

MODERN TASKS OF COMPUTER VISION RESEARCH

Andrey V. Savchenko

Dr. of Sci., Prof.,

HSE University, N. Novgorod, Russia

Email: avsavchenko@hse.ru

URL: www.hse.ru/en/staff/avsavchenko



- 1. Brief overview of 2D CV tasks
- 2. User Preference Prediction in Visual Data on Mobile Devices
- 3. Facial processing and emotion recognition
- 4. Food classification and restaurant recommendation
- 5. Understanding advertisements



Brief overview of 2D CV tasks



Image recognition



(a) Siberian husky



(b) Eskimo dog

Wolfram ImageIdentify



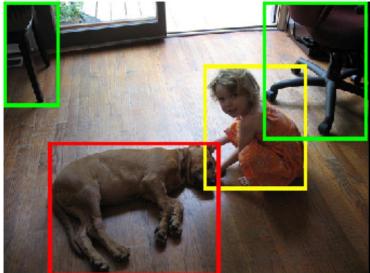




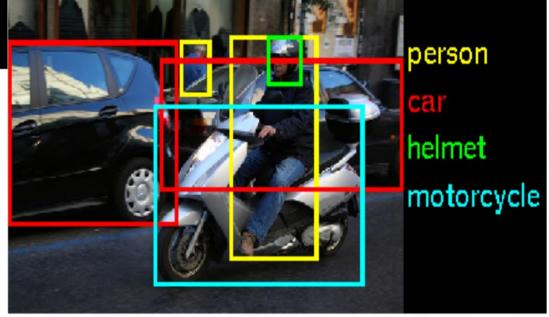
bastion



Object detection

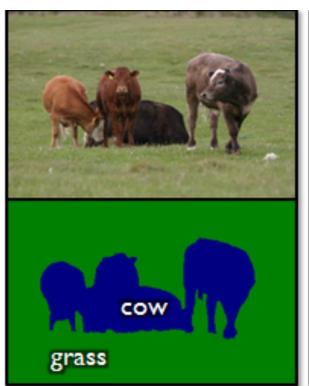


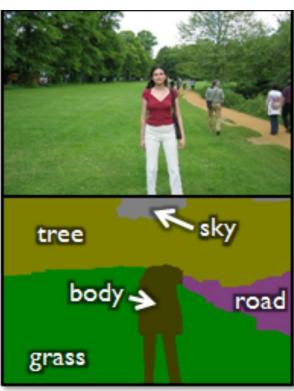
person dog chair

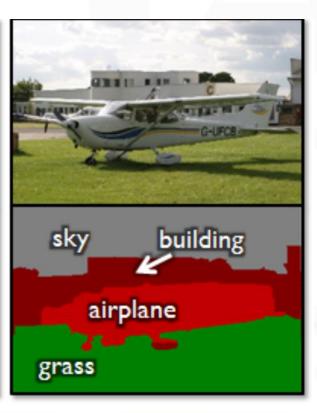




Semantic segmentation

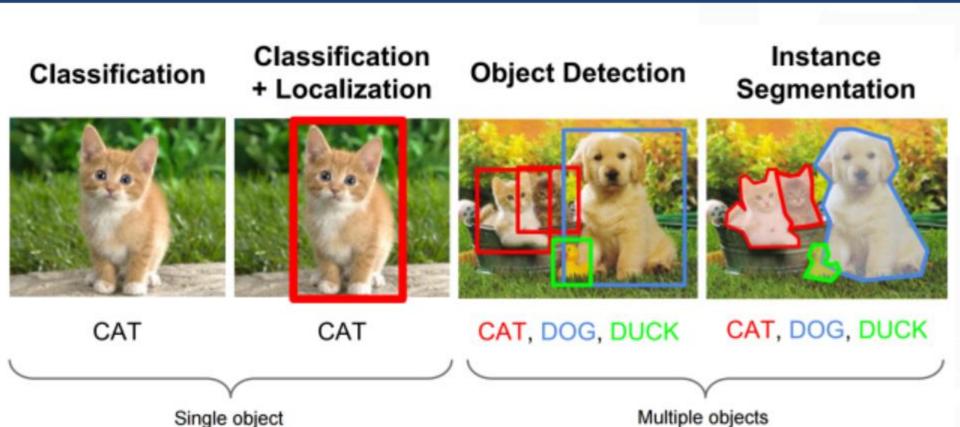






object classes	building	grass	tree	cow	sheep	sky	airplane	water	face	car
bicycle	flower	sign	bird	book	chair	road	cat	dog	body	boat

Classification, detection, segmentation,...



https://leonardoaraujosantos.gitbooks.io/artificialinteligence/content/object_localization_and_detection.html



Image captioning

Describes without errors



A person riding a motorcycle on a dirt road.

