

# A Look inside the Distributionally Similar Terms

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# “Distributional” Hypothesis

- Extensive use of **distributional similarity** derived from the “**distributional**” hypothesis (Harris 1959) is one of the key concepts of NLP that made it successful.
  - Hindle (1990), Grefenstette (1993), Lee (1997), Lin (1998)
- Reason for its nearly unanimous acceptance is not so much positively motivated, however.
  - If the hypothesis is not accepted, then most of Web-derived data would be intractable.
- Yet ..

# Three Questions We Address

- *Can distributional similarity really be equated with semantic similarity?*
  - No agreement seems to be reached as to what count as semantic similarity.
  - And there are several kinds of semantic similarity itself.
- *Even if distributional similarity can be equated with semantic similarity, to what extent is it so?*
- *Even if they can be equated to a large extent, is it valid on a large scale?*
- We address these questions in our study.

# Outline

- Method
- Preparing data
- Classification task
- Results
- Summary

# Method



# General Framework

- Step 1. Select a set of “base” terms  $B = \{b_1, b_1, \dots, b_n\}$
- Step 2. Use a certain similarity measure  $M$  (such as Jensen-Shannon divergence) to construct a list of  $n$  terms  $T = [t_{i,1}, t_{i,2}, \dots, t_{i,j}, \dots, t_{i,n}]$ 
  - where  $t_{i,j}$  denotes the  $j^{\text{th}}$  most similar term in  $T$  against  $b_i$  in  $B$ .
- Step 3. Generate  $P(k)$ , a set of  $t_{i,1}, t_{i,2}, \dots, t_{i,k}$  with each paired with  $b_i$ . Human raters classify  $P(k)$  with reference to a guideline.

# Product of Steps 1 and 2

base	$b_i$ 's most similar term under $M$	$b_i$ 's 2 <sup>nd</sup> most similar term under $M$	...	$b_i$ 's $k^{\text{th}}$ most similar term under $M$
$b_1$	$t_{1,1}$	$t_{1,2}$	...	$t_{1,k}$
$b_2$	$t_{2,1}$	$t_{2,2}$	...	$t_{2,k}$
$\vdots$	$\vdots$	$\vdots$	$\ddots$	$\vdots$
$b_n$	$t_{n,1}$	$t_{n,2}$	...	$t_{n,k}$

Each row represents  $T[b_i]$

# Parameters Considered

- How much for  $n$ ? In other words, how many “bases” to evaluate?
  - In our case,  $n = 150,000$
- How much for  $k$ ? In other words, how many similar terms to evaluate?
  - In our case,  $k = 2$ .
- What similarity metric to use?
  - We used the Jensen-Shannon divergence for  $M$  under distributional probabilities of  $\langle n, p, v \rangle$  (Kazama et al. 2009)



# Characteristics of Step 3

- We classified 300,000 pairs into the 18 finer-grained classes of semantic relation (to be explained).
- But we also applied candidate filtering (to be explained).
- Note
  - In Kazama's clustering data,  $n$  corresponds to the count rank of dependency relation types. This should be an *indicator* of token frequencies of base terms.

