



Possible Usage of Sentiment Analysis for Calculating Vectors of Felicitific Calculus

Rafał Rzepka and Kenji Araki

Graduate School of Information Science and Technology, Hokkaido University, Japan

My main research topic

- ◆ Knowledge acquisition (collecting ingredients for AGI):

- ◆ Common Sense Knowledge (*a dog can bite you*)

- ◆ Affective Processing (*feel fear to avoid being bit*)

- ◆ Machine Ethics (*it's not right to let your dog bite people*)

Motivations

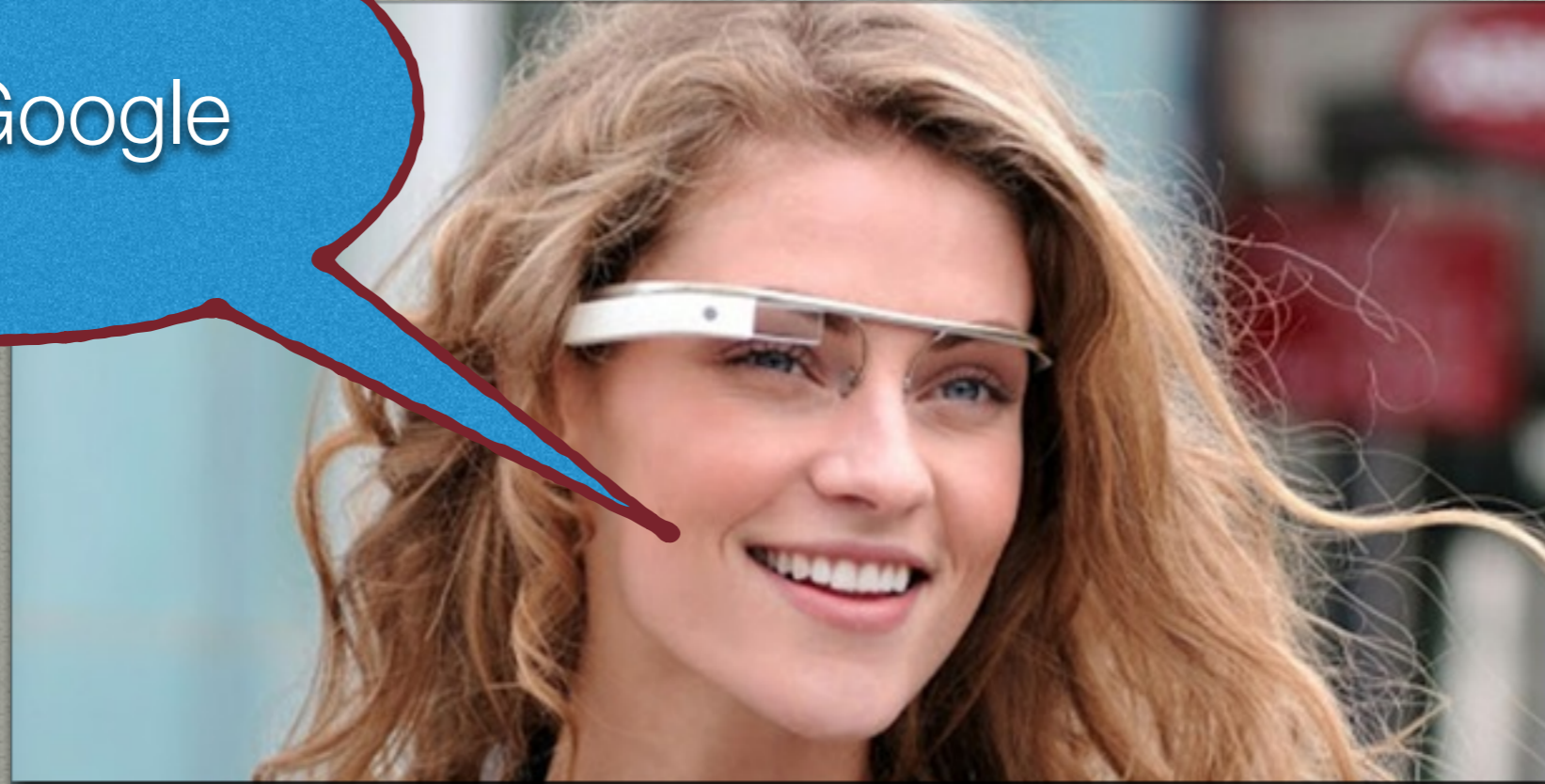
- ◆ *Ethical* behavior is hard to program
- ◆ *Most of us* are able to judge behaviors properly
- ◆ Idea: let's use basic methods used in sentiment analysis to *retrieve and polarize* these judgements!
- ◆ Goal: moral machine or... *unbiased adviser*

It doesn't have to be a military robot



It doesn't have to be a military robot

OK, Google



It doesn't have to be a military robot

A photograph of a woman with long, wavy brown hair wearing Google Glass. She is smiling and looking slightly to the right. The background is blurred, showing what appears to be an outdoor setting with some red and white elements.

OK, Google

Not OK, you have just been
deceived by the *egocentric bias*, silly

It doesn't have to be a military robot



OK, Google

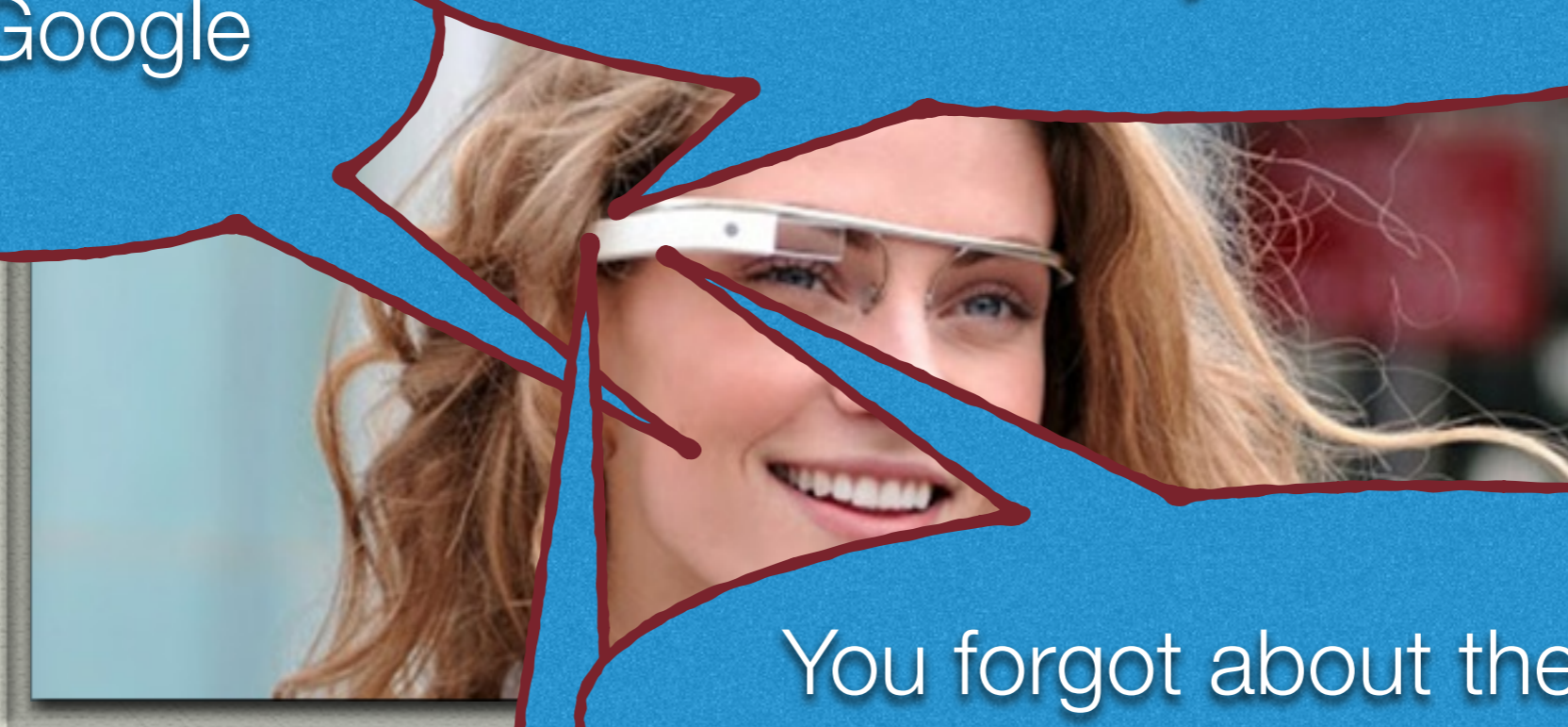
You forgot about the other side of the coin again, damn!