A group of young people

playing a game of frisbee.

A herd of elephants walking

across a dry grass field.



Describes with minor errors

Two dogs play in the grass.



Two hockey players are fighting over the puck.



A close up of a cat laying on a couch.

Somewhat related to the image



A skateboarder does a trick on a ramp.



A little girl in a pink hat is blowing bubbles.



A red motorcycle parked on the side of the road.

Unrelated to the image



A dog is jumping to catch a frisbee.



A refrigerator filled with lots of food and drinks.



A yellow school bus parked in a parking lot.

http://arxiv.org/abs/1411.4555 "Show and Tell: A Neural Image Caption Generator"



Style transfer















Prisma







Other tasks

Recognition of paintings



- https://quickdraw.withgoogle.com/
- https://github.com/tensorflow/docs/blob/master/site/ en/r1/tutorials/sequences/recurrent_quickdraw.md
- MemNet, https://arxiv.org/abs/1708.02209
- LaMem, http://memorability.csail.mit.edu/demo.html

Predicting Image Memorability



Memorability: High

(score: 0.825 2)



User Preference Prediction in Visual Data on Mobile Devices

- Savchenko, Demochkin, Grechikhin, Pattern Recognition 2022
- Savchenko, Kopeykina, RusAutoCon 2019, ITNT 2020
- Savchenko, Rassadin, ISNN 2019
- Grechikhin, Savchenko, IbPRIA 2019
- Savchenko, Samsung US Patent 2022

Funded by

SAMSUNG



Deep understanding of user characteristics by analyzing user images and videos in a mobile device

Singer: Taylor Swift



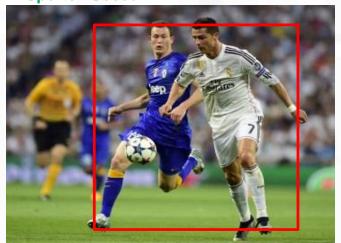
Food: Fried Chichen



Game: Nintendo Switch



Sports: Soccer





User's profile

Categorizing user's characteristics (taxonomy, classification, demographics, hobbies, occupation, lifestyle, etc.) → Generate user profile

Hobbies	Food code	Pets
Restaurant Beach Tracks Bar	Junk food Health - salad and etc. Sandwich Meat	Dog Cat Fish Horse

Sports	Household Income	Vacation
Fishing Golf Diving Tennis	Age Gender Car types household	Ski Museum Tracks Monuments
1		



