Finding Deceptive Opinion Spam by Any Stretch of the Imagination

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- Consumers increasingly rate, review and research products online
- Potential for opinion spam
 - Disruptive opinion spam
 - Deceptive opinion spam



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★★★★★ Great Customer Service!!, April 7, 2011

By <u>akaempf</u> ✓ - <u>See all my reviews</u>

Amazon Verified Purchase (What's this?)

This review is from: Apple iPad 2 MC984LL/A Tablet (64GB, Wifi + AT&T 3G, White) NEWEST MODEL (Personal Computers)

"WE SHIP TECH" is a great reliable company. I ordered the iPad2 late 3/30 @ 10:50pm and received the iPad2 4/1. When I wrote an email to them on the 3/31 they responded in about 20 min max. It's so hard to find great customer service and not get scammed these days that "We Ship Tech" is a breath of fresh air!! I would surely use them again and highly recommend them to anyone who expects great products & service. Thank you We Ship Tech!!!!!

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★★★★ Works Just as expected, May 14, 2007

By <u>Laurie B. Cook</u>

- See all my reviews

REAL NAME

This review is from: Belkin F5U301 CableFree 4-Port USB 2.0 Hub with Dongle (Electronics)

Supplies good range and does provide true wireless USB. Software worked right out of the box. I have been recommending this nifty little device to all my friends. Very useful device.

Which of these two hotel reviews is deceptive opinion spam?

Date of review: Jun 9, 2006

4 people found this review helpful

I have stayed at many hotels traveling for both business and pleasure and I can honestly stay that The James is tops. The service at the hotel is first class. The rooms are modern and very comfortable. The location is perfect within walking distance to all of the great sights and restaurants. Highly recommend to both business travellers and couples.

Date of review: Jun 9, 2006

4 people found this review helpful

My husband and I stayed at the James Chicago Hotel for our anniversary. This place is fantastic! We knew as soon as we arrived we made the right choice! The rooms are BEAUTIFUL and the staff very attentive and wonderful!! The area of the hotel is great, since I love to shop I couldn't ask for more!! We will definatly be back to Chicago and we will for sure be back to the James Chicago.

Which of these two hotel reviews is deceptive opinion spam?

Answer:

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Overview

- Motivation
- Gathering Data
- Human Performance
- Classifier Performance
- Conclusion

- Label existing reviews
 - -Can't manually do this
 - -Duplicate detection (Jindal and Liu, 2008)
- Create new reviews
 - -Mechanical Turk

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James Chicago

Hotel class ★★★★

55 East Ontario, Corner of Rush and Ontario, Chicago, IL 60611

877.526.3755 Hotel website E-mail hotel

What travelers say about James Chicago

Great location (33) Magnificent mile (14)

Room service (20) Very good (13)

Michigan avenue (13)

Comfortable bed (10)

Friendly and helpful (8)

Home → United States → Illinois (IL) → Chicago → Chicago Hotels → James Chicago

Reviews you can trust

Very nice (18)

Trader joe (16)

Boutique hotel (15)



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- Average time spent:
 - > 8 minutes
- Average length:
 - > 115 words

- 400 truthful reviews
 - -TripAdvisor.com
 - Lengths distributed similarly to deceptive reviews

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- Why bother?
 - -Validates deceptive opinions
 - -Baseline to compare other approaches

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- 80 truthful and 80 deceptive reviews
- 3 undergraduate judges
 - Truth bias
- 2 meta-judges

			TRUTHFUL			DECEPTIVE		
		Accuracy	P	R	\mathbf{F}	P	R	F
HUMAN	JUDGE 1	61.9%	57.9	87.5	69.7	74.4	36.3	48.7
	JUDGE 2	56.9%	53.9	95.0	68.8	78.9	18.8	30.3
	JUDGE 3	53.1%	52.3	70.0	59.9	54.7	36.3	43.6

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Performed at chance (p-value = 0.1)

HUMAN

JUDGE

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3	53.1%	52.3	70.0	59.9	54.7	36.3	43.6	

• 80 truthful a Performed at chance (p-value = 0.5) | ive reviews

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Classified fewer than 12% of opinions as deceptive!

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META	MAJORITY	58.1%	54.8	92.5	68.8	76.0	23.8	36.2	
	SKEPTIC	60.6%	60.8	60.0	60.4	60.5	61.3	60.9	

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- 80 truthful and 80 dece
- 3 undergraduate judges No more truth bias!