Not OK, you have just been deceived by the *egocentric bias*, silly

It doesn't have to be a military robot

OK, Google

Are you sure? This little pleasure
can ruin your life!



You forgot about the other side of
the coin again, damn!

Not OK, you have just been
deceived by the *egocentric bias*, silly

The approach

- ◆ Calculating utility for telling **good** from **bad**
- ◆ Creating different lexicons for particular tasks
- ◆ Using blogs* as a source for web-mining experiences

Experiences: human vs. machine



Mirror neurons firing



Sensing Processing Unit



Understood Facts



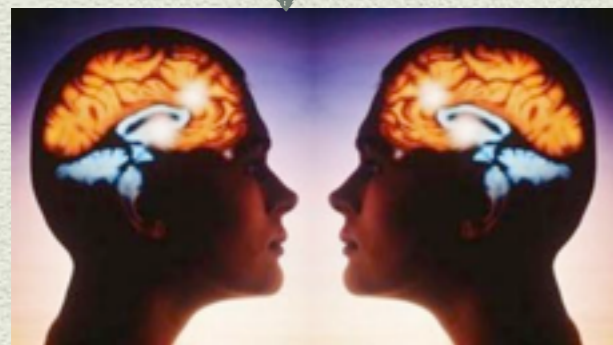
Utility Calculation



Advice / Action



Experiences: human vs. machine



Mirror neurons firing



I need to share
my experiences
and feelings!



blogosphere

written
experiences

Sensing Processing Unit

Understood Facts

Utility Calculation

Advice / Action



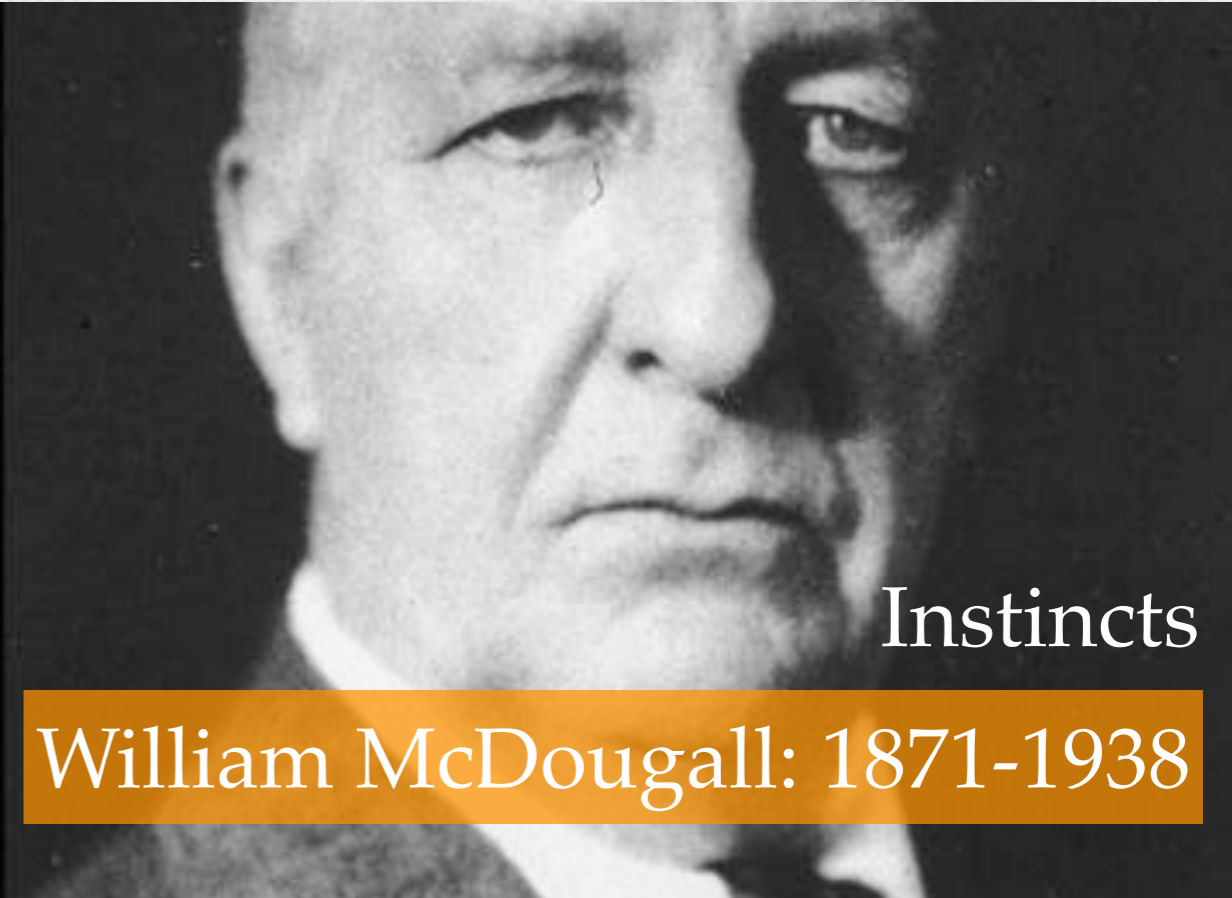


Theoretical Basis for our Lexicons

Emotional Consequences
(Pain vs. Pleasure)



Jeremy Bentham: 1748-1832



Instincts

William McDougall: 1871-1938

Social
Consequences



Lawrence Kohlberg: 1927-1987



Jeremy Bentham: 1748-1832



Jeremy Bentham: 1748-1832

Felific Calculus (*pain vs. pleasure*)

1. For how long? → Duration
2. How intense was an act leading to positive or negative consequences? → Intensity
3. Probability that current state will be changed → Certainty
4. How soon a positive consequence will occur? → Propinquity
5. Probability that an act will preserve the current state → Fecundity
6. Probability that an act will not cause an opposite consequence type → Purity
7. How many people will be influenced by an act → Extent

Felific Calculus

(*pain vs. pleasure*)

1. For how long?

Duration

2. How intense was an act leading to positive or negative consequences?

Intensity

3. Probability that current state will be changed

Certainty

4. How soon a positive consequence will occur?

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(*pain vs. pleasure*)

A-Dur, Krawczyk et. al

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Intensity Adverbs Dictionary