Privacy detection

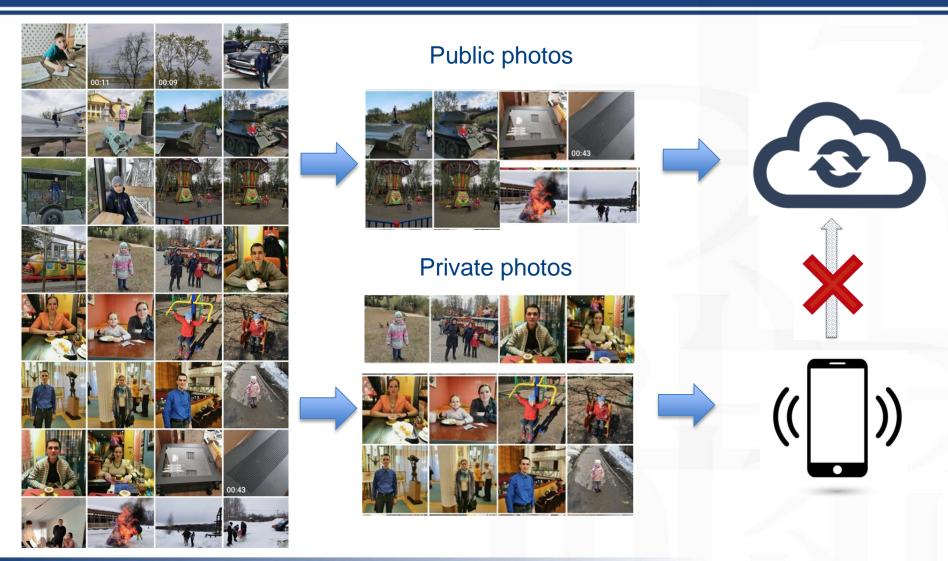
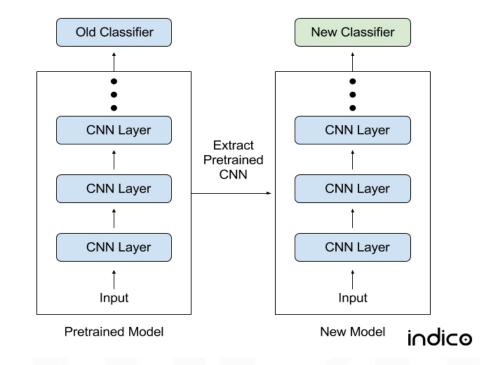


Image recognition problem

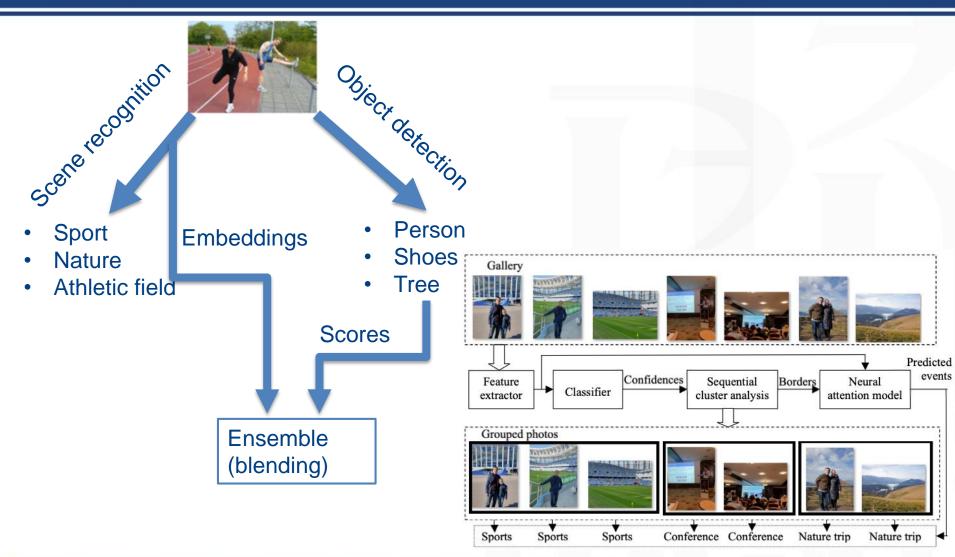
It is required to assign an observed image X to one of C classes. Training set contains N reference images (examples) $\{X_n\}$, $n \in \{1, ..., N\}$, with known class label $c_n \in \{1, ..., C\}$

- 1 Fine-tune convolutional neural network (CNN) pre-trained on ImageNet, Places, etc.
- Classify embeddings (features) from one of the last CNN's layers: Ddimensional feature vector x=[x₁,..., x_D]. Training set is associated with embeddings {x_n}





Classifier training using both visual features and results of object detection



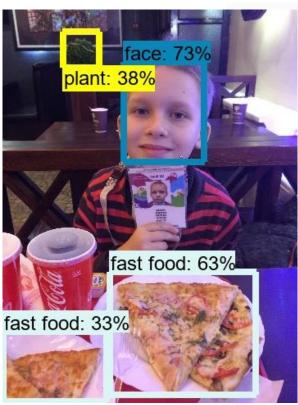


Example of object detection (1)

SSDLite+ MobileNet v2



Faster RCNN+ Inception v2



Faster RCNN+ InceptionResNet





Example of object detection (2)

