

Master Degree in Data Science and Economics

## Text Mining and Sentiment Analysis



# Sentiment Analysis and Opinion Mining

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sed noli modo



*“The only thing worse than being talked about,  
is not being talked about.”*

*— Oscar Wilde.*

**Sentiment analysis and opinion mining** is the field of study that analyzes people's **opinions, sentiments, evaluations, attitudes, and emotions** from **written language**.

It is one of the most active research areas in **natural language processing** and is also widely studied in **data mining, Web mining, and text mining**.

# Levels of Analysis

## Document level

To classify whether a whole opinion document expresses a positive or negative sentiment.

This level of analysis assumes that each document expresses opinions on a single entity.

## Sentence level

To determine whether each sentence expressed a positive, negative, or neutral opinion.

This level of analysis is closely related to *subjectivity* classification in that it tries to distinguish between *objective* and *subjective* sentences.

## Entity level

To directly detect opinions. An opinion consists of a sentiment (positive or negative) and a target (of opinion).

For example, the sentence “*the iPhone’s call quality is good, but its battery life is short.*” evaluates two aspects: call quality (positive) and battery life (negative), of iPhone (entity).

# Levels of Analysis (example)

	DOCUMENT	SENTENCE	ENTITY
<i>Rented in Phoenix.</i>		N	N → <i>Phoenix</i>
<i>Last minute rental so getting into a car with a cigarette burn on the drivers seat told me this was not a non smoking car.</i>	+	-	- → <i>car</i>
<i>The Good counter lady was nice, polite and efficient, so that, thanks to Phoenix I could put captain Bars on my sons Army uniform.</i>		+	+ → <i>counter lady</i> + → <i>Phoenix</i>

# Sentiment Lexicon

One of the most straightforward ways of addressing sentiment analysis is to define a **lexicon of sentiment words** or **opinion words** (or even small key phrases) that may be used as indicators for sentiment.

- Baccianella, S., Esuli, A., & Sebastiani, F. (2010, May). Sentiwordnet 3.0: an enhanced lexical resource for sentiment analysis and opinion mining. In Lrec (Vol. 10, No. 2010, pp. 2200-2204). <https://github.com/aesuli/sentiwordnet>
- Musto, C., Semeraro, G., & Polignano, M. (2014, December). A Comparison of Lexicon-based Approaches for Sentiment Analysis of Microblog Posts. In DART@ AI\* IA (pp. 59-68).
- De Albornoz, J. C., Plaza, L., & Gervás, P. (2012, May). SentiSense: An easily scalable concept-based affective lexicon for sentiment analysis. In LREC (Vol. 12, pp. 3562-3567).



# Issues with Sentiment Lexicon (I)

Word orientation depends on the **application context** and **word sense disambiguation**



*this camera **sucks!***

*this vacuum cleaner really **sucks***



# Issues with Sentiment Lexicon (II)

*A sentence containing sentiment words may not express any sentiment*

Interrogative  
sentences pretty  
often do not  
express  
sentiment

*can you tell me which Sony  
camera is **good** ?*

... but  
sometimes they  
do

*Does anyone know how to  
repair this **terrible** printer ?*

Also conditional  
sentences pretty  
often do not  
express  
sentiment

*if i can find a **good** camera  
in the shop, i will buy it*

... but  
sometimes they  
do

*if you are looking for a  
**good** car, get toyota  
camry.*



# Issues with Sentiment Lexicon (III)

*“... no dark sarcasm in the classroom ...”*



*What a **great** car! it stopped working in two days.*

# Issues with Sentiment Lexicon (IV)

*Many (objective) sentences without sentiment words can also imply opinions*



*this fridge uses a lot  
of energy*



*after sleeping on the  
mattress for two  
days, a valley has  
formed in the  
middle*