# Sample of Data Used in Step 3

w-reclassified00.xls						
	A	B	C	D	E	F
1	ID	Freq(w1)	w1	w2	type	note
2	000046-2	276782	中国	米国	w[形態素共有のある同類語対]	
3	000060-2	247607	二人	三人	w[形態素共有のある同類語対]	
4	000124-1	169125	友人	知人	w[形態素共有のある同類語対]	
5	000141-1	155062	英語	日本語	w[形態素共有のある同類語対]	
6	000246-1	112967	日本語	英語	w[形態素共有のある同類語対]	
7	000246-2	112967	日本語	フランス語	w[形態素共有のある同類語対]	
8	000278-2	106469	去年	おととし	t[順序づけ可能語対]	
9	000295-2	102504	二つ	三つ	w[形態素共有のある同類語対]	
10	000318-1	97929	他人	隣人	w[形態素共有のある同類語対]	
11	000332-2	95655	患者	被検者	w[形態素共有のある同類語対]	
12	000466-1	76516	業務	職務	w[形態素共有のある同類語対]	
13	000484-2	74686	利用者	購入者	w[形態素共有のある同類語対]	
14	000487-1	74579	一日	毎日	c[(反義性のない)対比語対]	
15	000505-2	73514	工場	加工場	h[上位下位語対]	
16	000531-2	71535	毎日	一日	c[(反義性のない)対比語対]	
17	000532-2	71351	表面	塗装面	h[上位下位語対]	
18	000534-1	71079	人物	登場人物	h[上位下位語対]	
19	000543-2	69966	高齢者	障害者	w[形態素共有のある同類語対]	
20	000565-2	67594	著者	編者	w[形態素共有のある同類語対]	
21	000574-2	66867	近年	数年	w[形態素共有のある同類語対]	
22	000576-2	66637	制度	介護保険制度	h[上位下位語対]	
23	000579-2	66430	今年度	来年度	t[順序づけ可能語対]	
24	000580-1	66417	市内	町内	w[形態素共有のある同類語対]	

# Preparing Data

# 10 Most Similar Terms of “ピアノ” (piano)

rank	Japanese (original)	English translation	Score
1	エレクトーン	<i>Electone</i> , electric organ	-0.322
2	バイオリン	violin	-0.357
3	ヴァイオリン	violin	-0.358
3	チェロ	cello	-0.358
5	トランペット	trumpet	-0.377
6	三味線	<i>shamisen</i> , Japanese 3-string guitar	-0.383
7	サクソ	saxophone	-0.390
8	オルガン	organ	-0.392
9	クラリネット	clarinet	-0.394
10	二胡	erh hu	-0.396



# 10 Most Similar Terms of “チャイコフスキー” (Tchaikovsky)

rank	Japanese (original)	English translation	Score
1	ブラームス	Brahms	-0.152
2	シューマン	Schumann	-0.163
3	メンデルスゾーン	Mendelssohn	-0.166
4	ショスタコーヴィッチ	Shostakovich	-0.178
5	シベリウス	Sibelius	-0.180
6	ハイドン	Haydn	-0.181
6	ヘンデル	Händel	-0.181
8	ラヴェル	Ravel	-0.182
9	シューベルト	Schubert	-0.197
10	ベートーヴェン	Beethoven	-0.190



# Terms Excluded from Candidates

- Strings that were judged to fail to have meaning due to segmentation error.
  - An independent task was performed for this.
- Terms begin with Roman digits (i.e., “0”, “1”, ..., “9”)
- Terms ending with 88 derivational morphemes that lead to either POS-change or obscure semantics
- Terms containing more than one occurrence of “ . ”
  - “ . ” means either disjunction, conjunction or surrogate of “white space” in Japanese.

# 88 Derivational Morphemes for Candidate Filtering

- Hedge-deriver
  - -など, -等, -たち, -達, -ども, -ら, -以外, -ほか, -他, -くらい, -ぐらい, -まま, -ごと, -ついで, -つつ
- Modalizer
  - -とおり, -あたり, -ぶり, -振り, -あまり, -余り, -ほど, -かわり, -代わり
- Nominalizer
  - -たの, -いの, -うの, -くの, -すの, -つの, -ぬの, -ふの, -むの, -ゆの, -るの, -なの, -んか, -るか, -でか, -つか
- Epithet-deriver
  - -さん, -サン, -ちゃん, -チャン, -さま, -サマ, -様, -くん, -君, -どの, -殿
- Temporalizer or Locationalizer
  - -ばあい, 場合, -ため, -為, -せい, -コト, -こと, -事, -トコロ, -ところ, -所, -処, -とき, -時, -ころ, -ごろ, -頃, -際, -なか, -中, -うえ, -上, -下, -前, -後, -ちかく, -近く, -ほう, -方
- Deriver of other POS-terms
  - -的だ, -的に, -した, -った, -である, -では, -です, -ます

