름 세보는 제품이에요. 상품평도 괜찮고 주변분들도 써보기잘 할 구매를 해두노. 좋. 팔이 너 무 건강이다. 한참은 제품인거 같아요. 가격도 저렴하고.. 여름은 심하지 않는데 거울엔 너무 [더보기]</p> <p>• 조회수 : 2건 • 댓글수 : 0건 sej***** 2011-06-11 구매</p>			
<p>[옵션]●상품선택▶ 세타필 클렌저 237ml[구입 시 안티박테리얼비누증정,●상품필수선택_01대용량_모델:568g] 예기한데 발라줄려고 신는데 제가발라보니 품질적합이 있는거같기하구.ㅋ.ㅋ 아직에킨텐트 안겠네요.ㅎㅎ 예기한데발라줄려고 신는데 제가발라보니 품질적합이 있는거같기하구.ㅋ.ㅋ</p> <p>• 조회수 : 3건 • 댓글수 : 0건 miyo***** 2011-06-10 구매</p>			

Customers' reviews of internet shopping mall

Best seller 100 products are considered initially

- lotion and cream are selected for analysis

## 2. Structurizing and filtering

Dictionary

**Start List**  
(Selecting Dictionary)

Selecting Dictionary

Stop List  
(Filtering Dictionary)

Filtering Dictionary

**Part of speech  
filtering**

Non-informative		informative		
Part of speech	filter	Part of speech		filter
Aux	auxiliary verb	Filtered	Abbr	abbreviation
Conj	conjunction	Filtered	Adj	<b>adjective</b>
Det	determiner	Filtered	Adv	adverb
Interj	interjection	Filtered	Noun	Noun
Part	particle	Filtered	Num	Number
Prep	<b>preposition</b>	Selected	Prop	<b>proper noun</b>
Pron	pronoun	Filtered	Verb	Verb
		VerbAdj	Verbal adjective	Selected

**synonym dictionary**

Category	Term	Parent
Noun	Price	price
Noun	Value	
Noun	Cost	
Noun	delivery	
Noun	shipping	delivery
Noun	아가	
Noun	아이	
Noun	애기	
Noun	상품	제품
Noun	제조일자	제조일
Noun	제조날짜	
Adj	고맙다	감사하다
Adj	마르다	건조하다
Verb	소개하다	권하다

## 3. SVD

10 Maximum loaded terms					
SVD 1		SVD 2		SVD 3	
Term	Value	Term	Value	Term	Value
용량	0.6529	단점	0.8724	맘놓고	0.7184
바르다	0.6223	유지되다	0.8683	2010년	0.7148
크림	0.5881	질환	0.8669	바디클렌저	0.7024
많다	0.5115	외출	0.8648	벗기다	0.6889
촉촉하다	0.4980	탁월하다	0.8440	감사하다	0.6489
건조하다	0.4907	은은하다	0.7997	2011년	0.6218
발	0.4545	무취	0.7930	걱정없이	0.5790
저렴하다	0.4451	성분	0.7804	포장	0.5734
건조	0.4365	미끈거리다	0.7625	년	0.5440
바르다	0.4179	악하다	0.6962	깨진다	0.5354

### 10 minimum loaded terms

이상하다	0.0976	크림통	-0.1671	상처	-0.1877
거치다	0.0943	피부가 보들보들	-0.1677	고민	-0.1896
요긴하다	0.0907	많다	-0.1684	증상	-0.1911
뽁뽁이	0.0900	간지럽다	-0.1745	심하다	-0.2027
건강	0.0864	촉촉하다	-0.1969	건조하다	-0.2131
남녀노소	0.0795	크림	-0.2028	겨울	-0.2158
교환하다	0.0776	올다	-0.2036	느낌	-0.2337
설명서	0.0670	부담스럽다	-0.2131	긁다	-0.2340
g	0.0548	듬뿍듬뿍	-0.2315	바르다	-0.2386
교환	0.0538	바르다	-0.2620	기嬖다	-0.2445

