

BIG DATA AND CUSTOMER EMOTION DYNAMICS: AUTOMATED ANALYSES IN CHAT SERVICES



Yom-Tov, Ashtar, Altman, Natapov, Barkay, Westphal, &
Rafaeli. *Companion of the The Web Conference 2018.*

CUSTOMER EMOTION

- Emotions prevail in service interactions (Grandey, 2004; Groth & Grandey, 2012)

Presumption.

High frequency of emotions, and specifically of negative emotions



**„Your service is awful. I’m an angry customer.
Solve my problems and make me feel better!“**

CUSTOMER EMOTION DYNAMICS

- Emotions are dynamic (Hareli & Rafaeli, 2008; van Kleef, 2014; Weiss & Cropanzano, 1996)

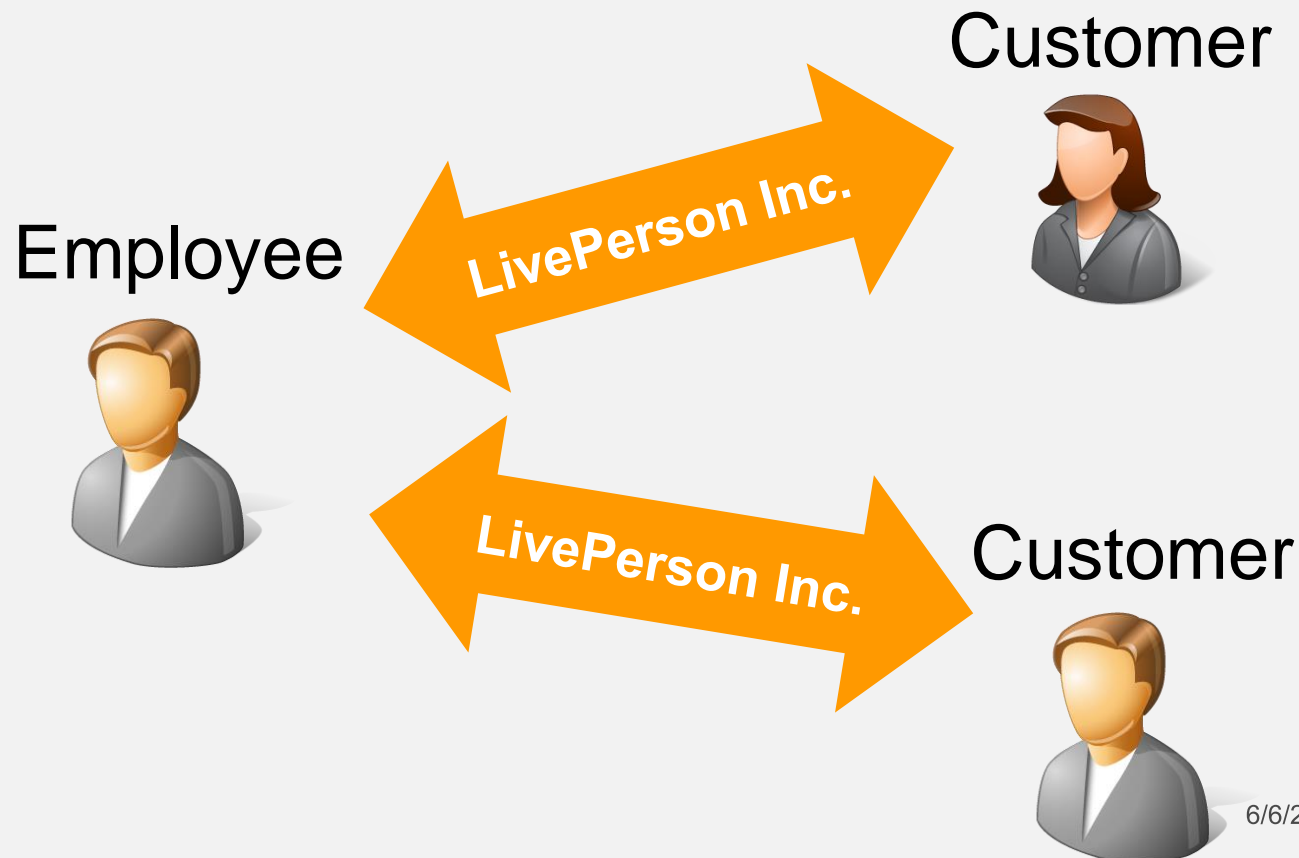
...but...