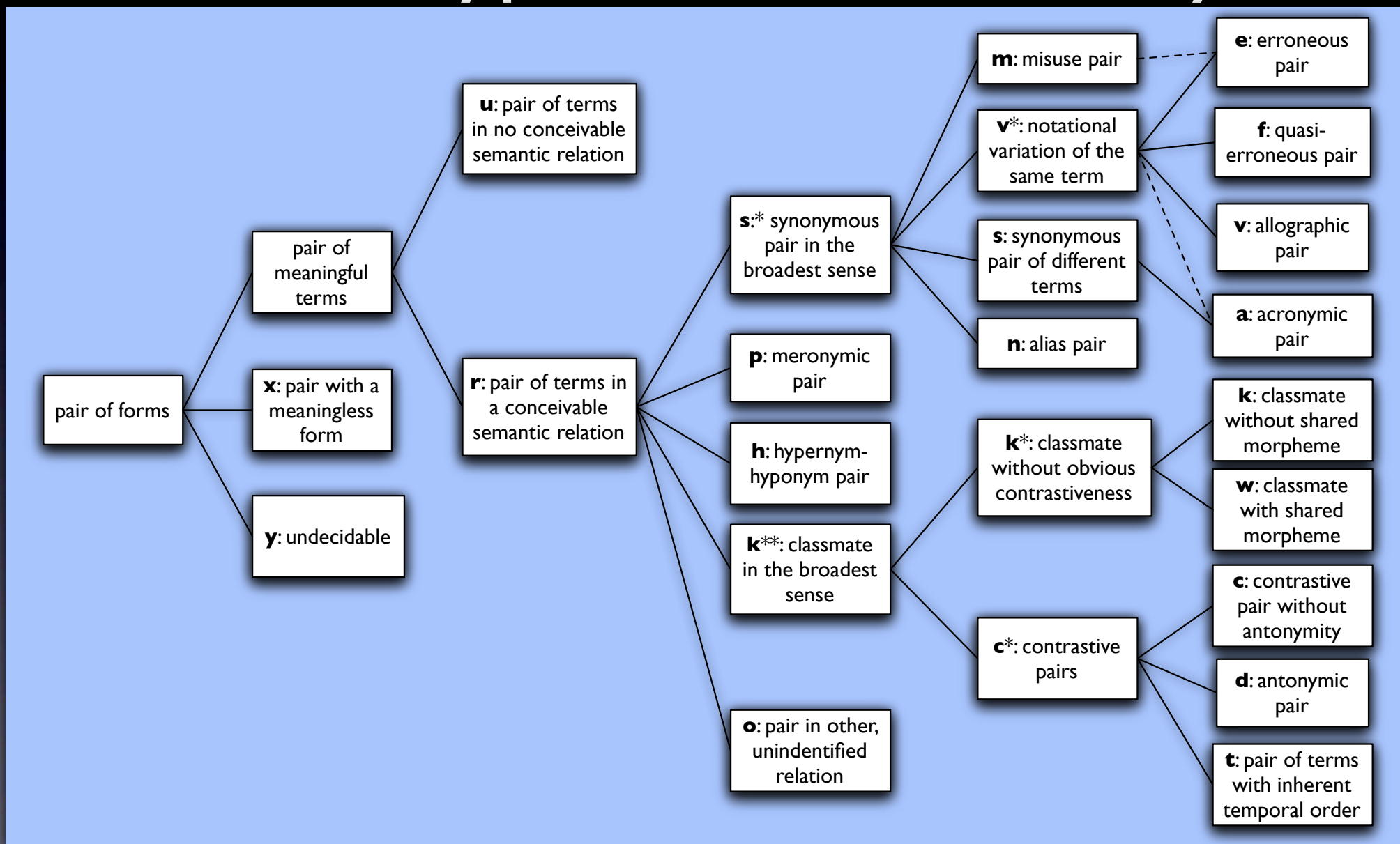
# Classification Task

*Its design and practice*

# Factoring out “semantic similarity”

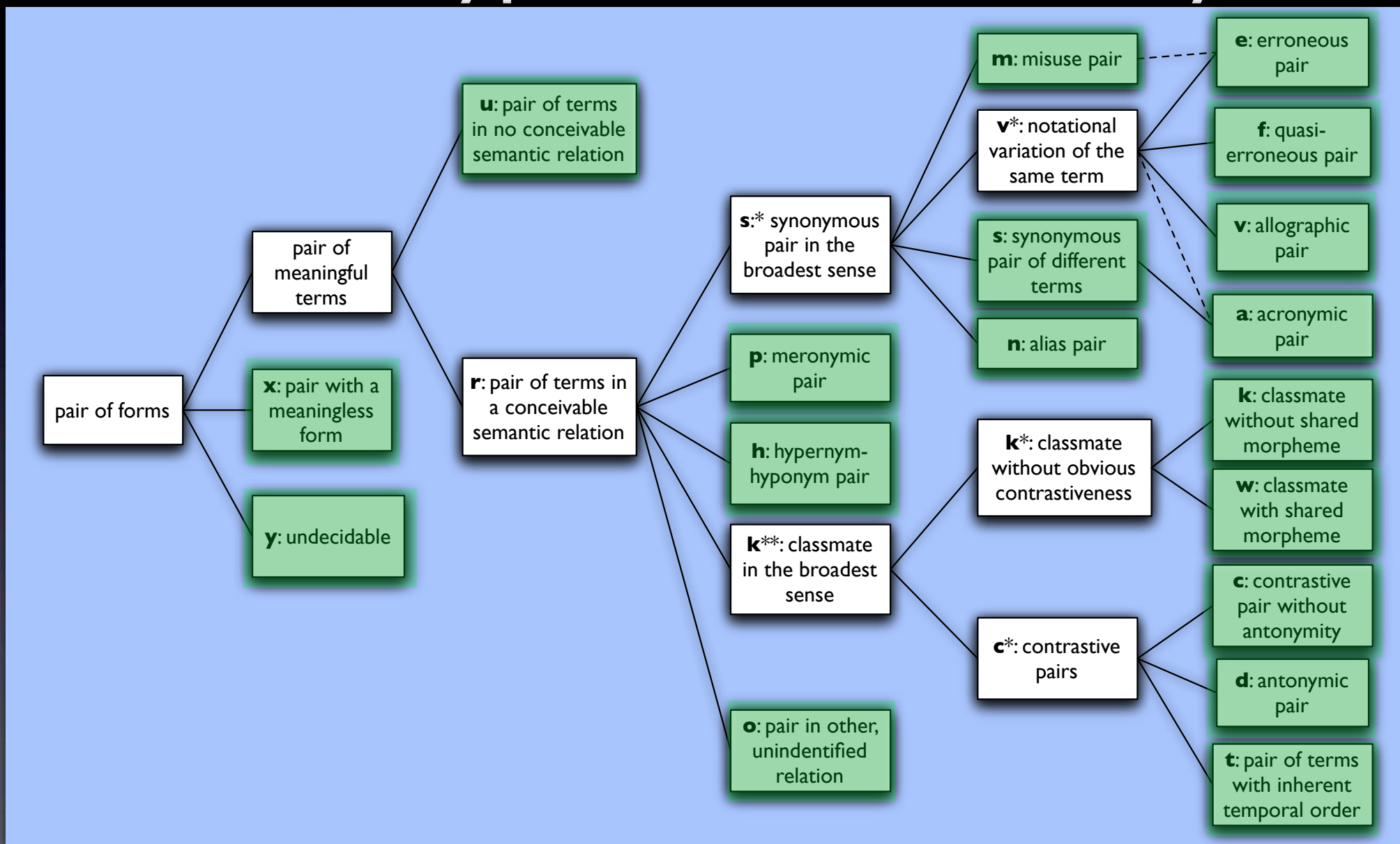
- We employed 18 finer-grained classes build on four basic “components” of semantic similarity
  1. *synonymic* relation
  2. *hypernym-hyponym* relation
  3. *meronymic* relation
  4. *classmate* relation
- They are designed based on research like Fellbaum, ed. (1998), Murphy (2003)