### SVD 1



- **volume, a lot of, moist, dry, cheap**
- **exchange, g, manual, health, important**



### SVD 2



- **defect , disease, soft, scentless, slippery**
- **apply, generously, moist, tender**



### SVD 3



- **2010, 2011, year, packing, unclothe**
- **scrape, winter, dry, symptom, abrasion**



## 4. clustering

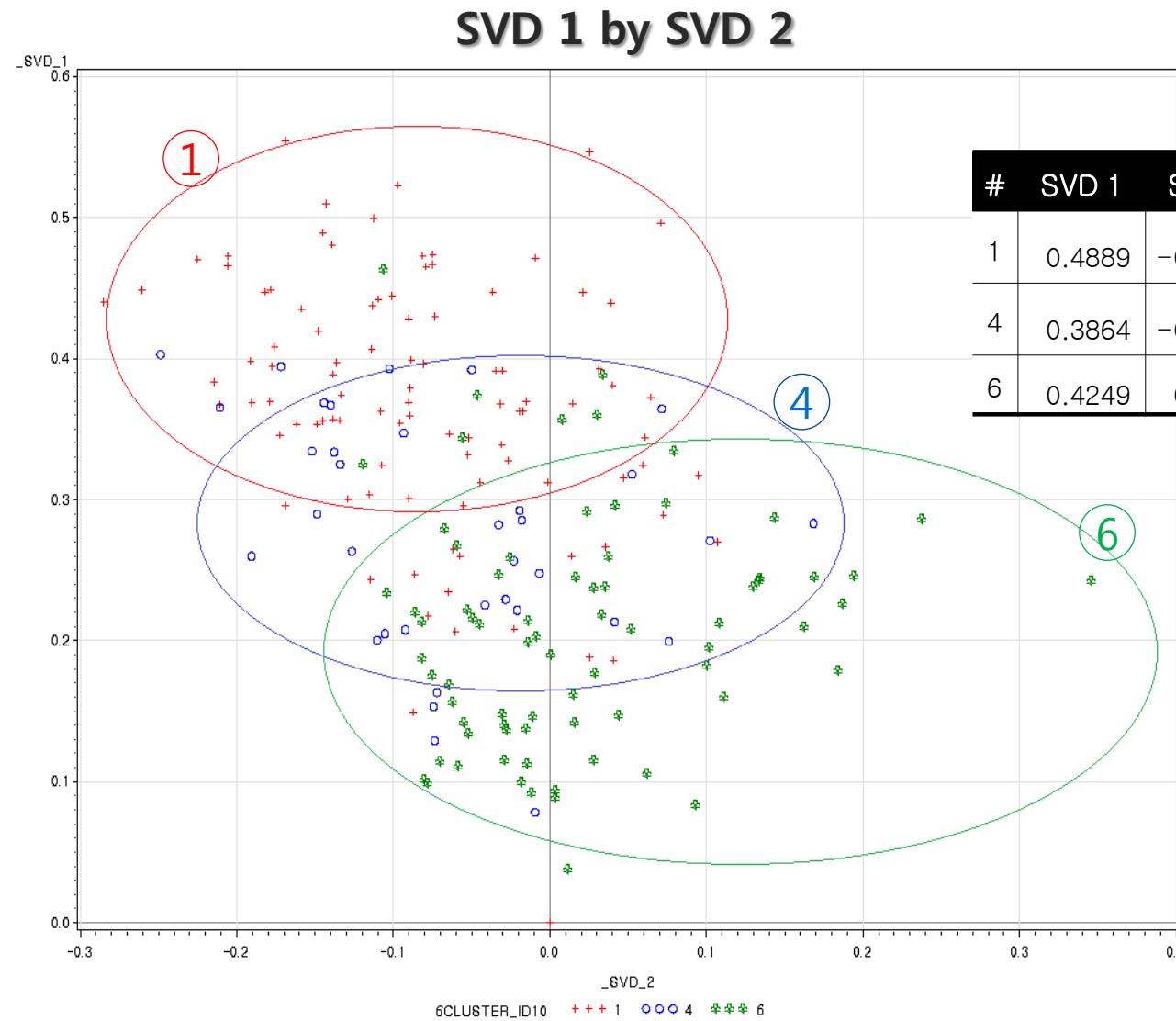
**Criterion for determine  
number of clusters**