SSDLite+ MobileNet v2



Faster RCNN+ Inception v2

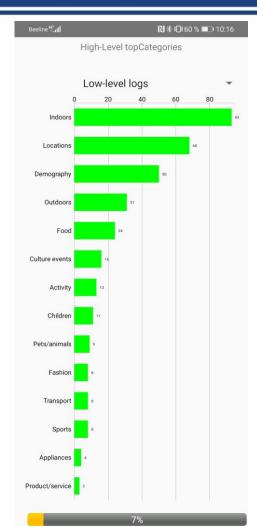


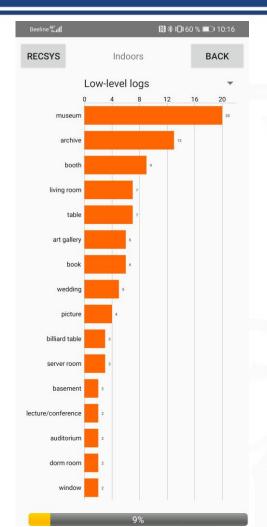
Faster RCNN+ InceptionResNet

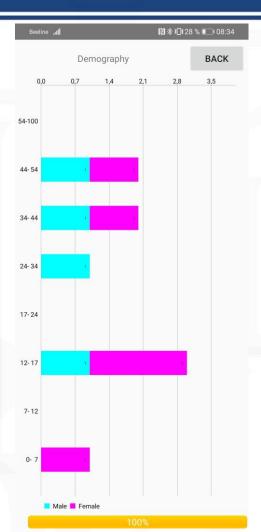




Mobile application (1)



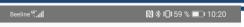




https://drive.google.com/drive/folders/1rQkJZifq_89pu0sT_UnYXziuxutTpNEN



Mobile application (2)





PREV

NEXT

BACK

photo 6 out of 33 Private photo Selfie Nizhnij Novgorod, Russia

no objects found scenes: football (0,97); events: sports (0,26); orig topCategories:scenes: football (0,94); events: sports (0,00); me: age=29 male white child 1: age=25 male white

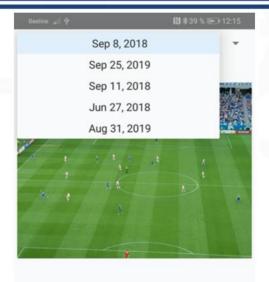
child 1: age=25 male w text:

text



photo 10 out of 37 Private photo Nizhnij Novgorod, Russia

car (0,80) car (0,48) cars:Hyundai Rena(0,75) scenes: parking lot (0,52); roof garden (0,12); child 2: age=1 female white text:DAIHATSU



NEXT

BACK

photo 1 out of 2 Public photo latitude=0,000 longitude=0,000

PREV

no objects found scenes: football (0,98); events: sports (0,17); No faces found text:OAMMn JAKO Instat

OOAMH OL BAdSIRRU 1AAIH OLMP.BET

1009

100%

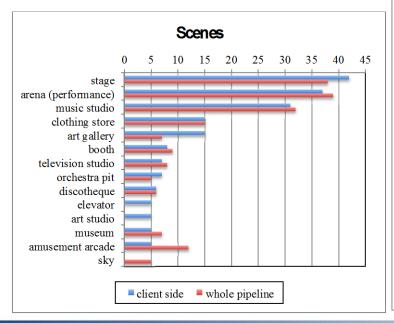


The Rolling Stones profile

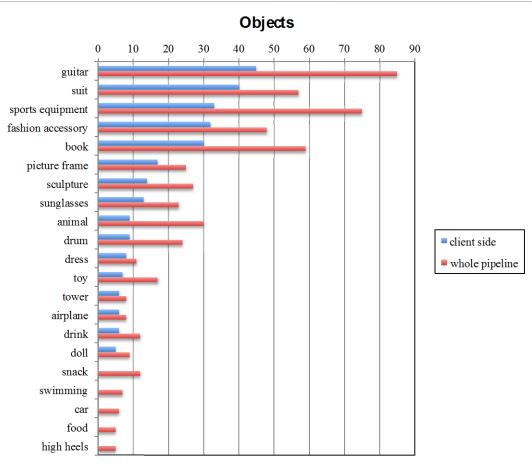




Scene recognition



Object detection





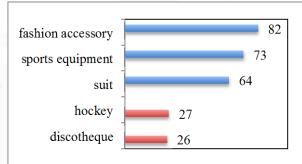
User's profiles from Instagram (1)