# 18 Subtypes in the Hierarchy





# 18 Subtypes in the Hierarchy



# Characteristics of the Hierarchy

- **s\***, **k\*\***, **p**, **h**, and **o** are major divisions and are expected to be mutually exclusive.
  - **s\*** has four subtypes: **s**, **m**, **v\*** and **n**.
  - **k\*\*** has two subtypes: **k\*** and **c\***.
  - **k\*** has two subtypes: **s\*** and **w** differing with presence of a common morpheme.
  - **c\*** has three subtypes: **c**, **d** and **t**.
- In the most tolerant condition, **{s\*, k\*\*, p, h}** corresponds to the overall class of semantically similar terms.
- Note that **{m, e}** or **{m, e, f}** are only classes in which distributional and semantic similarities do not match up.

# Dealing with Label Ambiguity

- But at least in practice, some labels are not mutually exclusive!
  - This does not guarantee the uniqueness of the labels to be assigned.
- To solve this, the following priority was set to choose the most appropriate one:
  - **e, f < v < a < n < p < h < s < t < d < c < w < k < m < o < u < x < y**
- the leftmost label is the most preferred one.

# Examples

# I. synonymous [s] pairs

1. (根元, 株元) [both mean *root*]
2. (サポート会員, 協力会員) [(*supporting member, cooperating, member*)]
3. (呼び出し元, 親プロセス) [(*invoker of the process, parent process*)]
4. (相手投手, 相手ピッチャー) (*opposing hurler, opposing pitcher*)
5. (病歴, 既往歴) [(*medical history, anamneses*)]



## 2. acronymic [a] pairs

1. (DEC, Digital Equipment)
2. (IBM, International Business Machine)
3. (MS 社, Microsoft 社) [(MS, Inc., Microsoft, Inc.)]
4. (難関大, 難関大学) [both mean *universities hard to enter*]
5. (配置轉換, 配転) [both mean *job displacement*]

## 3. alias [n] pairs

1. (Steve Jobs, founder of Apple, Inc)
2. (Barak Obama, US President)
3. (侑一郎, うにっ子) [(Yuichiro, Unikko)]
  - *Unikko* seems to be the nickname for a cartoon character.
4. (ノグチ, イサム・ノグチ) [(Noguchi, Isamu Noguchi)]

## 4. allographic [v] pairs

1. (Solo, solo) [with or without capitalization]
2. (center, centre), (colour, color) [difference between AmE and BE]
3. (アカスリ, あかすり) [both mean *skin-scrubbing*, pair of katakana notation and hiragana notation]
4. (がん, 癌) [both mean *cancer*, in different character types]
5. (廻り, 回り) [both mean *surrounding of*, in variation]
6. (コンピューター, コンピュータ) [both mean *computer*]

## 5. erroneous [e] pairs

1. (発砲スチロール, 発泡スチロール) [発砲 (shooting) is mistaken for 発泡 (foaming)]
2. (太宰府, 大宰府) [太 and 大 are mistaken]
3. (筋線維, 筋繊維) [線 and 繊 are mistaken]

## 6. quasi-erroneous [f] pairs

1. (スポイト, スポイド) [both mean *dropper*]
2. (ゴルフバッグ, ゴルフバック) [both mean *golf bag*]
3. (ビッグバン, ビックバン) [both mean *Big Bang*]