K	RMS Std	Pseudo F	Hartigan
2	0.1314	0.423	-244.956
3	0.1369	0.158	-239.009
4	0.1307	2.397	-172.884
5	0.1281	5.652	-181.068
<b>6</b>	<b>0.1132</b>	<b>6.687</b>	<b>-209.477</b>
7	0.1240	3.364	11.238
8	0.1236	2.043	-179.220
9	0.1185	4.598	29.122
10	0.1155	3.066	-176.718

**Cluster profiles**

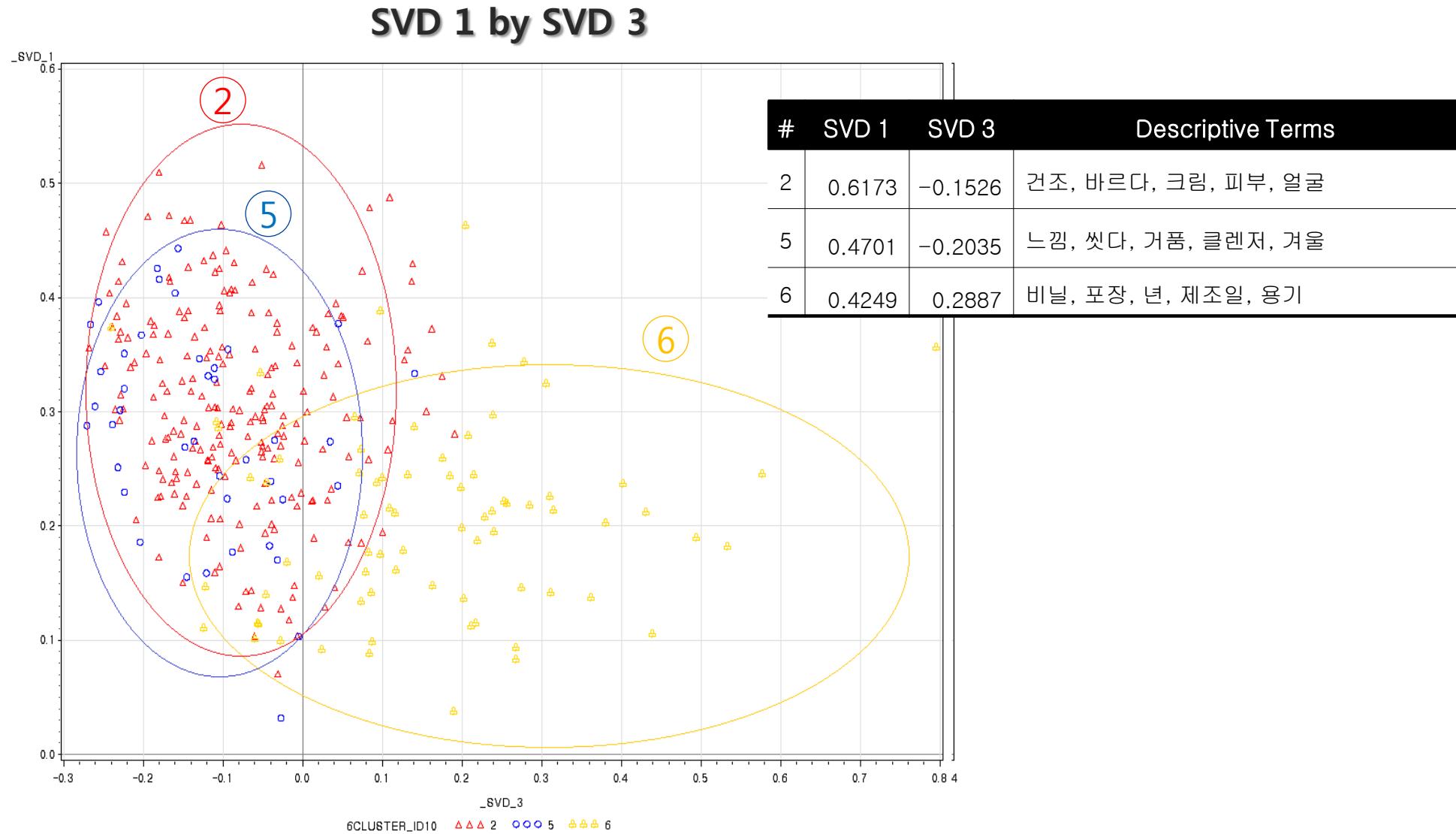
#	SVD 1	SVD 2	SVD 3	Freq	Descriptive Terms
1	0.4889	-0.0992	0.0933	131	용량, 많다, 저렴하다, 틈빠름빠, 보습
2	0.6173	-0.1532	-0.1526	221	건조, 바르다, 크림, 피부, 얼굴
3	0.5065	0.0444	0.0250	99	빠르다, 저렴하다, 추천하다, 배송, 향기
4	0.3864	-0.0849	-0.1207	36	가격, 저렴하다, 배송, 향기, 아토피
5	0.4701	0.0930	-0.2035	42	느낌, 씻다, 거품, 클렌저, 겨울
6	0.4249	0.0237	0.2887	76	비닐, 포장, 년, 제조일, 용기