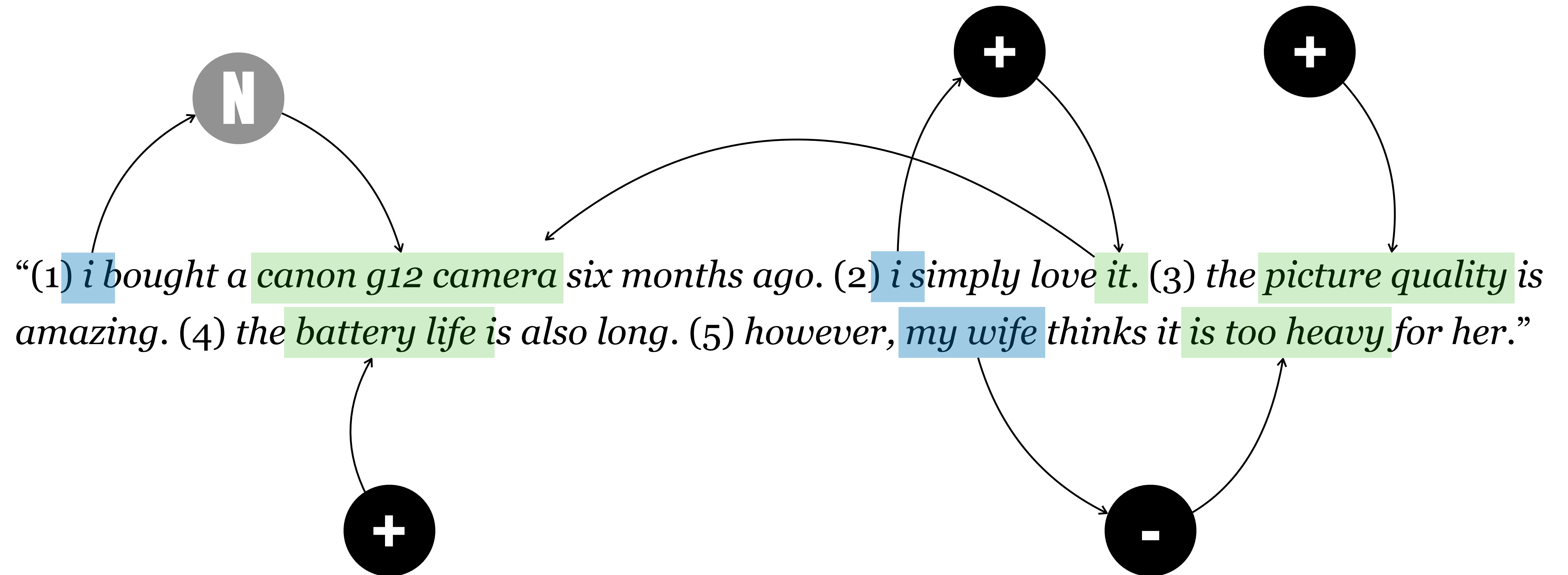


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# **Problems and Definitions in Sentiment Analysis**



# Opinion definition (I)

**Definition (Opinion):** An opinion is a quadruple,  $(g, s, h, t)$ , where ***g*** is the opinion (or sentiment) **target**, ***s*** is the sentiment about the target (either polarity or score or even an emotion), ***h*** is the **opinion holder**, and ***t*** is the **time** when the opinion was expressed.

# Entity definition (I)

**Definition (entity):** An entity  $e$  is a product, service, topic, issue, person, organization, or event. It is described with a pair,  $e: (t, W)$ , where  $t$  is a **hierarchy of parts**, sub-parts, and so on, and  $W$  is a set of **attributes** of  $e$ . Each part or sub-part also has its own set of attributes.



# Entity example

A particular model of camera is an entity, e.g., Canon G12. It has a set of attributes, e.g., picture quality, size, and weight, and a set of parts, e.g., lens, viewfinder, and battery. Battery also has its own set of attributes, e.g., battery life and battery weight.

A topic can be an entity too, e.g., tax increase, with its parts “tax increase for the poor,” “tax increase for the middle class,” and “tax increase for the rich.”