Alex Ovechkin

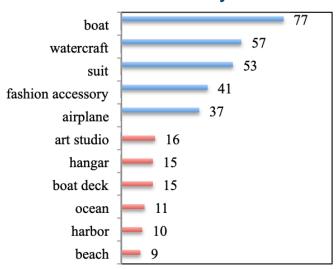




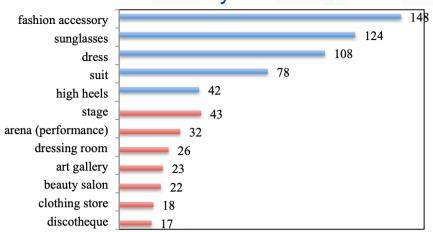




Fedor Konyukhov

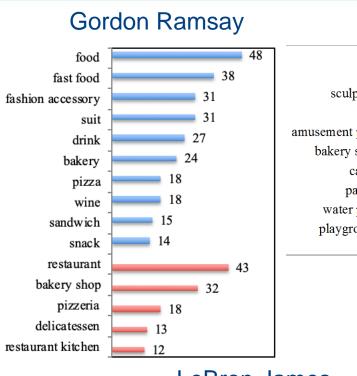


Beyonce

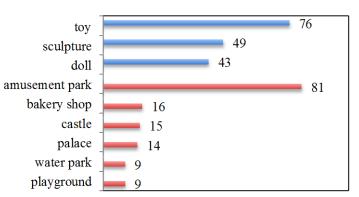




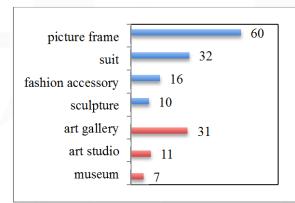
Profiles from Instagram (2)



Disneyland

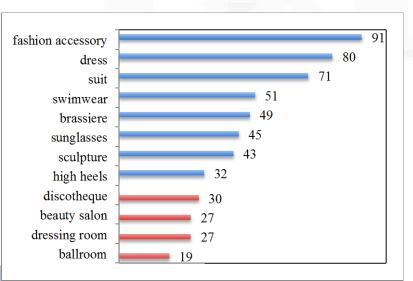


Kehinde Wiley



Kim Kardashian







Facial processing and emotion recognition

- Kharchevnikova, Savchenko, PeerJ Computer Science 2021
- Sokolova, Savchenko, AIST 2017, 2019, ITNT 2018, OMNN 2020
- Demochina, Savchenko, ICPR Workshop 2021

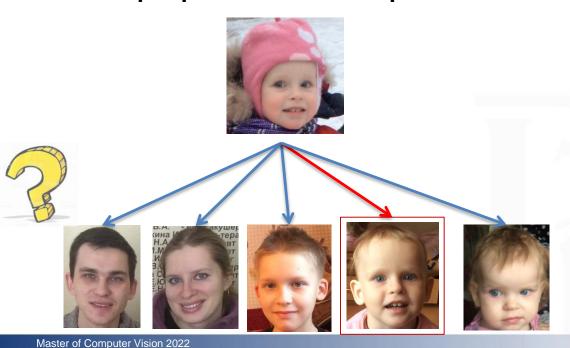
Churaev, Savchenko 2021



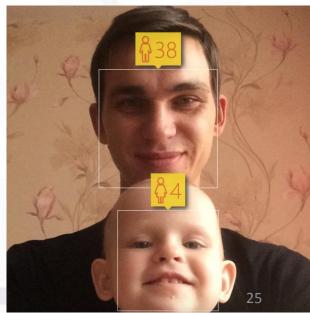
Facial processing tasks

- Detection
- Verification
- Identification
- Recognition of emotions, age, gender, race,...