- Emotion measurements are static (Donaldson & Grant-Vallone, 2002)



Technology-Mediated Service
Creates Opportunities!

THE FRAMEWORK: CHAT-BASED SERVICE



Jane Smith



12:34 PM

Hi, I'm Jane. H

Wen wi

2 min ago

I understand that you would like to access your My Account. We are working to fix the problem

2 min ago

That os not good enough. It has been 9 days

1 min ago

I know how important this is for you. Let me go ahead and access your account so I can assist you with your query. Would that be Ok?

1 min ago

No - I don't trust you

1 min ago

Sd off goodbye

NATURE OF CHAT-BASED SERVICE

Multiple levels of analysis:

- I. Atomic (message) level
- II. Cumulative level (interaction/subset)

RESEARCH GOALS

Need to design new model for detection of customer emotion in chat interactions







1. Tool development and validation

Need to understand customer emotion dynamics and their connection to service outcomes



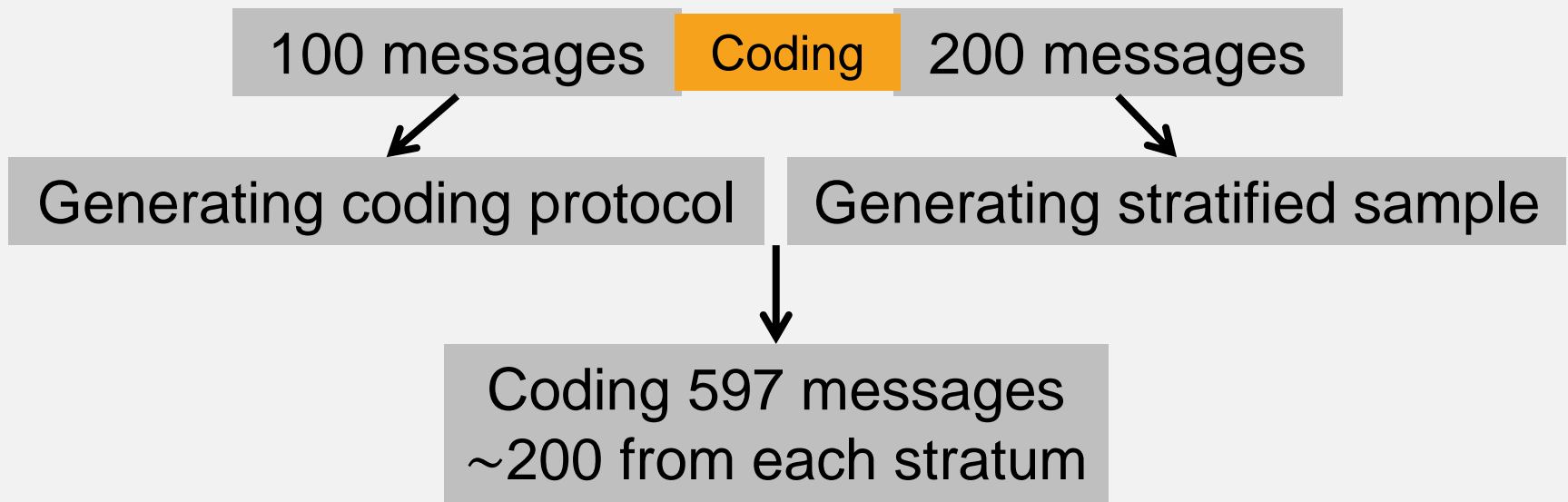
2. Tool application

PART I.I: TOOL DEVELOPMENT

- Lexicon-based approach
- Chat data, provided by  LIVEPERSON
-    score for each customer message
- Tool adjustments:
 - (1) Service, (2) Brand, (3) Customer language

PART I.2: TOOL VALIDATION

Human coding (Larsson, 1993; Amabile et al., 2004)



PART I.2: TOOL VALIDATION



Negative emotion class

<i>Model</i>	<i>Precision</i>	<i>Recall</i>	<i>F₁</i>	<i>F_{0.5}</i>
CustSent	0.719	0.236	0.355	0.51
Stanford	0.335	0.509	0.404	0.36
LIWC	0.479	0.115	0.186	0.294
SentiStrength	0.494	0.216	0.3	0.393



Positive emotion class

<i>Model</i>	<i>Precision</i>	<i>Recall</i>	<i>F₁</i>	<i>F_{0.5}</i>
CustSent	0.866	0.569	0.687	0.784
Stanford	0.546	0.339	0.418	0.486
LIWC	0.491	0.717	0.583	0.524
SentiStrength	0.813	0.677	0.739	0.781

PART II: TOOL APPLICATION

Sample:

- 390,438 Telecom interactions, 8,526,814 messages (October to December 2016)

Variables:

- Customer 😞 😐 😊
- Service quality measures:
- Operational Data

First Call Resolution (FCR):
“Was your service need
resolved in this interaction?”