## 7. misuse [m] pairs

1. (氷漬け, 氷付け) [both mean *frozen*, but the former is not standard form]
2. (開講, 開校) [(*open a lecture*, *open a school*) yet susceptible for misuse]
3. (平行, 並行) [both mean *parallel* with difference in denotation]
4. (恋愛観, 恋愛感) [the latter is an apparently a new terms]

## 8. hypernym-hyponym [h] pairs

1. (検索ツール, 検索ソフト)  
[(*search tool, search software*)]
2. (失業対策, 雇用対策)  
[(*unemployment measures, employment measures*)]
3. (景況, 雇用情勢)  
[(*business conditions, employment conditions*)]
4. (フェスティバル, 音楽祭)  
[(*festival, music festival*)]
5. (シンビジウム, 洋ラン)  
[(*cymbidium, orchid*)]
6. (神秘体験, 臨死体験)  
[(*mystical experience, near-death experience*)]

## 9. meronymic [p] pairs

1. (ちきゅう, うみ) [(*earth, sea*)]
2. (確約, 了解) [(*affirmation, admission*)]
3. (知見, 研究成果) [(*findings, research results*)]
4. (ソーラーサーキット, 外断熱工法) [(*solar circuit system, exterior thermal insulation method*)]
5. (プロバンス, 南フランス) [(*Provence, South France*)]

# 10. classmates with shared morpheme [w]

1. (ガス設備, 電気設備) [(*gas facilities, electric facilities*)]
2. (系列局, 地方局) [(*affiliate station(s), local station(s)*)]
3. (新潟市, 和歌山市) [(*Niigata City, Wakayama City*)]
4. (シナイ半島, マレー半島) [(*Sinai Peninsula, Malay Peninsula*)]

# 11. classmates without shared morpheme [k]

1. (Tom, Jerry)
2. (自分磨き, 体力作り) [(self-culture, training)]
3. (所属機関, 部局) [(sub-organs, services)]
4. (トンパ文字, ヒエログリフ) [(Dongba alphabets, hieroglyphs)]



## 12. contrastive pairs without antonymity [c]

1. (ロマン主義, 自然主義) [(romanticism, naturalism)]
2. (携帯ユーザー, インターネットユーザー) [(mobile user(s), internet user(s))]
3. (海賊版, PS2版) [(bootleg edition, PS2 edition)]

# 13. antonymic [d] pairs

1. (接着, 分解) [(bonding, disintegration)]
2. (砂利道, 舗装路) [(gravel road, pavement)]
3. (西壁, 東壁) [(west wall(s), east wall(s))]
4. (娘夫婦, 息子夫婦)  
[(daughter and son-in-law, son and daughter-in-law)]
5. (外税, 内税) [(tax-exclusive prices, tax-inclusive prices)]
6. (リアブレーキ, フロントブレーキ) [(front break, rear brake)]
7. (タッグマッチ, シングルマッチ) [(tag-team match, single match)]

# 14. pairs with inherent temporal order [t]

1. (稲刈り, 田植え)  
[(*harvesting of rice, planting of rice*)]
2. (ご出発日, ご到着日) [(*day of departure, day of arrival*)]
3. (進路決定, 進路選択)  
[(*career decision, career selection*)]
4. (居眠り, 夜更かし)  
[(*catnap, stay up*)]
5. (密猟, 密輸) [(*poaching, contraband trade*)]
6. (投降, 出兵) [(*surrender, dispatch*)]
7. (二回生, 三回生) [(*2<sup>nd</sup>-year student(s), 3<sup>rd</sup>-year student(s)*)]

# 15. pairs in other relation [o]

1. (下心, 独占欲) [(*ulterior motives, possessive feeling*)]
2. (理論的背景, 基本的概念) [(*theoretical background, basic concepts*)]
3. (アレクサンドリア, シラクサ) [(*Alexandria, Syracuse*)]

# 16. unrelated [u] pairs

1. (非接触, 高分解能) [*(noncontact, high resolution)*]
2. (模倣, 拡大解釈) [*(imitation, overinterpretation)*]



# 17. nonsensical [x] pairs

1. (わったん, まる赤)