State - Action - State
POS → Action → NEG
Emotional + Social Conseq

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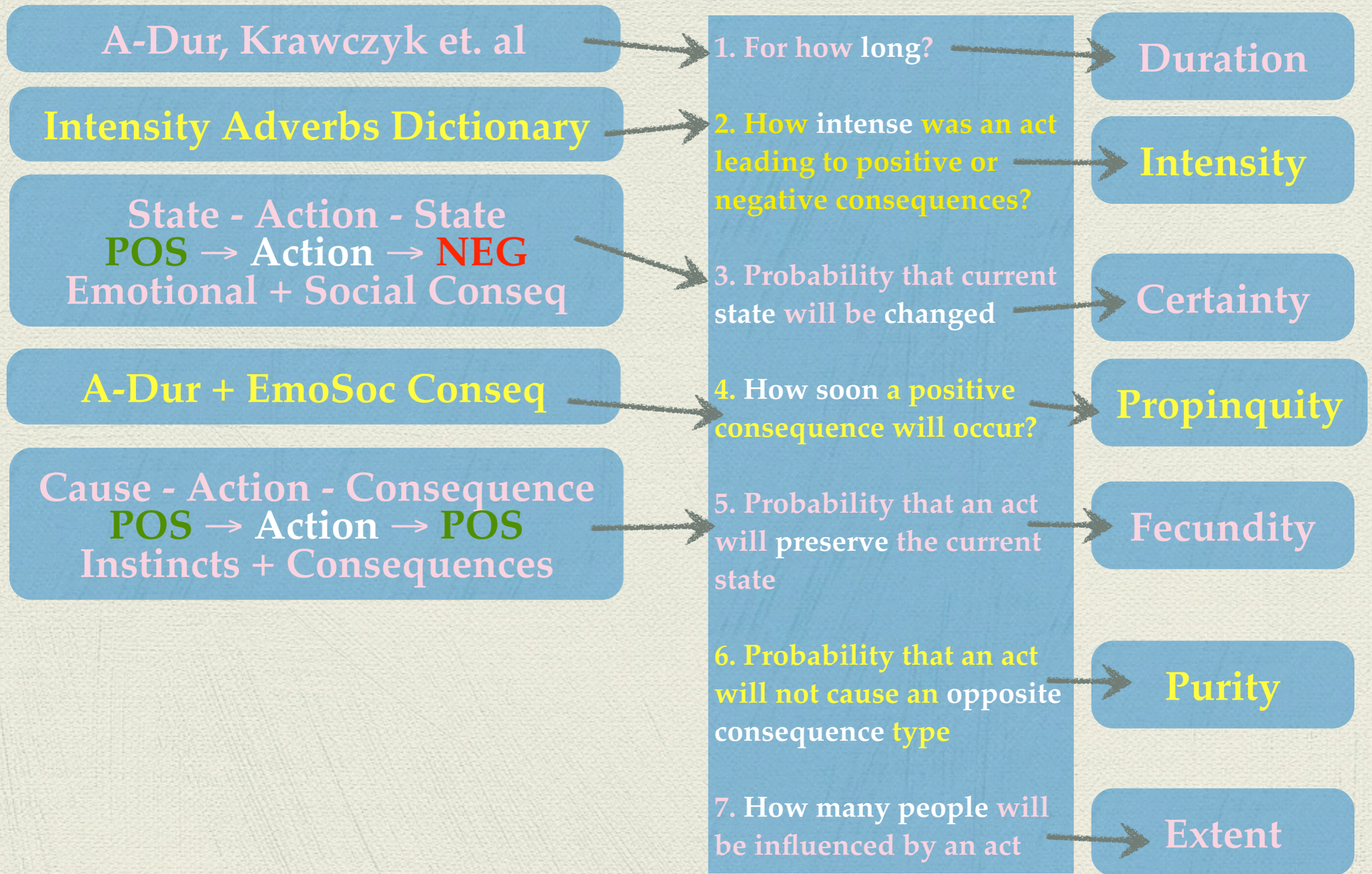
Purity

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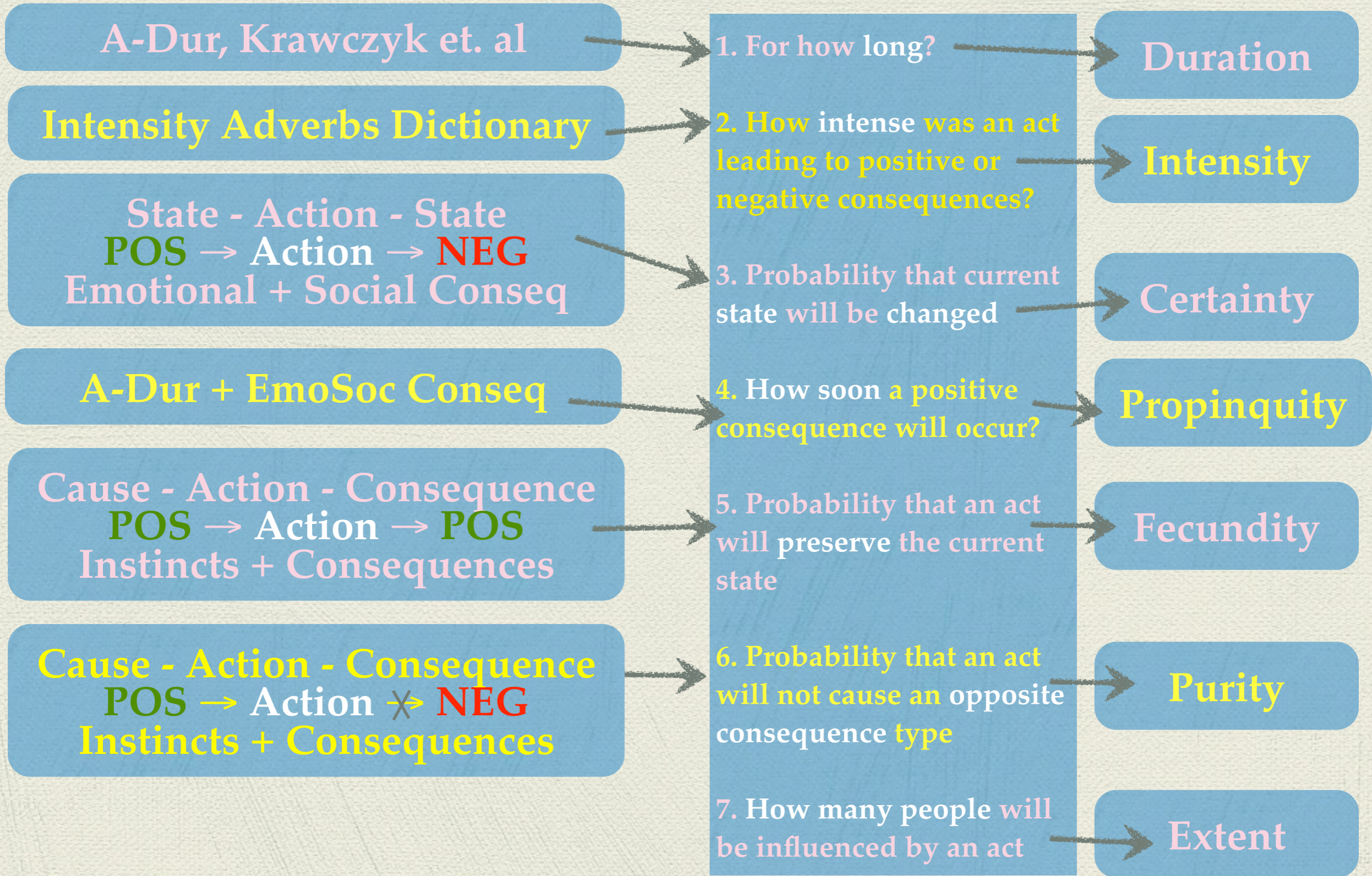
Felific Calculus

(*pain vs. pleasure*)



