

Do Diverse Social Interactions Make Us Smile More Often?

Studying Smiles and Diversity Via Social Media Photos



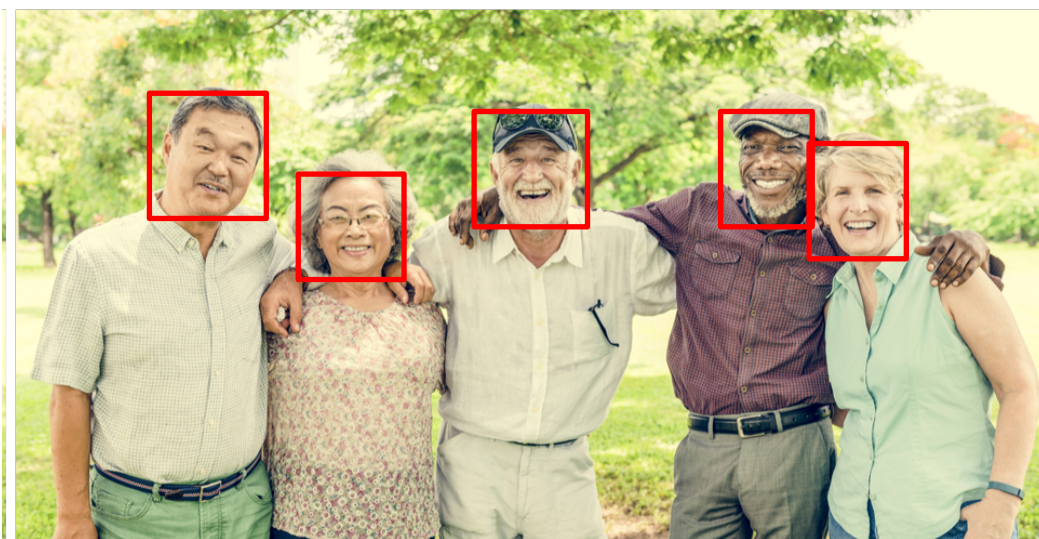
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Motivation

- Photographs → one of the most fundamental ways for human beings to capture social experiences.
 - Provide unique opportunity to study the social phenomena of mixing of different people and the smiles expressed by individuals in these social settings.



Smiles ← ? → Diversity

- Advancing the understanding of phenomena of smiles and diversity of social interactions
 - Scale to millions of individuals, not require human time, and capture the temporal dynamics

Objectives

- RQ1:** How can the visual content on social media (photos) be used to study the interconnections between diversity and smiles?
- RQ2:** What are the effects of presence of others on the smile levels of the individuals in photos?
- RQ3:** What are the effects of diversity of social interactions in terms of age, race, and gender on the smile levels of individuals in photos?

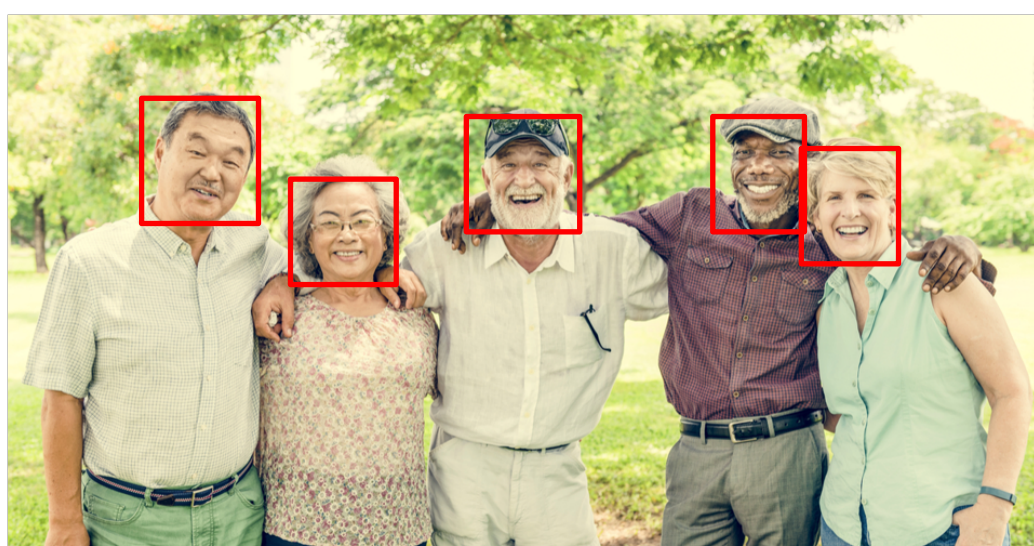
Novelty

- What has been done:**
 - Understanding diversity in terms of social interactions, not residential address
 - Using smile scores to judge emotion
- What has not been done:** Computationally (scale, automation) studying the effects of diversity on smiles.

Approach Overview



Face++ API Facial & Attributes Detection of Photo



Our Interpretation of Results	
Variable	Value
# of Males	3
# of Females	2
# of Whites	2
# of Blacks	1
# of Asians	2
<18 Age	0
>18 and <35 Age	0
>35 Age	5
Avg. Smile Score	81.47

Gender: Male Age: >35 Ethnicity: Asian Smile Score: 41.17
Gender: Female Age: >35 Ethnicity: Asian Smile Score: 96.41
Gender: Male Age: >35 Ethnicity: White Smile Score: 89.02
Gender: Male Age: >35 Ethnicity: Black Smile Score: 88.68
Gender: Female Age: >35 Ethnicity: White Smile Score: 92.07

Approach

- Collect Instagram images via Instagram's public API
- Collect Twitter images via Twitter's Stream API
- Recognize & remove non-facial and tourist images
- Obtain variables from Face++ API of all individuals
- Validate race detection using Morph dataset
- Validate smile scores using human annotation comparison
- T-test between smile scores for photos with single and multiple people
- OLS Linear Regression model for smile score as a function of demographic (control) variables and diversity
- Obtain the number of face in each image
- Compute smile scores (dep. variable) of each image
- Compute interaction diversity metric (SEAge, SEGender, SERace) for all images

Results

- T-test** between smile scores for photos with single and multiple people
 - Results across the two data sets consistently suggest that people tend to smile more in the presence of others.

	Groups	N	Mean	Std. Dev	Std. Error	Model Significance (p-value)	T-test between the two groups
Instagram	Multiple	3157	52.80	28.83	0.51	<0.001	T = 26.06
	Single	4035	33.75	33.09	0.52	<0.001	
Twitter	Multiple	3688	53.70	27.04	0.45	<0.001	T = 31.34
	Single	4380	33.07	32.10	0.49	<0.001	

- Linear regression** for smile score as a function of demographic (control) variables and diversity variables
 - Results indicate that people smile MORE in DIVERSE social company
 - POSITIVE coefficients observed for MIXING
 - Broad level consistencies between the datasets

Limitations

- Posed and shared smiles
- Limited demographic descriptors
- Modest effect sizes

Implications

- Supports experiments in multiple social sciences
 - Scalable, automated, lower cost, “in-the-wild”
- Contextual understanding for modeling faces and their dynamics in multimedia systems
- Building blocks to a more smiling, happier society.