## 5. Cluster visualization by SVD

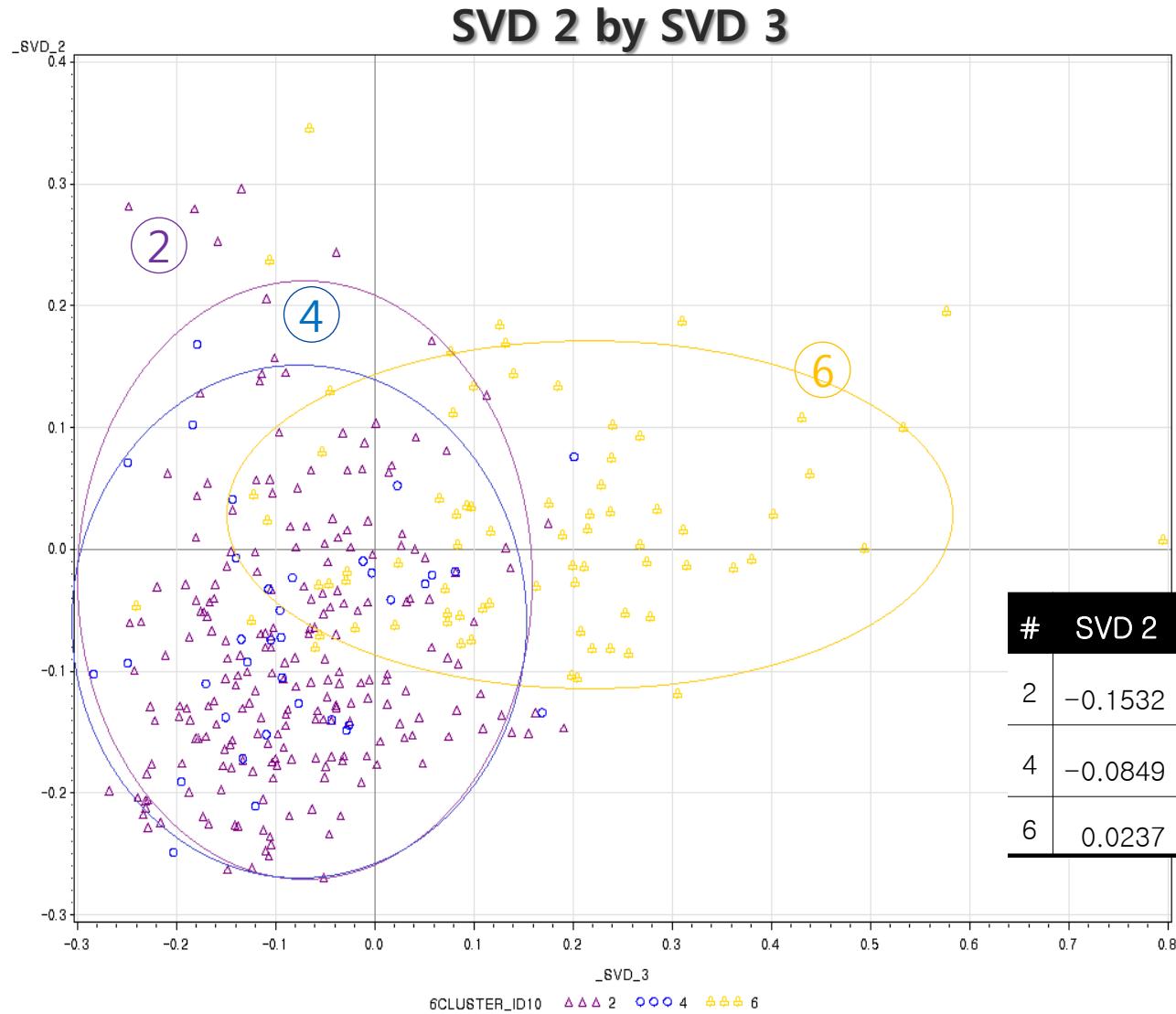


#	SVD 1	SVD 2	Descriptive Terms
1	0.4889	-0.0992	용량, 많다, 저렴하다, 듬뿍듬뿍, 보습
4	0.3864	-0.0849	가격, 저렴하다, 배송, 향기, 아토피
6	0.4249	0.0237	비닐, 포장, 년, 제조일, 용기