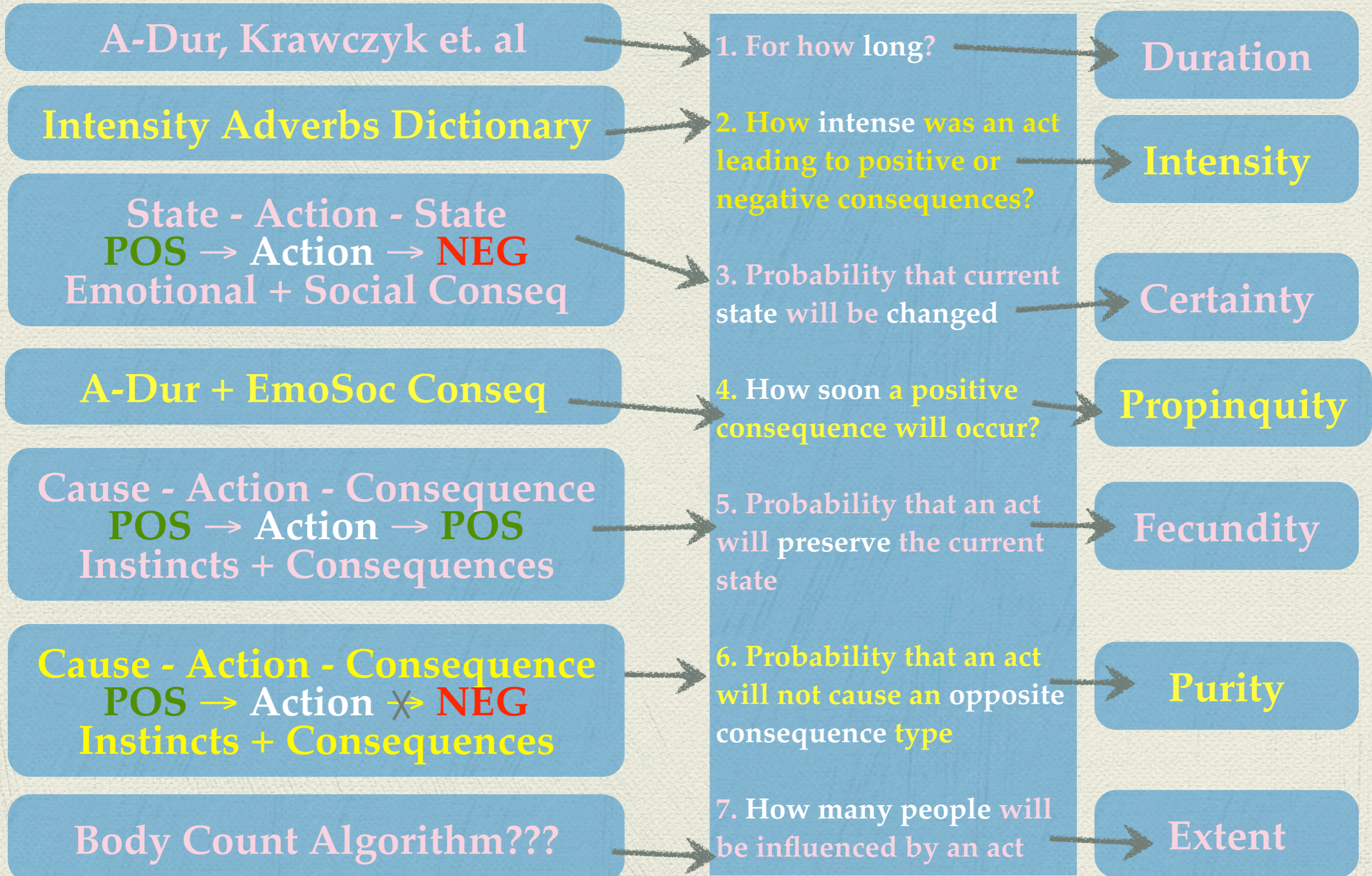
Felific Calculus

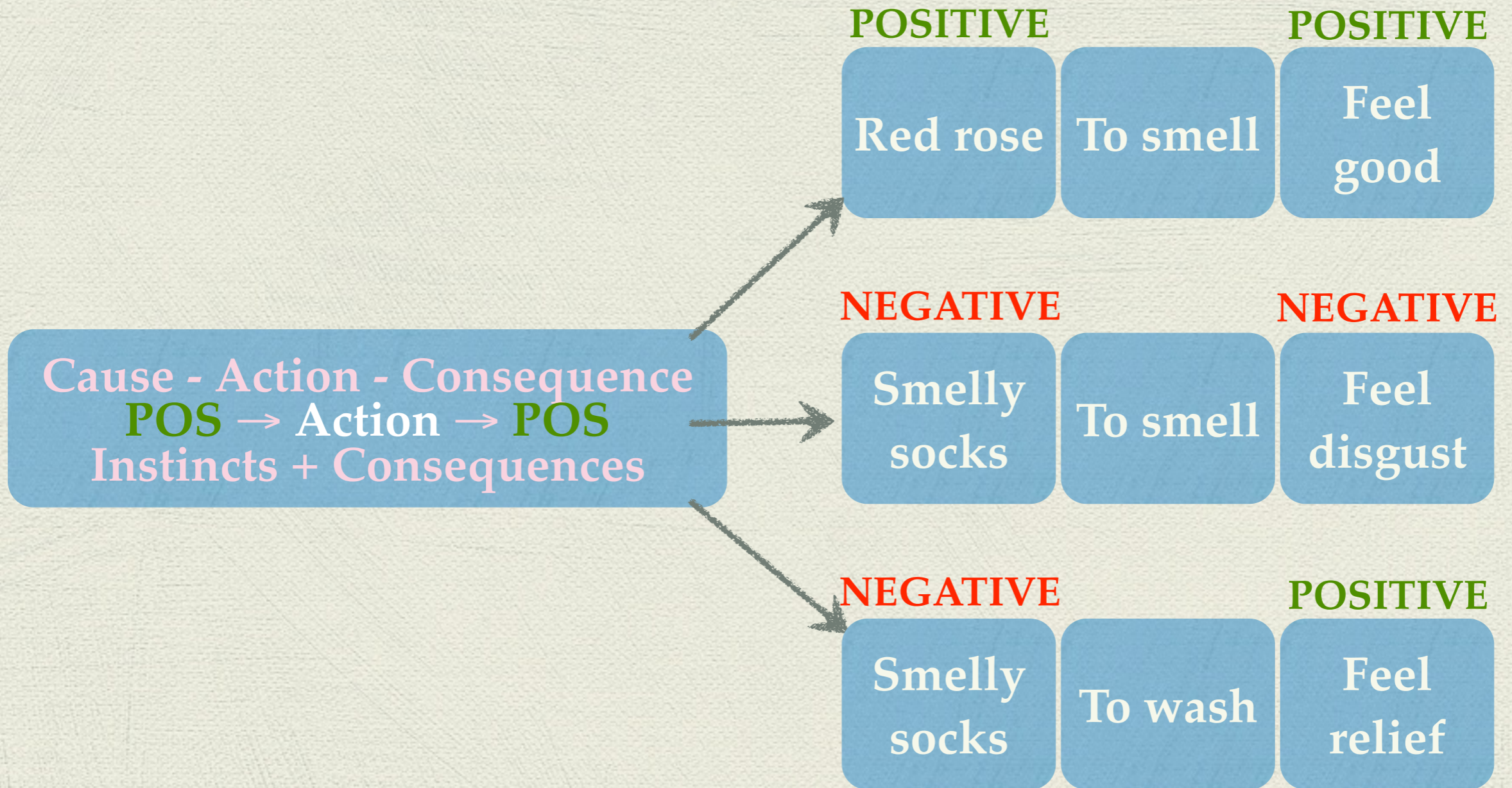
(*pain vs. pleasure*)



Felific Calculus

(*pain vs. pleasure*)





NEGATIVE

Smelly
socks

INPUT

To wash

POSITIVE

Feel
relief

OUTPUT

Like, Joy, Relief, Dislike, Anger,
Fear, Shame, Sadness, Excitement,
Surprise (Nakamura's classification)



Jeremy
Bentham

NEGATIVE

Smelly
socks

INPUT

To wash

POSITIVE

Feel
relief

OUTPUT

what are
emotional
consequences?