CUSTOMER EMOTION DYNAMICS & SERVICE QUALITY

- Cognitive Appraisal Theory & Justice Theories (Folkman et al. 1986; Folger & Cropanzano, 1998)

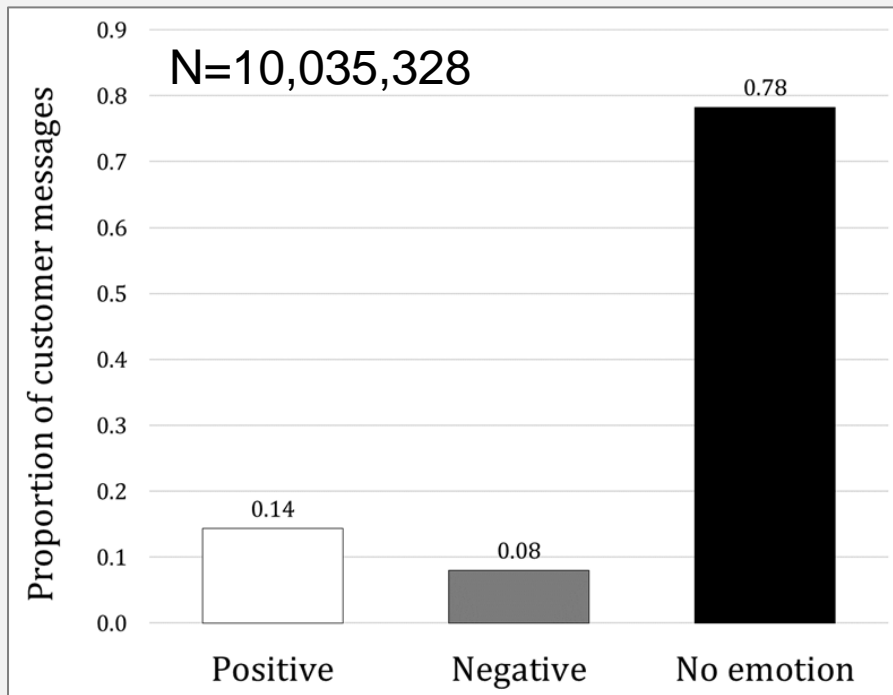
Hypothesis 1. *Customer emotions during interactions are dynamic, evolving from initial negative into more positive emotions*

- Service failure and recovery logic (e.g. DeWitt et al. 2008; Casidy & Shin, 2015)

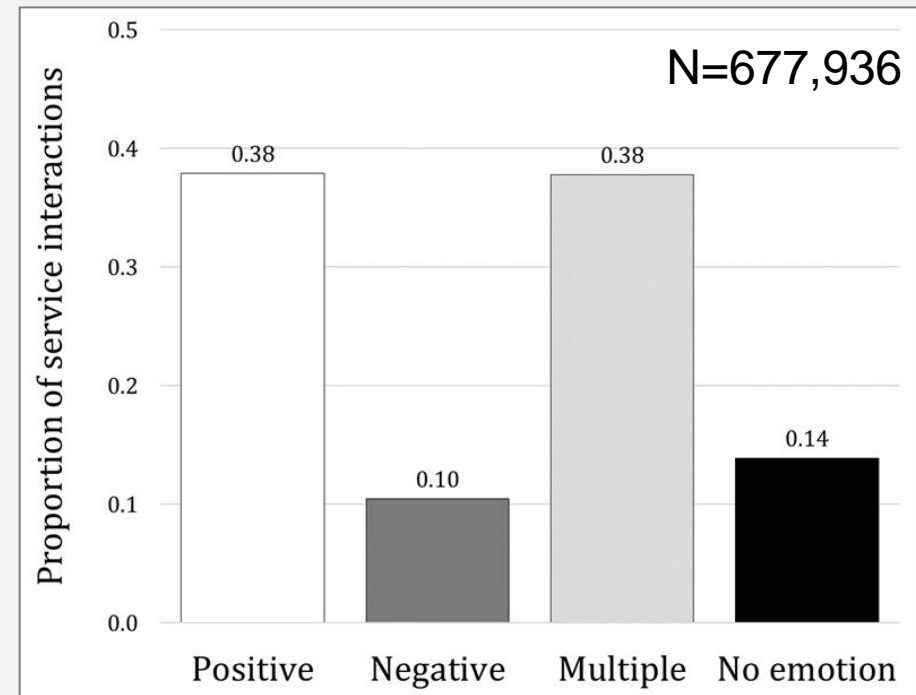
Hypothesis 2. *The magnitude of change in emotion from negative to positive reflects service quality*