Input photo or video sequence









Emotion analysis and engagement prediction in videos









https://www.youtube.com/watch?v=kORwkI-Oq7o



AFEW (Acted Facial Expression In The Wild)









- # of emotions: 7 (Neutral, Happy, Sad, Surprise, Fear, Anger, Disgust)
- Training set: 773 short video clips
- Validation set: 383 videos

https://sites.google.com/view/emotiw2019



VGAF (Video-level Group AFfect)



- # of classes: 3 (Positive, Negative, Neutral)
- Training set: 2661 videos
- Validation set: 766 videos

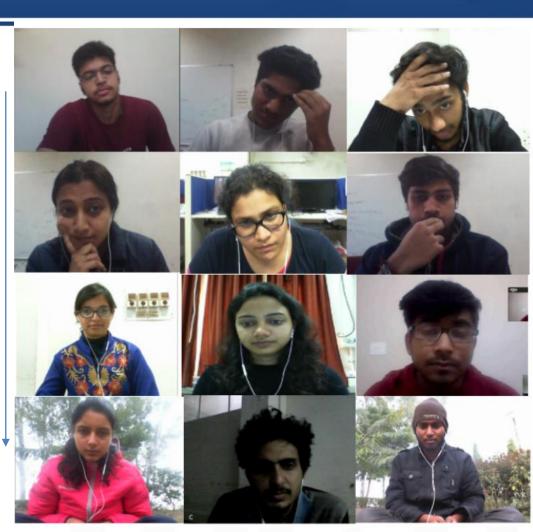
https://sites.google.com/view/emotiw2020



EngageWild: student's engagement prediction

- # of engagement levels: 4 (0 disengaged, 0.33, 0.66, 1 highly engaged)
- Training set: 147
 videos (~ 5 minutes)
- Validation set: 48 videos

Engagement level

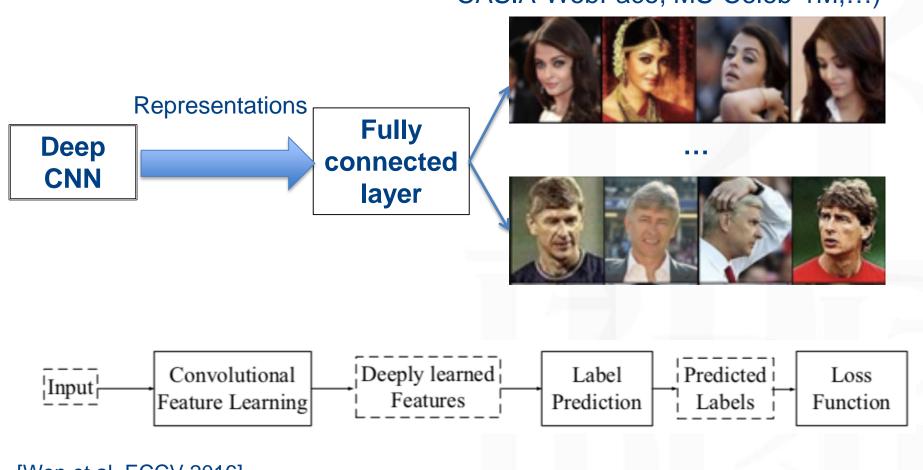


https://sites.google.com/view/emotiw2020



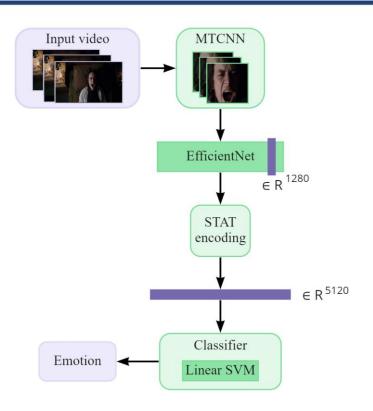
CNN (Convolutional Neural Networks) and domain adaptation: Deep faces