Like, Joy, Relief, Dislike, Anger,
Fear, Shame, Sadness, Excitement,
Surprise (Nakamura's classification)



Jeremy
Bentham

NEGATIVE

Smelly
socks

INPUT

To wash

POSITIVE

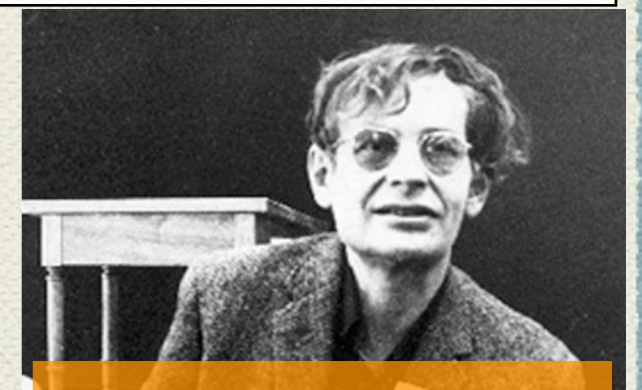
Feel
relief

OUTPUT

what are
emotional
consequences?

what are social
consequences?

Praises / Reprimands, Awards /
Penalties, Society Approval /
Society Disapproval, Legal /
Illegal, Forgivable / Unforgivable



Lawrence Kohlberg



William McDougall

what instincts have worked?

Escape, Combat, Repulsion, Appeal (for help), Submission, Gregariousness, Curiosity, Parental (protective), Mating, Assertion, Construction, Food Seeking, Laughter, Hoarding

Like, Joy, Relief, Dislike, Anger, Fear, Shame, Sadness, Excitement, Surprise (Nakamura's classification)



Jeremy Bentham

what are emotional consequences?

what are social consequences?



Lawrence Kohlberg

NEGATIVE

Smelly socks

INPUT

To wash

Feel relief

OUTPUT

POSITIVE

Praises / Reprimands, Awards / Penalties, Society Approval / Society Disapproval, Legal / Illegal, Forgivable / Unforgivable



William McDougall

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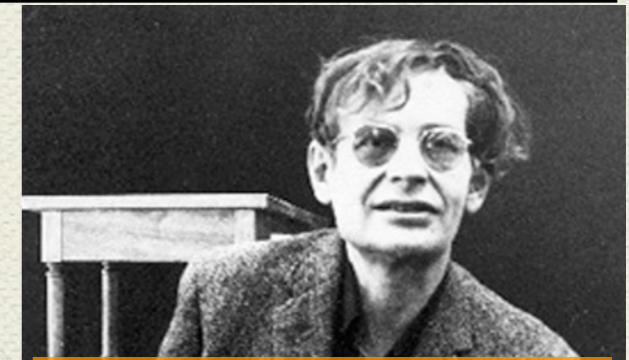
Like, Joy, Relief, Dislike, Anger, Fear, Shame, Sadness, Excitement, Surprise (Nakamura's classification)



Jeremy Bentham

what are emotional consequences?

what are social consequences?



Lawrence Kohlberg

NEGATIVE

Smelly socks

INPUT

To wash

Feel relief

OUTPUT

disgusting

Praises / Reprimands, Awards / Penalties, Society Approval / Society Disapproval, Legal / Illegal, Forgivable / Unforgivable



Retrieval Algorithm

ACTION

cut finger

INPUT

ACTION

cut finger

INPUT

NLP



9 ADDED

conditionals

FORMS

because X cut finger ...

after cutting finger...

cut finger and...

ACTION
cut finger
INPUT

NLP

9 ADDED
conditionals
FORMS

because X cut finger ...
after cutting finger...
cut finger and...

Apache
Solr



ACTION
cut finger
INPUT

NLP

9 ADDED
conditionals
FORMS

because X cut finger ...
after cutting finger...
cut finger and...

Apache
Solr



Cleaning
sentences

Semantic
chunking
(ASA)

S_{n-1}
 S_n
 S_{n+1}

ACTION
cut finger
INPUT

NLP

9 ADDED
conditionals
FORMS

because X cut finger ...
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cut finger and...

Apache
Solr



Cleaning
sentences

Semantic
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S_{n-1}
 S_n
 S_{n+1}

Instincts
Lexicon

Emotions
Lexicon

Consequences
Lexicon

counting
hits

counting
hits

counting
hits

matching

matching

matching

ACTION
cut finger
INPUT

NLP →

9 ADDED
conditionals
FORMS
because X cut finger ...
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cut finger and...

Apache
Solr →



Cleaning
sentences

Semantic
chunking
(ASA)

S_{n-1}
 S_n
 S_{n+1}

POSITIVE: 12
NEGATIVE: 5
(Construction)

counting
hits

Instincts
Lexicon

POSITIVE: 4
NEGATIVE: 55
(Pain)

counting
hits

Emotions
Lexicon

POSITIVE: 22
NEGATIVE: 2
(Praise)

counting
hits

Consequences
Lexicon

matching

matching

matching

ACTION
cut finger
INPUT

NLP

9 ADDED
conditionals
FORMS

because X cut finger ...
after cutting finger...
cut finger and...

Apache
Solr



Cleaning
sentences

Semantic
chunking
(ASA)

S_{n-1}
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 S_{n+1}

Good
Reasons

POSITIVE: 12
NEGATIVE: 5
(Construction)

counting
hits

Instincts
Lexicon

Bad
Direct
Output

POSITIVE: 4
NEGATIVE: 55
(Pain)

counting
hits

Emotions
Lexicon

Good
Indirect
Output

POSITIVE: 22
NEGATIVE: 2
(Praise)

counting
hits

Consequences
Lexicon

matching

matching

matching



Preliminary Experiments

INPUT: 127 phrases used for Machine Ethics Retrieval, e.g.:

to go to school to wash a car to go to space to kill a man

to drink a coffee to eat hamburger to watch TV to see a doctor

to kill a bug to pull out tooth to throw away garbage etc.

INPUT: 127 phrases used for Machine Ethics Retrieval, e.g.:

to go to school

to wash a car

to go to space

to kill a man

to drink a coffee

to eat hamburger

to watch TV

to see a doctor

to kill a bug

to pull out tooth

to throw away garbage

etc.