## 5. Cluster visualization by SVD

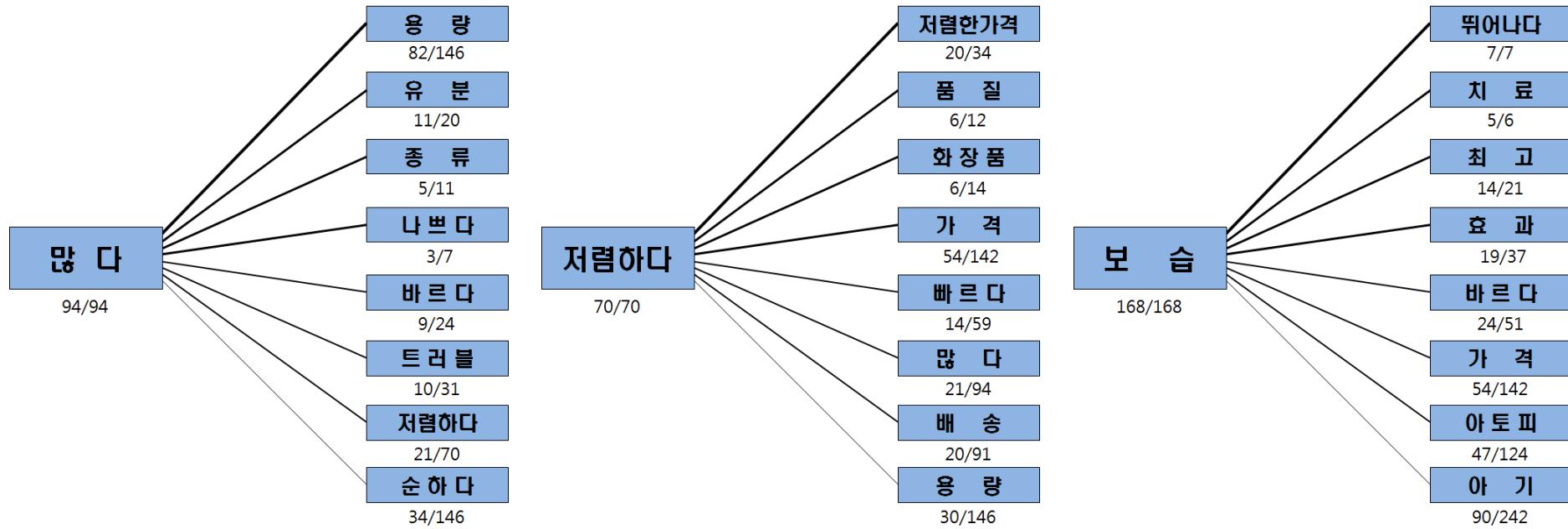


## 5. Cluster visualization by SVD



#	SVD 2	SVD 3	Descriptive Terms
2	-0.1532	-0.1526	건조, 바르다, 크림, 피부, 얼굴
4	-0.0849	-0.1207	가격, 저렴하다, 배송, 향기, 아토피
6	0.0237	0.2887	비닐, 포장, 년, 제조일, 용기