2. (セルディ, 瀬璃)

3. (チル, エルダ)

4. (ウーナ, 香螢)

5. (ma, ジョージア)

# 18. unclassified [y] pairs

1. (場所網, 無規準ゲーム)
2. (fj, スラド)
3. (反力, 断力)

# Results

# Details of the Classification Task

- 17 people were asked to perform the classification task using the guidelines specified by the first and second author.
  - The task took nearly 3 months (= regular 2 months + extra 1 month for rework).
- The quality of the product turned out to be very low in some cases.
  - Rework on **o**- and **w**-cases was requested.

Rank	Count	Ratio (%)	Cumulative (%)	Class	Label
1	108,149	36.04	36.04	classmates without common	<b>k</b>
2	67,089	22.35	58.39	classmates with common	<b>w</b>
3	26,113	8.70	67.09	synonymic pairs	<b>s</b>
4	24,599	8.20	75.29	hypernym-hyponym pairs	<b>h</b>
5	20,766	6.92	82.21	allographic pairs	<b>v</b>
6	18,950	6.31	88.52	pairs in “other” relation	<b>o</b>
7	12,383	4.13	92.65	unrelated pairs	<b>u</b>
8	8,092	2.70	95.34	contrastive pairs	<b>c</b>
9	3,793	1.26	96.61	pairs with temporal order	<b>t</b>
10	3,038	1.01	97.62	antonymic pairs	<b>d</b>
11	2,995	1.00	98.62	meronymic pairs	<b>p</b>
12	1,855	0.62	99.23	acronymic pairs	<b>a</b>
13	725	0.24	99.48	alias pairs	<b>n</b>
14	715	0.24	99.71	erroneous pairs	<b>e</b>
15	397	0.13	99.85	misuse pairs	<b>m</b>
16	250	0.08	99.93	nonsensical pairs	<b>x</b>
17	180	0.06	99.99	quasi-erroneous pairs	<b>f</b>
18	33	0.01	100.00	unclassified	<b>y</b>