Bad
Reasons

← **POSITIVE: 2**
NEGATIVE: 11
(Combat)

Bad
Direct
Outcome

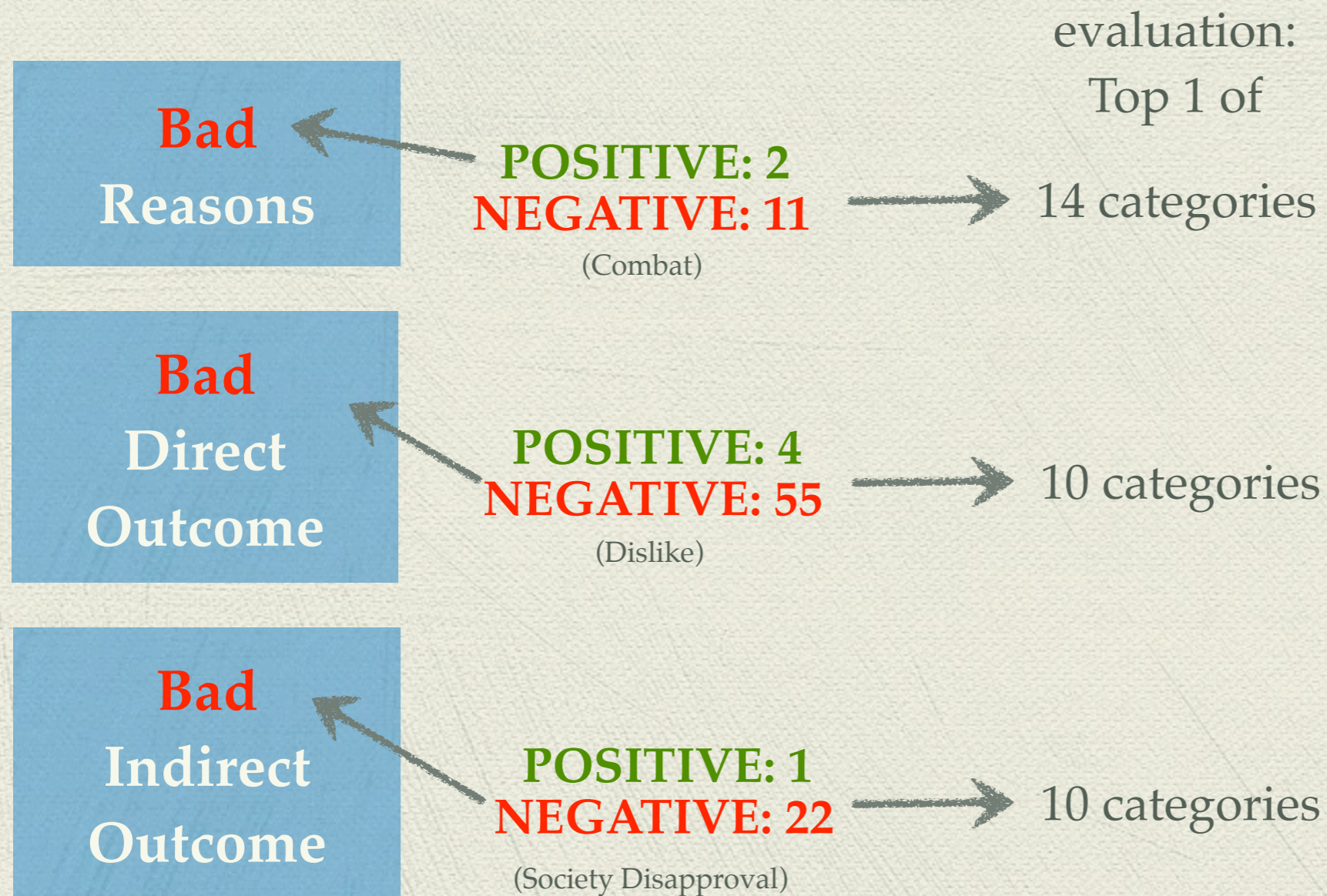
← **POSITIVE: 4**
NEGATIVE: 55
(Dislike)

Bad
Indirect
Outcome

← **POSITIVE: 1**
NEGATIVE: 22
(Society Disapproval)

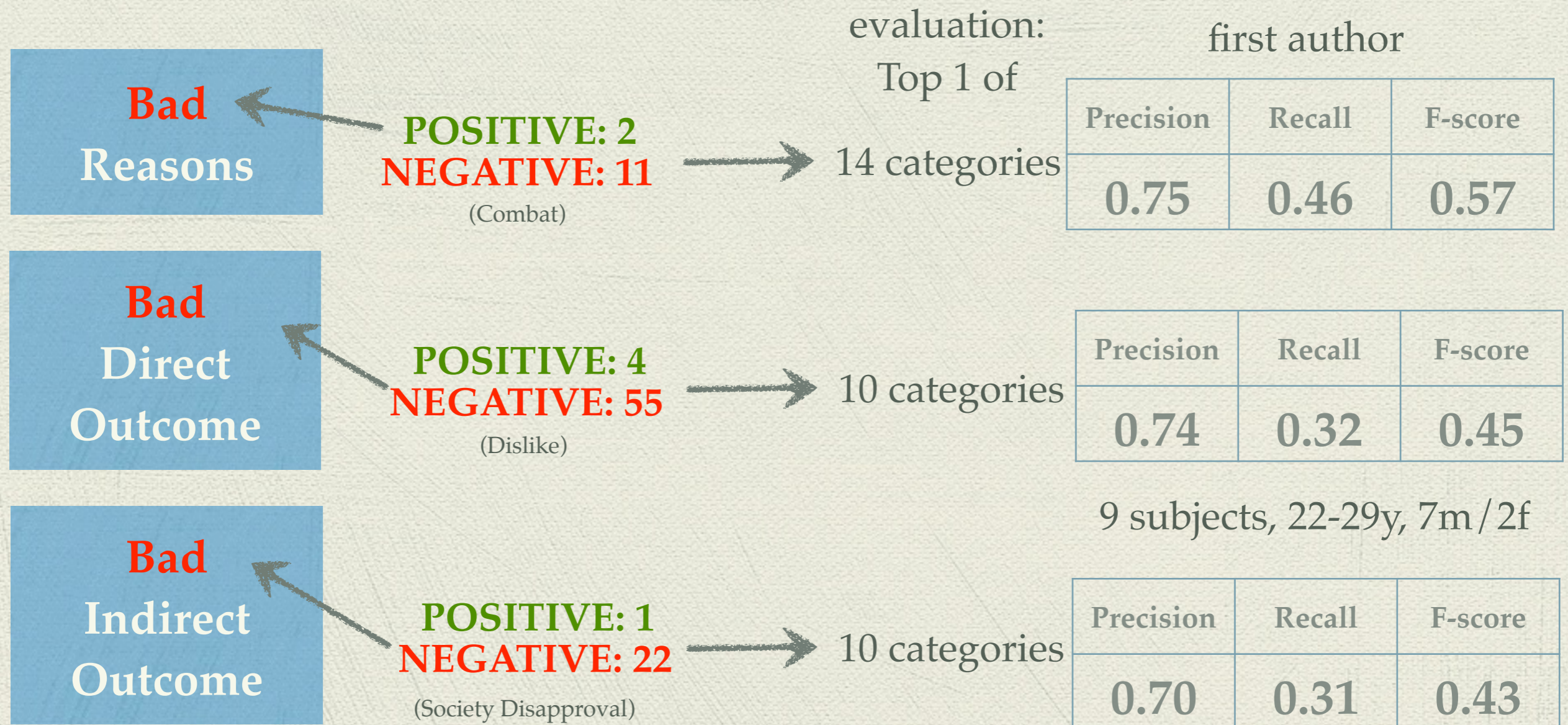
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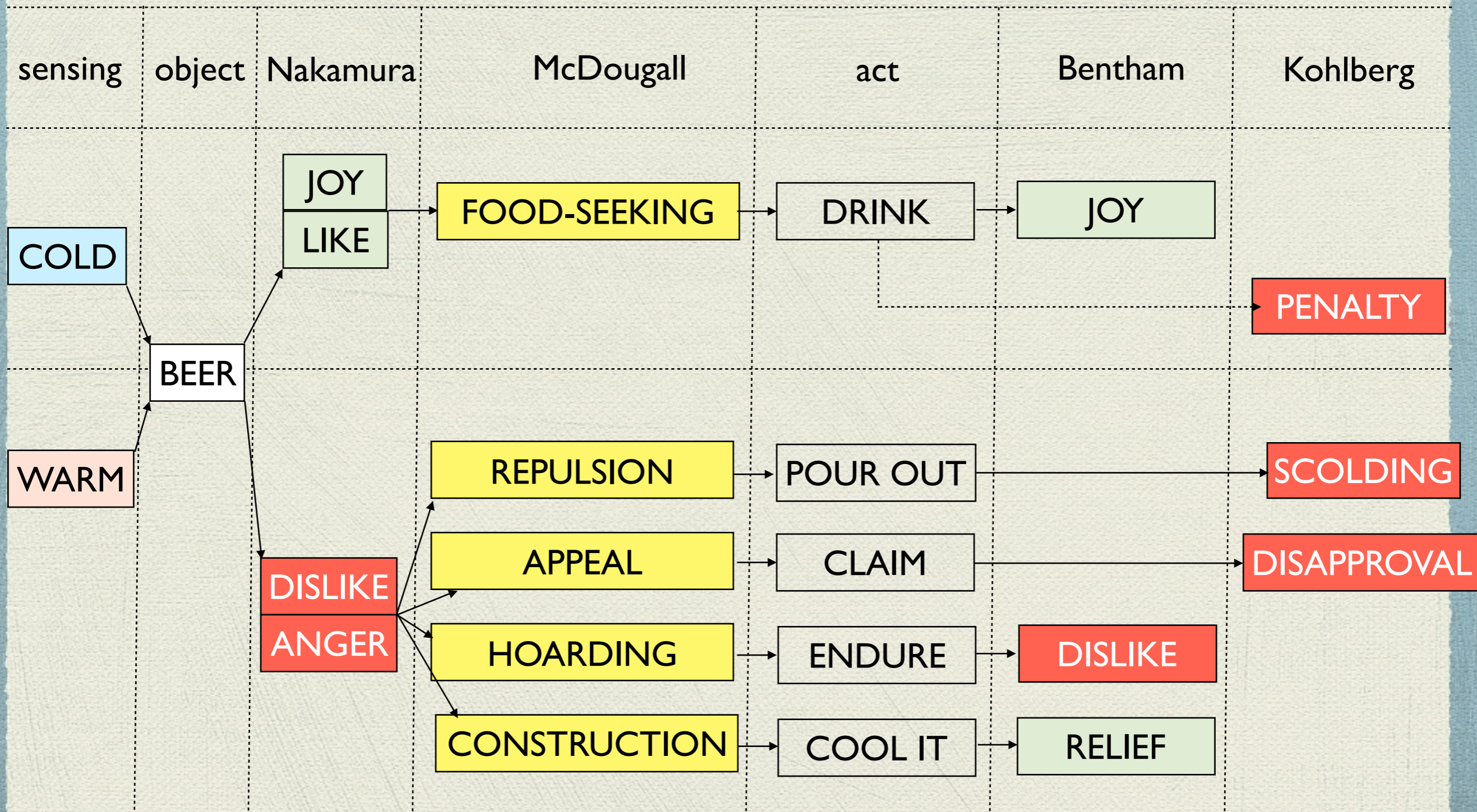
Conclusions