## 6. Concept link



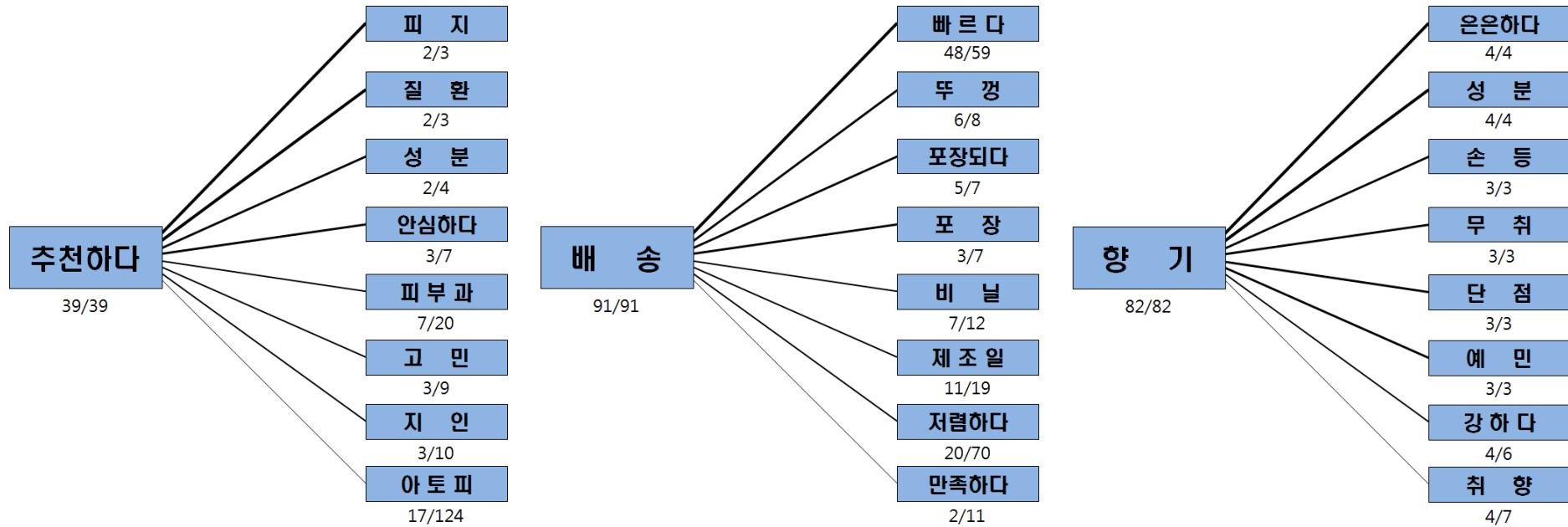
cluster1

Large volume, cheap price, moisturizing, cure

cluster1

Customers' reviews which are about cheap price, excellent moisturizing, and curing an atopic dermatitis.