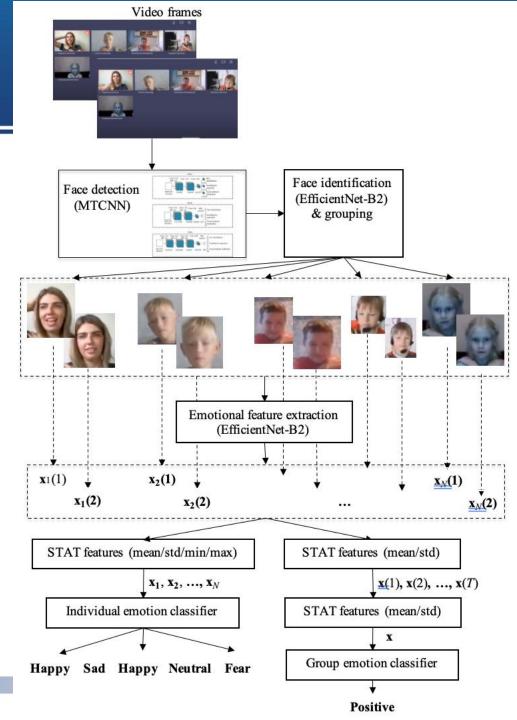
Large training set of celebrities (VGGFace2, CASIA-WebFace, MS-Celeb-1M,...)





Analysis of video conference







Android demo app





photo 47 out of 246
Private photo
latitude=0,000 longitude=0,000

footwear (0,39)
scenes:beach (0,41); boardwalk (0,29);
age=25 female white
age=38 male black
age=27 female white

child 2: age=2 male white

age=34 male black

text:



NEXT

photo 285 out of 1295 Public photo latitude=0,000 longitude=0,000

no objects found scenes:stage (0,71); age=33 female asian age=25 female asian age=24 female asian age=13 female asian text:

PREV





BACK

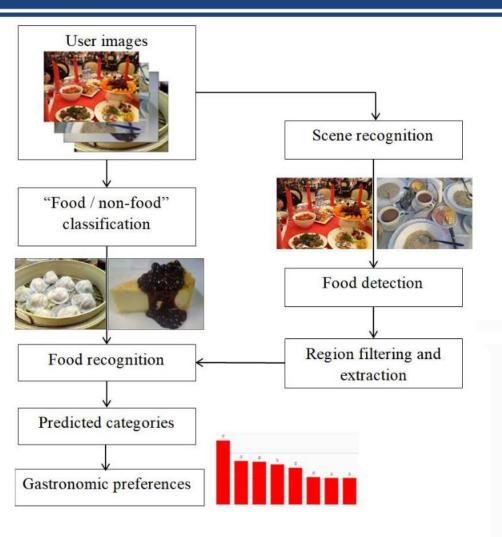


Food classification and restaurant recommendation

Miasnikov, Savchenko, ICIAR 2020



Food detection and recognition in a gallery of mobile device









Restaurant recommendation



Scenes/Events dining room (0.31) wedding (-0.51)

Objects

Food:0.768

Fast food:0.473

Fireplace:0.422

Fast food:0.407

Snack:0.398

Table:0.389 Snack:0.313

Snack:0.304

YELP label classification

food (0.86)

YELP photo classification

has_alcohol (0.93)

Select city for recommendation Las Vegas

Analyze



Scenes/Events restaurant (0.62) birthday (-0.06)

Objects

Food:0.692 Furniture:0.604 Bottle:0.602 Plate:0.518 Chopsticks:0.480 Fast food:0.459 Table:0.412 Plate:0.393 Tableware:0.361

YELP restaurants

has alcohol (0.92) has table service (0.97) YELP cuisine Chinese (0.31) Japanese (0.22)

> YELP labels food (0.99)

Recommended restaurants

Tercommended restminants		
Name	Cuisine	Stars
Red Plate	Chinese	5.0
Bar Sake & Robata Grill	Japanese, Sushi Bars	5.0
Ichi Belle	Japanese, Asian Fusion	5.0
Bar Charlie	Japanese	5.0
Tetsuro's Sayonara, Aloha, Going Away Uye At Japanese Curry n	Japanese	5.0



Understanding advertisements

Savchenko, Alexeev, Nikolenko et al, COLING 2020



Automatic Understanding of Image Advertisements

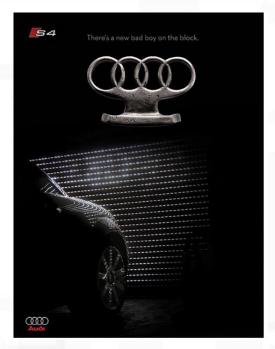


WWF anti-deforestation ad

Topic: environment.

Sentiment: alarmed.

Symbols: environment.