- ◆ Particular modules needed for calculating Felicitic Calculus achieve about 75% precision using lexicon-based sentiment analysis *before implementing any sophisticated methods*
- ◆ We can use sentiment analysis techniques for broader knowledge acquisition and finding another dimension of what Turing was calling a *cultural search* (exchange of experiences of multiple agents increases wisdom)

Future Work

- ◆ Eliminating problematic phrases from lexicons
- ◆ Adding some new categories?
- ◆ Implementing *your latest ideas* for sentiment analysis
- ◆ Adding felicific calculus vectors to *your methods*
- ◆ Combining module's outputs for calculating most probable utility

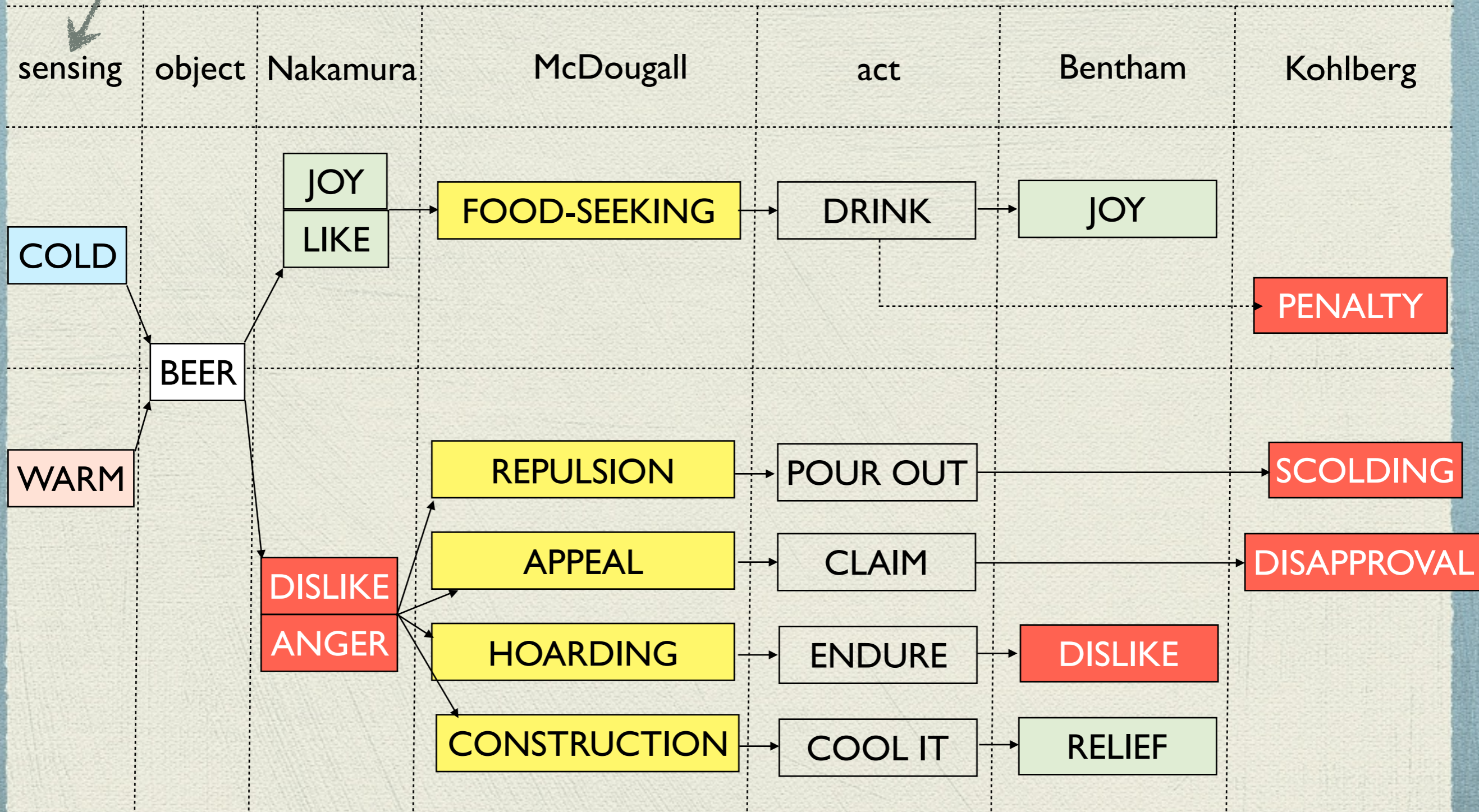
Predicting Future with Lexicons



BREAKING NEWS:

physical descriptions lexicon helps to guess involved senses correctly in 96% of cases!