Findings

CUSTOMER EMOTION TYPE AND FREQUENCY



Frequency of emotion
in customer messages

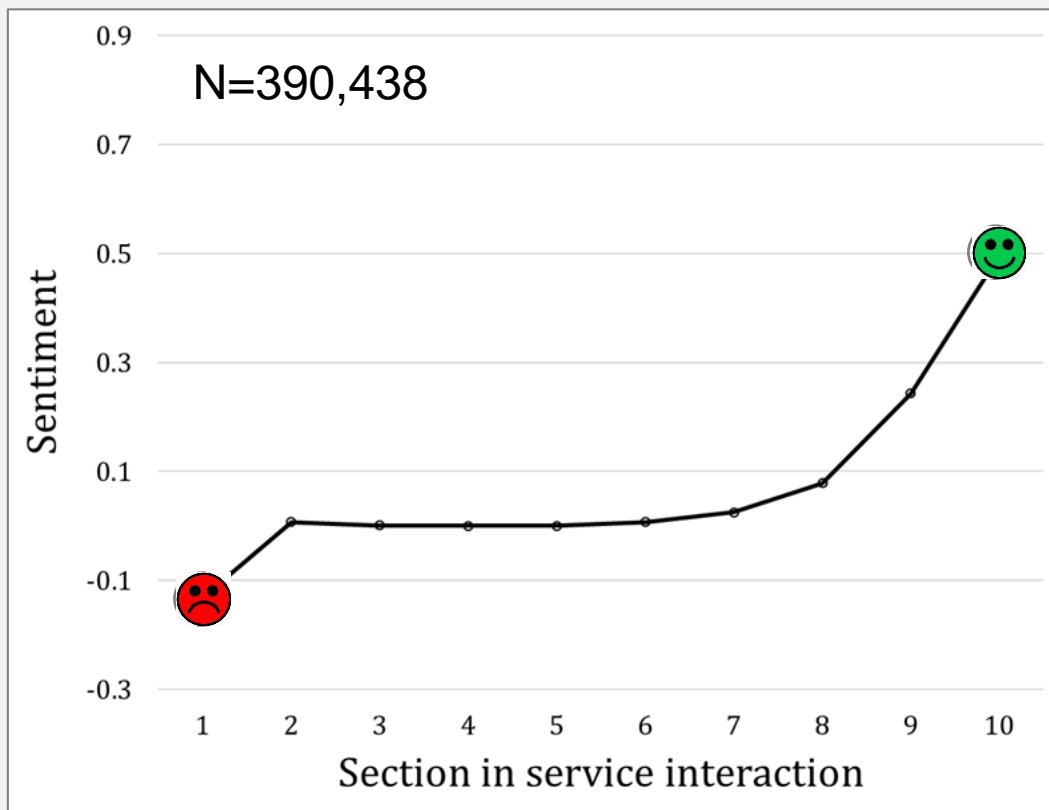


Frequency of emotion
in full service interactions

Findings


CUSTOMER EMOTION DYNAMICS

Hypothesis 1



Paired sample t-test:

$M_{\text{difference}}=0.63$,
 $t(390437)=450.52$,
 $p<0.001$

 Single sample t-test:
 $t(390437)=-138.72$,
 $p<0.001$

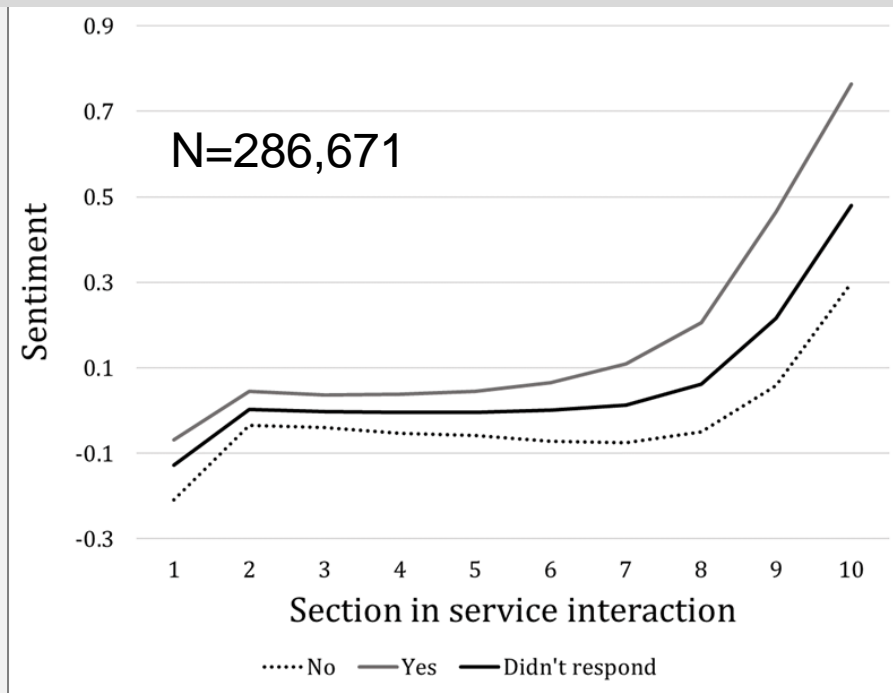
 $t(390437)=458.5$,
 $p<0.001$

Findings

CUSTOMER EMOTION DYNAMICS & SERVICE QUALITY

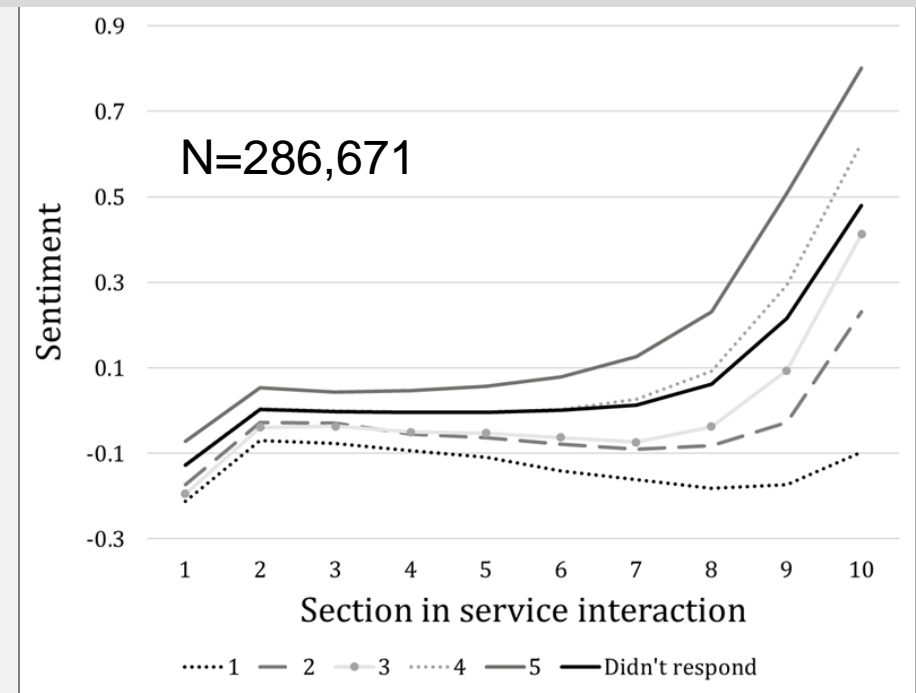
Hypothesis 2

$\chi^2(10)=28481.386, p<0.001$



Sentiment in different sections
by Service Recovery response

$\chi^2(10)=44725.318, p<.001$



Sentiment in different sections
by CSAT response

CONTRIBUTIONS

I. Theoretical contributions



- Understanding meaning and trends of customer emotion dynamics in chat

II. Methodological contributions



- Introducing a service-adapted sentiment analysis tool

III. Managerial contributions



- Enabling prevention of customer dissatisfaction and service failures

LIMITATIONS AND FUTURE RESEARCH

- Model accuracy
- Detection of employee emotion
 - Emotional labor (Rafaeli & Sutton, 1987)
- Integration of other aspects of customer behavior
 - key strokes, customer engagement history
- Customer emotion and employee performance
 - (Altman, 2017; Ashtar, 2017)

Contact: westphal@campus.technion.ac.il



Anat Rafaeli



Galit Yom-Tov



Shelly Ashtar



Daniel Altman



Michael Natapov



Neta Barkay