## Canon G12 10 MP Digital Camera with 5x Optical Image Stabilized Zoom and 2.8 Inch Vari-Angle LCD

by [Canon](#)



482 ratings

| 31 answered questions

Available from these sellers.

Style: **US Version**

- 10.0-megapixel sensor and the DIGIC 4 Image Processor combine to create Canon's HS SYSTEM for improved low light performance
- Shoot 720p HD video in stereo sound; HDMI output
- Canon's Hybrid IS compensates for angular and shift camera shake during close-up shooting
- 5x optical zoom with Optical Image Stabilizer; 28mm wide-angle lens; optical viewfinder
- Capture images and video to SD/SDHC/SDXC memory card, MultiMediaCard, MMC Plus Card, HC MMC Plus Card (not included)

Used (19) from \$120.00

# Opinion definition (II)

**Definition (opinion):** An opinion is a quintuple,  $(e, a, s_{ea}, h, t)$ , where  $e$  is an **entity**,  $a$  is an aspect of  $e$ ,  $s_{ea}$  is the **sentiment** on aspect  $a$  of entity  $e$ ,  $h$  is the **opinion holder**, and  $t$  is the **time** when the opinion is expressed by  $h$ . The sentiment  $s_{ea}$  is positive, negative, or neutral, or expressed with different strength /intensity levels, e.g., 1–5 stars as used by most review sites on the Web.

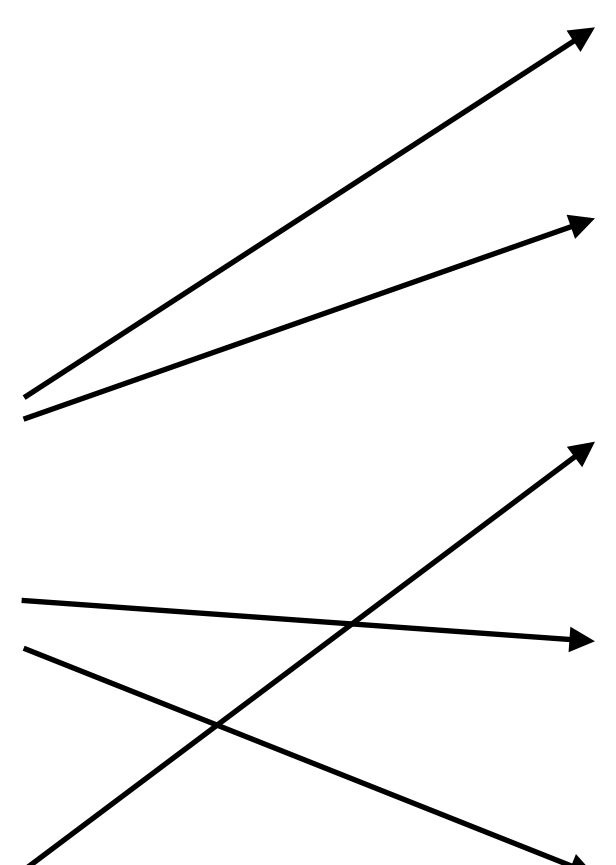
**Objective of sentiment analysis:** Given an opinion document  $d$ , discover all opinion quintuples  $(e, a, s_{ea}, h, t)$  in  $d$ .



**Example:** Posted by: **big John** Date: **Sept. 15, 2011**

(1) *i bought a **Samsung camera** and **my friends** brought **a canon camera** yesterday. (2) in the past week, we both used the cameras a lot. (3) the **photos** from **my Samy** are **not that great**, and the **battery life** is short too. (4) **my friend** was **very happy** with his **camera** and **loves its picture quality**. (5) **i** want a **camera** that can take good **photos**. (6) **i** am going to return **it** tomorrow.*