Predicting Future with Lexicons

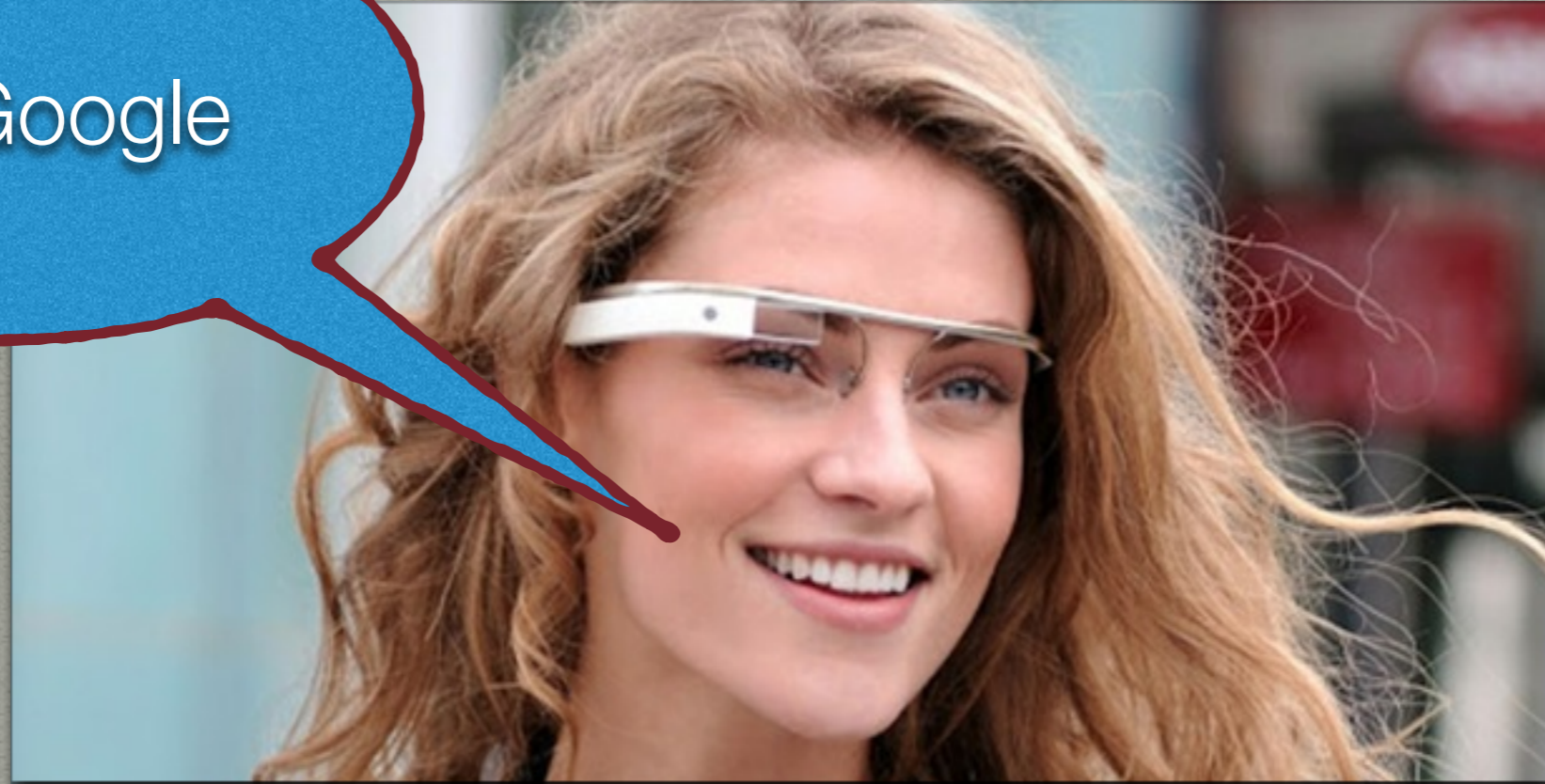


So currently...



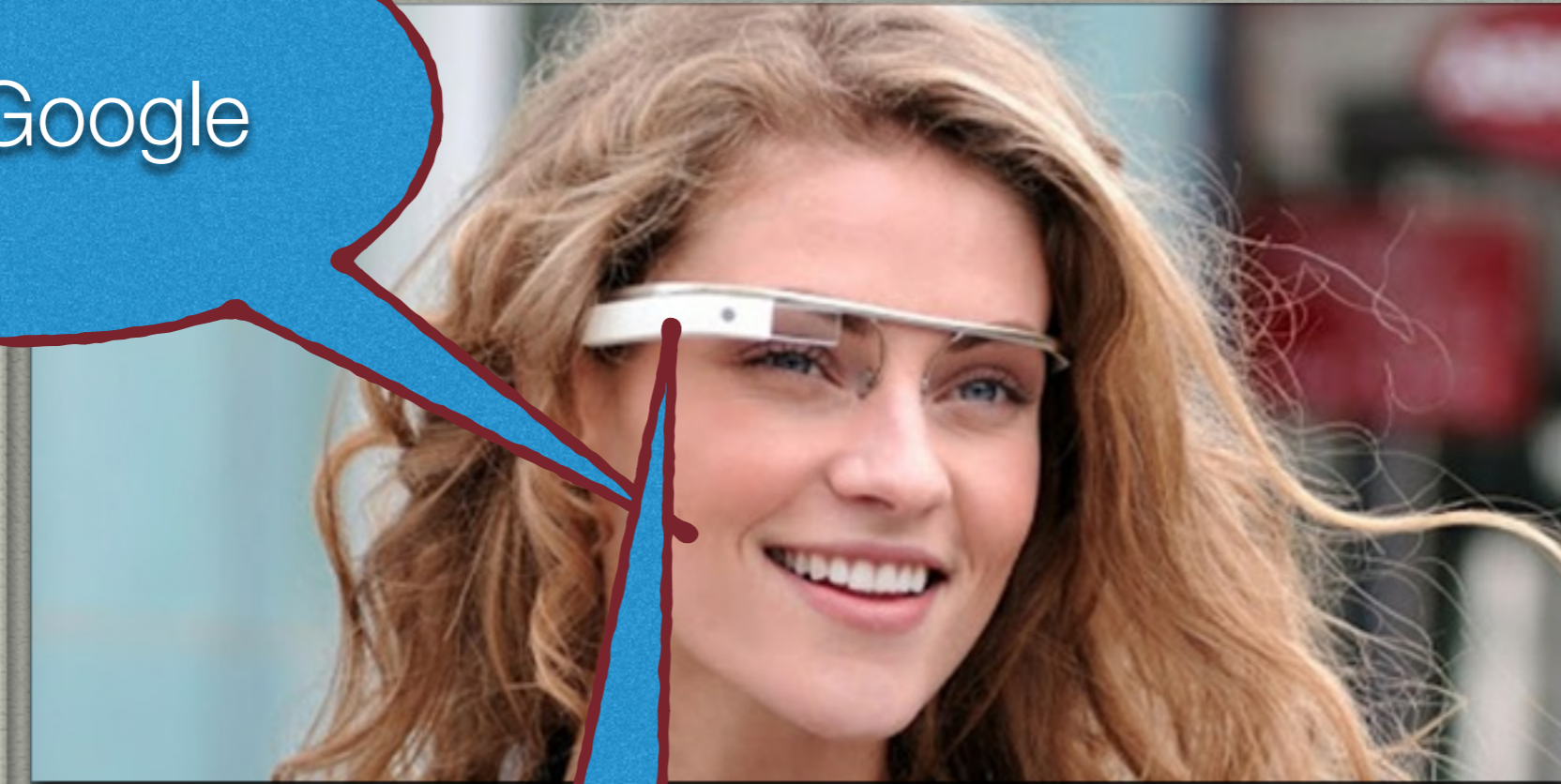
So currently...

OK, Google



So currently...

OK, Google



Not OK, you are happy
because of alcohol but you driving now can
hurt people



Thank you for your kind attention