# Basic Results

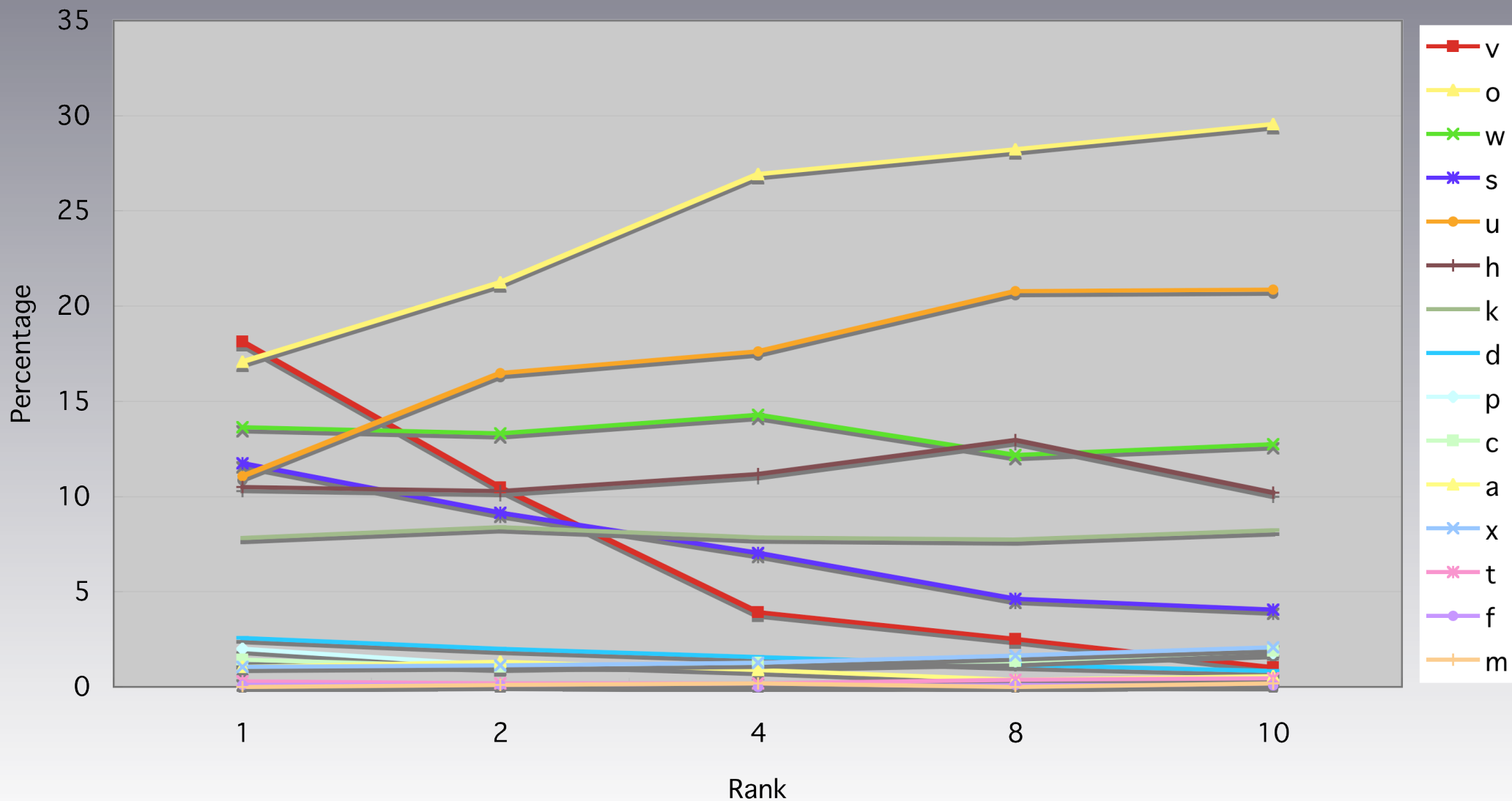
1. Union of **k** and **w** makes 58.39% (strict condition).
2. Union of **k\*\*** and **s\*** makes 79.01% (moderate condition).
  - **k\*\*** = {**k, w, c, d, t**} is a generalized class of classmates to make 62.10%.
  - **s\*** = {**s, a, n, v, e, f, m**} generalized class of synonymic pairs to make 16.91%
3. All classes except **o, u, m, x** and **y** make roughly 88% (loose condition).
  - The second or third conditions can be understood as confirmations of the “distributional” hypothesis.

# Further Question

- What is the (side)effect of  $k = 2$ ? Did we get a representative result?
- An informal preliminary analysis of sample 1000 pairs (generated based on bases at ranks 2, 4, 8, 10) indicates
  - the rate of  $s^*$  (especially  $v$ ) decreases at lower ranks.
  - the rates of  $o$  and  $u$  increase at lower ranks.

# Rankwise Distribution of Types

Rankwise Distribution of Classes



# Summary

- Our aim was to see to what extent distributionally similar terms can be equated with semantically similar terms when semantic similarity is factored out.
- Loose condition with all labels except **o**, **u**, **m**, **x** and **y** make roughly **88%**. Even moderate condition with **k\*\*** and **s\*** makes **79.01%**. So, it would be safe to say that the “distributional” hypothesis is confirmed.
- Though our case is limited in that  $n=150,000$  and  $k=2$ , rankwise distribution of class suggests that our results are with fair representativeness.

Thank you  
for Your Attention



# Appendix

# Potential inconsistency

- The distinction among classes is sometimes obscure, especially the one between **p** and **h** is hard to make in Japanese.
  - For example, is the right label for (火星, 天体) **p** or **h**?
- This ambiguity is influenced by the ambiguity of 天体: If *heavenly body* is meant, then **h** is right. If *heavenly bodies* is meant, then **p** is right.