- -Truth bias
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Overview

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- Three feature sets
 - -Genre identification
 - -Psycholinguistic deception detection
 - Text categorization
- Linear SVM

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Approach	Features	Accuracy	P	\mathbf{R}	\mathbf{F}	P	\mathbf{R}	\mathbf{F}
GENRE IDENTIFICATION	POS	73.0%	75.3	68.5	71.7	71.1	77.5	74.2

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Outperforms human judges! (p-values = {0.06, 0.01, 0.001})

TRUTHFUL/INFORMATIVE			DECEPTIVE/IMAGINATIVE			
Category	Variant	Weight	Category	Variant	Weight	
	Singular	0.008		Base	-0.057	
NOUNS	Plural	0.002		Past tense	0.041	
NOUNS	Proper, singular	-0.041		Present participle	-0.089	
	Proper, plural	0.091	VERBS	Singular, present	-0.031	
	General	0.002		Third person	0.026	
ADJECTIVES	Comparative	0.058 singular, pre		singular, present	0.020	
	Superlative	-0.164		Modal	-0.063	
PREPOSITIONS	General	0.064	ADVERBS	General	0.001	
DETERMINERS	General	0.009	ADVERDS	Comparative	-0.035	
COORD. CONJ.	General	0.094	PRONOUNS	Personal	-0.098	
VERBS	Past participle	0.053	FRONOUNS	Possessive	-0.303	
ADVERBS	Superlative	-0.094	PRE-DETERMINERS	General	0.017	

- Rayson et. al. (2001)
 - Informative on left, imaginative on right

TRUTH	TRUTHFUL/INFORMATIVE			DECEPTIVE/IMAGINATIVE			
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	Proper, plural	0.091	VERBS	Singular, present	-0.031		
	General	0.002		Third person	0.026		
ADJECTIVES	Comparative	0.058	e.g., best, finest singular, present		0.020		
	Superlative *	-0.164		Modal	-0.063		
PREPOSITIONS	General	0.064	ADVERBS	General	0.001		
DETERMINERS	General	0.009	ADVERDS	Comparative	-0.035		
COORD. CONJ.	General	0.094	PRONOUNS	Personal	-0.098		
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• Rayson et. al. (2001) e.g., most

- Informative on left, imaginative on right

- Linguistic Inquire and Word Count (Pennebaker et al., 2007)
 - -Counts instances of ~4,500 keywords
 - Regular expressions, actually
 - Keywords are divided into 80 dimensions across 4 broad groups

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- Linguistic processes
 - -e.g., average number of words per sentence
- Psychological processes
 - e.g., talk, happy, know, feeling, eat
- Personal concerns
 - e.g., job, cook, family
- Spoken categories
 - e.g., yes, umm, blah

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GENRE IDENTIFICATION	POS	73.0%	75.3	68.5	71.7	71.1	77.5	74.2
PSYCHOLINGUISTIC	LIWC	76.8%	77.9	76.0	76.6	76.4	77.5	76.9
DECEPTION DETECTION	LIWC	10.670	11.2	70.0	10.0	10.4	11.5	10.9

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Outperforms PoS! (p-value = 0.02)

- Text categorization (n-grams)
 - -Unigrams
 - -Bigrams⁺
 - Includes unigrams
 - -Trigrams⁺
 - Includes unigrams and bigrams

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PSYCHOLINGUISTIC	LIWC	76.8%	77.2	76.0	76.6	76.4	77.5	76.9
DECEPTION DETECTION	LIWC	10.070	11.2	10.0	10.0	10.4	11.9	10.5
	UNIGRAMS	88.4%	89.9	86.5	88.2	87.0	90.3	88.6
TEXT CATEGORIZATION	BIGRAMS	89.6%	90.1	89.0	89.6	89.1	90.3	89.7
TEXT CATEGORIZATION	LIWC+BIGRAMS	89.8%	89.8	89.8	89.8	89.8	89.8	89.8
	TRIGRAMS	89.0%	89.0	89.0	89.0	89.0	89.0	89.0

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TEXT CATEGORIZATION	BIGRAMS	89.6%	90.1	89.0	89.6	89.1	90.3	89.7
TEXT CATEGORIZATION	LIWC+BIGRAMS	89.8%	89.8	89.8	89.8	89.8	89.8	89.8
	TRIGRAMS	89.0%	89.0	89.0	89.0	89.0	89.0	89.0

Outperforms all other methods!

LIWC+B	BIGRAMS
TRUTHFUL	DECEPTIVE
-	chicago
	my
on	hotel
location	, and
)	luxury
$\operatorname{allpunct}_{\operatorname{LIWC}}$	experience
floor	hilton
(business
the_hotel	vacation
bathroom	i
small	spa
helpful	looking
\$	while
hotel	husband
other	$my_husband$

- Spatial difficulties (Vrij et al., 2009)
- Psychological distancing (Newman et al., 2003)

LIWC+B	IGRAMS
TRUTHFUL	DECEPTIVE
-	chicago
	my
★ on	hotel
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)	luxury
$allpunct_{LIWC}$	experience
★ floor	hilton
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the_hotel	vacation
★ bathroom	i
\star small	spa
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other	\star my_husband

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 - http://www.cs.cornell.edu/~myleott/op_spam
- Evaluated human deception detection performance
- Developed automated classifiers capable of nearly 90% accuracy
 - Relationship between deceptive and imaginative text
 - Importance of moving beyond universal deception cues

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Thank you. Questions?

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