## 6. Concept link



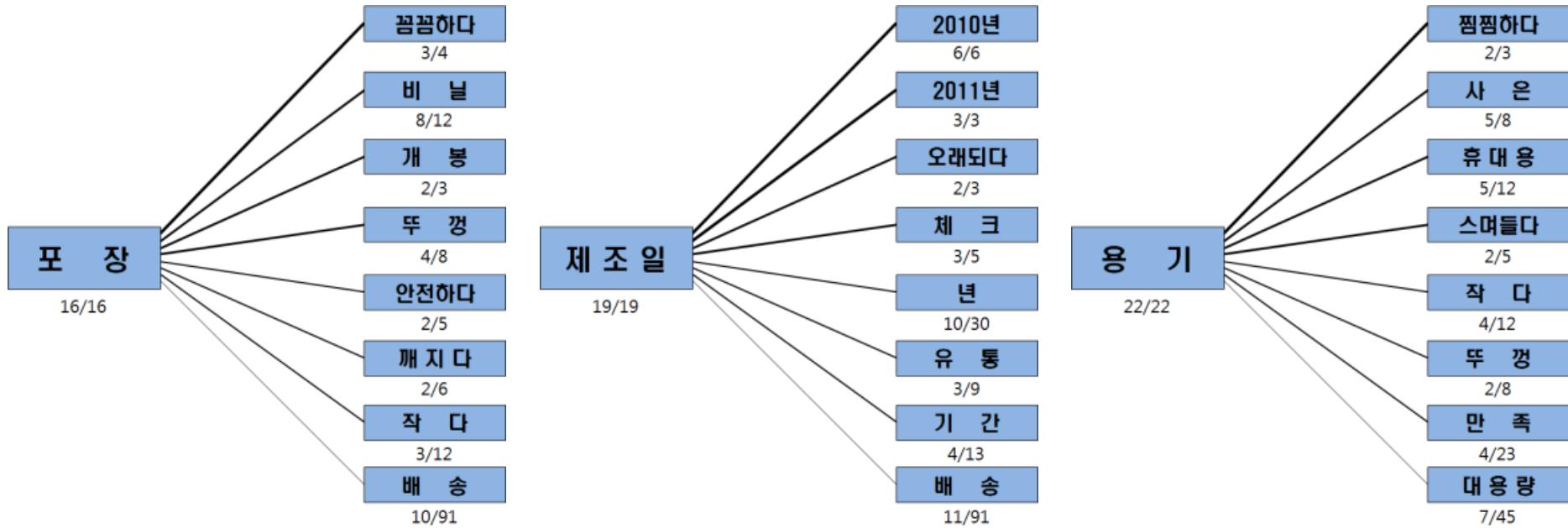
cluster3

Quick delivery, cheap price, recommendation, scent

cluster3

Customers' reviews which are about recommendation for dermatitis, quick delivery, good packing, and scent.

## 6. Concept link



cluster6

Packing, expiration date, negative word

clsuter6

Customers' reviews which are about pacing problem, expired product, etc.  
This cluster consists of negative documents.

## 7. Usage of results

Usage of Text mining results	
<b>Customer management</b>	<ul style="list-style-type: none"><li>▪ Understanding customers' response about the products</li></ul>
<b>marketing</b>	<ul style="list-style-type: none"><li>▪ Target marketing based on customers' preference</li></ul>
<b>promotion</b>	<ul style="list-style-type: none"><li>▪ Finding strength and weakness about the products</li><li>▪ Emphasize strength for selling promotion</li></ul>
<b>Product development</b>	<ul style="list-style-type: none"><li>▪ Realize the weak points of the products</li><li>▪ Re-design the products</li></ul>
<b>Settle a grievance</b>	<ul style="list-style-type: none"><li>▪ Realize customers' complain</li><li>▪ Reduce customers' complain by solving the problems</li></ul>

Thank you