<b>1</b>		<b>Samsung camera</b>	<b>Photo</b>	<b>-</b>	<b>Big John</b>	<b>15.9.2011</b>
<b>2</b>		<b>Samsung camera</b>	<b>Battery life</b>	<b>-</b>	<b>Big John</b>	<b>15.9.2011</b>
<b>3</b>		<b>Samsung camera</b>	<b>General</b>	<b>-</b>	<b>Big John</b>	<b>15.9.2011</b>
<b>4</b>		<b>Canon camera</b>	<b>General</b>	<b>+</b>	<b>Big John friend</b>	<b>15.9.2011</b>
<b>5</b>		<b>Canon camera</b>	<b>Photo</b>	<b>+</b>	<b>Big John friend</b>	<b>15.9.2011</b>



Named Entity  
Recognition  
(Ner)

Aspect  
Extraction

Opinion Holder  
Extraction

Time  
Extraction

Aspect  
Sentiment  
Classification

# Types of opinion

<b>Opinion</b>	<b>Regular</b>	<b>Comparative</b>
<b>Explicit</b>	<i>“coke tastes great,”</i>	<i>“coke tastes better than Pepsi.”</i>
<b>Implicit</b>	<i>“i bought the mattress a week ago, and a valley has formed,”</i>	<i>“the battery life of nokia phones is longer than Samsung phones.”</i>



# Subjectivity

**Definition (sentence subjectivity):** An *objective sentence* presents some *factual information* about the world, while a *subjective sentence* expresses some *personal feelings, views, or beliefs*.

- **A subjective sentence may not express any sentiment.** For example, “*I think that he went home*” is a subjective sentence, but does not express any sentiment, as well as “*I want a camera that can take good photos*”.
- **Objective sentences can imply opinions or sentiments due to desirable and undesirable facts.** For example, the following two sentences which state some facts clearly imply negative sentiments (which are implicit opinions) about their respective products because the facts are undesirable:

“*the earphone broke in two days.*”

“*i brought the mattress a week ago and a valley has formed* ”



# Emotions

**Definition (emotion):** Emotions are our subjective feelings and thoughts.

- **Rational evaluation:** Such evaluations are from rational reasoning, tangible beliefs, and utilitarian attitudes. For example, the following sentences express rational evaluations:

*“the voice of this phone is clear,”*

*“this car is worth the price,”* and

*“i am happy with this car.”*

- **Emotional evaluation:** Such evaluations are from non-tangible and emotional responses to entities which go deep into people’s state of mind. For example, the following sentences express emotional evaluations:

*“i love iPhone,”*

*“i am so angry with their service people”* and

*“this is the best car ever built.”*

Sometimes, sentiment ratings are referred to a combination of polarity and emotion:

Sentiment	Rational	Emotional
Positive	1	-2
Neutral	0	0
Negative	-1	2



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# Some Datasets & Resources

The “Tweet Emotion Intensity Dataset” data set taken from “8th Workshop on Computational Approaches to Subjectivity, Sentiment and Social Media Analysis (WASSA- 2017)”. <http://saifmohammad.com/WebPages/EmotionIntensity-SharedTask.html>

10000	How the fu*k! Who the heck! moved my fridge!... should I knock the landlord door. #angry #mad ##	anger	.938
10001	So my Indian Uber driver just called someone the N word. If I wasn't in a moving vehicle I'd have jumped out #disgusted	anger	.896
10002	@DPD_UK I asked for my parcel to be delivered to a pick up store not my address #fuming #poorcustomerservice	anger	.896
10003	so ef whichever butt wipe pulled the fire alarm in davis bc I was sound asleep #pissed #angry #upset #tired #sad #tired #hangry #####	anger	.896
10004	Don't join @BTCare they put the phone down on you, talk over you and are rude. Taking money out of my acc willynilly! #fuming	anger	.896