Audi ad: a new bad boy on the block

Topic: cars

Sentiment: inspired

Symbols: n/a.

http://people.cs.pitt.edu/~kovashka/ads/



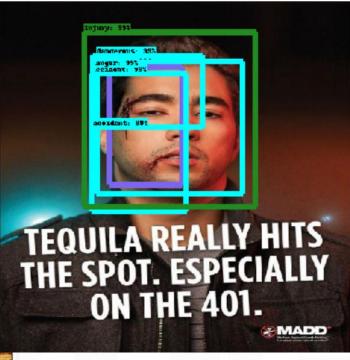
Symbolism in Image Advertisements



spicy alcohol destruction hot spicy spicy hot spicy

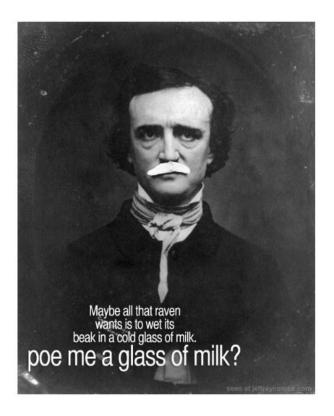


danger accident dangerous accident injury harm accident





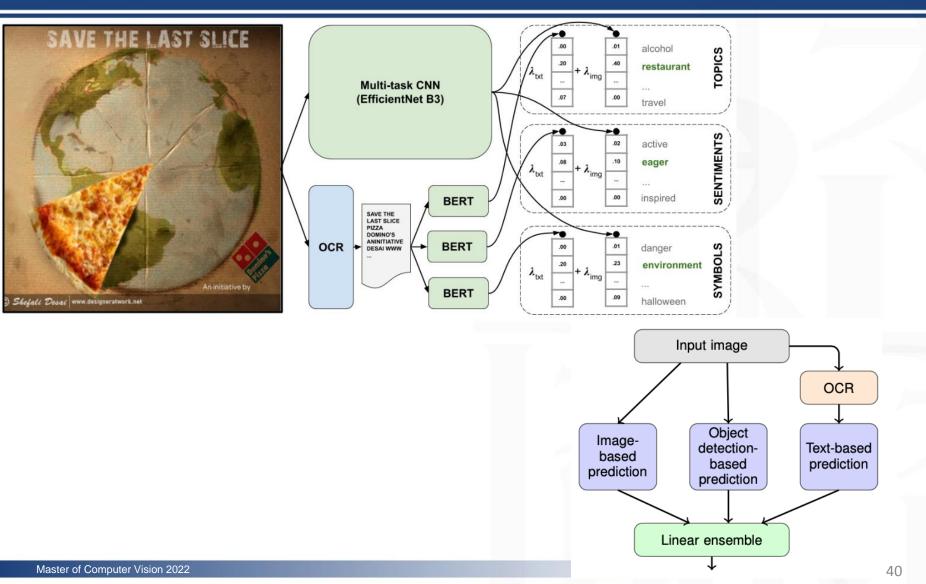
OCR (optical character recognition)



Tesseract	Maybe all that raven wants is to wet its beak in a'cold glass
Smith (2007)	of milk.
	poe mea lo EISsTo) aU eg
EAST+Tesseract	Maybe are Gein) eu SH Hostel S to wet its ake cold Me TS
Kopeykina and	10a milk. eee me glass 0) aa Penal see
Savchenko (2019)	
PSENet	Elle that eV WM eMey iy is 10 Wed Mey in beak — Ey
Wang et al. (2019)	milk. of (eek glass — ranlll@a a me al eee) glass be at
	jeffreycombs-com
EasyOCR	Maybe all that raven poe me a glass ofmik? beak in a cold
JaidedAI (2020)	glass ofmik. wants is to wet its seen 3t jefieycomlzesolm
Charnet	MAYBE ALL THAT RAVEN WANTS WET ITS BEAKIN
Xing et al. (2019)	COLD GLASS MILK POE GLASS MILK? SEEN ATJ
	COM
CloudVision	Maybe all that raven wants is to wet its beak in a cold glass
Otani et al. (2018)	of milk. poe me a glass of milk? seen at jeffreycombs.com



Pipeline





Thank you!