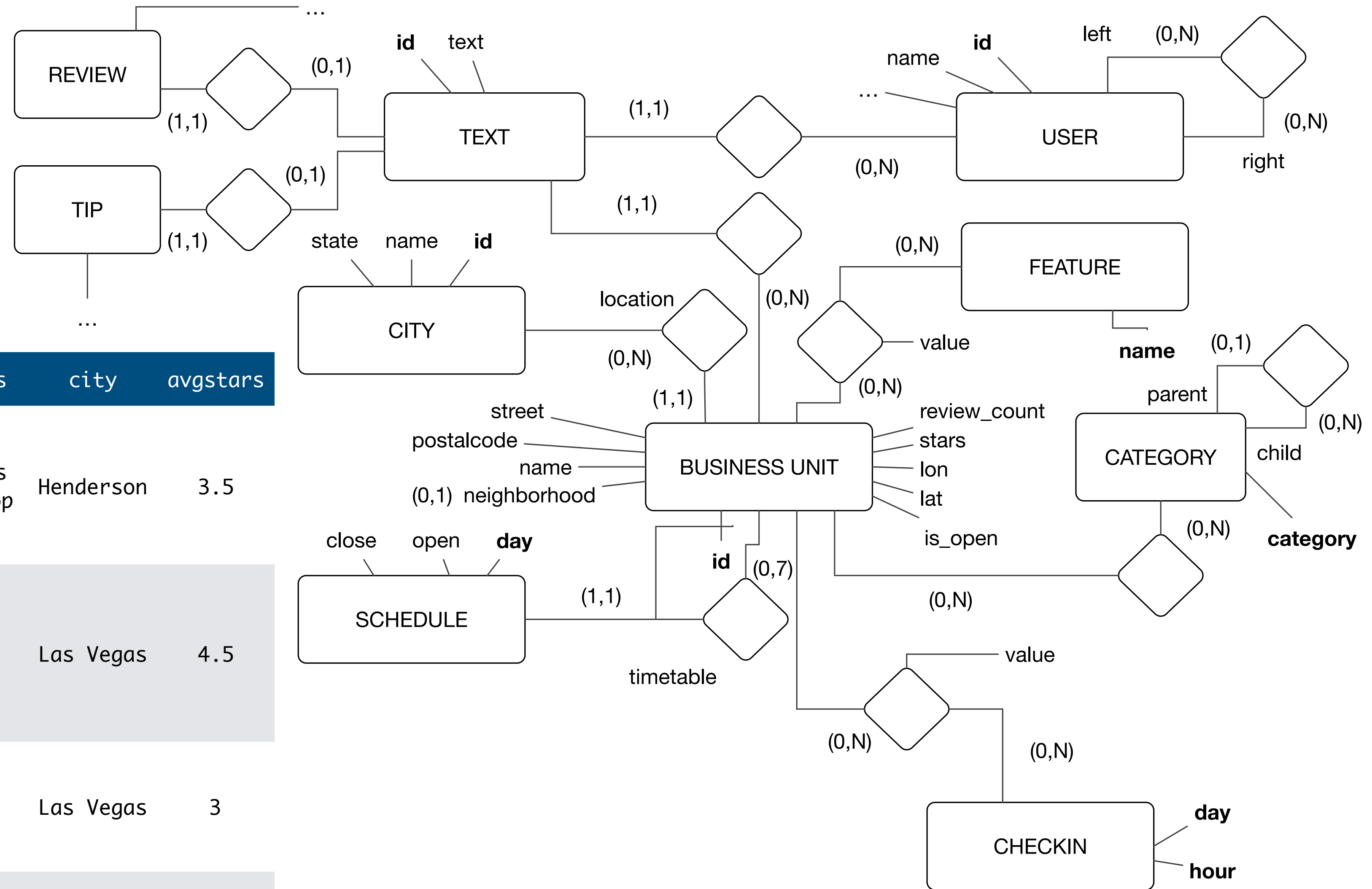
EmoBank: 10k sentences manually annotated with Valence, Arousal, and Dominance. Link: <https://github.com/JULIELab/EmoBank>

The emotional dimensions of **Valence**, **Arousal** and **Dominance** (VAD) are able to derive a **person's interest** (attraction), **level of activation** and perceived **level of control** for a particular situation from textual communication, such as emails.

Id	split	V	A	D	text
110CYL068_1036_1079	train	3.00	3.00	3.20	Remember what she said in my last letter? "
110CYL068_1079_1110	test	2.80	3.10	2.80	If I wasn't working here.
110CYL068_1127_1130	train	3.00	3.00	3.00	.."
110CYL068_1137_1188	train	3.44	3.00	3.22	Goodwill helps people get off of public assist...
110CYL068_1189_1328	train	3.55	3.27	3.46	Sherry learned through our Future Works class ...

# Yelp Dataset

(<https://www.yelp.com/dataset/challenge>)



content	stars	business	city	avgstars
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24 hour tacos? Yes please.

This place is a modest spot in the middle of a strip mall, seats about 2 dozen, and has a little self serve ...

3

Robertos  
Taco Shop

Henderson

3.5

2Nd time eating here today.1st time was great but now i dont think so. My husband ordered chicken fried steak. The gravey tasted very strange. ...

1

Egg & I

Las Vegas

4.5

A word of caution! Don't use their ATM if you have a credit union. My Fiancé and I visited last Saturday. We expected to grab a to-go burger, just to say we ...

3

Heart  
Attack  
Grill

Las Vegas

3

Absolutely amazing!! We just ordered take out and it was delicious (+was very well packed)...cannot imagine what its like to dine in :)

5

Khao San  
Road

Toronto

4



# Multimodal Datasets

## CMU-MOSI Dataset

The Multimodal Corpus of Sentiment Intensity (CMU-MOSI) dataset is a collection of 2199 opinion video clips. Each opinion video is annotated with sentiment in the range  $[-3,3]$ . The dataset is rigorously annotated with labels for subjectivity, sentiment intensity, per-frame and per-opinion annotated visual features, and per-milliseconds annotated audio features.

**<http://multicomp.cs.cmu.edu/resources/cmu-mosi-dataset>**

## IEMOCAP Database

The Interactive Emotional Dyadic Motion Capture (IEMOCAP) database is an acted, multimodal and multispeaker database, recently collected at SAIL lab at USC. It contains approximately 12 hours of audiovisual data, including video, speech, motion capture of face, text transcriptions. It consists of dyadic sessions where actors perform improvisations or scripted scenarios, specifically selected to elicit emotional expressions.

**<https://sail.usc.edu/iemocap>**



Other datasets	Dataset	Author	Year	Description	Format	Size	Emotion categories
	affectivetext	Strapparava & Mihalcea	2007	Classification of emotions in news headlines	SGML/txt	250 headlines	anger, disgust, fear, joy, sadness, surprise, V
	crowdflower_data	CrowdFlower	2016	Annotated dataset of tweets via crowdsourcing.	csv	40k tweets	anger, enthusiasm, fun, happiness, hate, neutral, sadness, surprise, worry, love, boredom, worry, relief, empty
	dailydialog	Li Yanrand et al.	2017	Manually labelled conversations dataset with topics and emotions	text	13k dialogs	anger, disgust, fear, joy, sadness, surprise
	emotion-cause	Diman Ghazi&Diana Inkpen&Stan Szpakowicz	2015	Automatically built dataset annotated with emotion and the stimulus using FrameNet's emotionsdirected frame	XML	820 sents + 1594 sents	anger, sad, happy, surprise, fear, disgust
	EmoBank	Sven Buechel	2017	Large-scale corpus annotated with emotion according to VAD scheme	text	10k	VAD
	fb-valence-arousal-anon	Preotiuc Pietro	2016	VA on Facebook	csv	2.8k posts	VA
	grounded_emotions	Liu, V.&Banea, C.&Mihalcea	2017	They look into wheter the effect of weather, news events, relates to the tweet sentiment	text	2.5k tweets	joy, sadness
	isear	Klaus R. Scherer and Harald Wallbott	1990	reported situations in which emotions were experienced	text (mdb/sav)	3000 docs	joy, fear, anger, sadness, disgust,shame, guilt
	tales-emotions	Cecilia Ovesdotter Alm	2005	Dataset of manually annotated tales used in a document classification task	text	15k sents	angry, disgusted, fearful, happy, sad, surprised, mood (positive, negative)