# Supplementary Material: Removing Bias in Multi-modal Classifiers: Regularization by Maximizing Functional Entropies 

Itai Gat*<br>Technion<br>Idan Schwartz<br>Technion<br>Alexander Schwing<br>UIUC<br>Tamir Hazan<br>Technion

In this supplementary material we study additional quantitative (see Sec. 1p) and qualitative (see Sec. 2) results. We detail our experimental settings (see Sec. 3). Finally, we provide the code we use to perform the experiments (see Sec. 4).

## 1 Quantitative Evaluation

To study robustness, in Tab. S1, we provide analysis for an ensemble of different models on VQACPv 2 . For each model, we used different values of $\lambda$, and different regularization, i.e., Eq. (17) and Eq. (18). Using our Top-3 ensemble, we improved the accuracy by $\sim 2 \%$ compared to our single model ( $56.74 \%$ vs. $54.55 \%$ ), and $\sim 4 \%$ compared to the previous single model baseline ( $56.74 \%$ vs. 52.05\%).

In Tab. S2, we show a per question type comparison on VQA-CPv2. Our method improved significantly the "how many people are" question type that requires counting ( $19.98 \% \mathrm{vs} .42 .81 \%$ ). However, for the "is there" question type which requires image detection, our method's performance dropped in a $19.6 \%$ accuracy ( $22.4 \%$ vs. $42 \%$ ).

Table S1: A comparison of different ensemble models on VQA-CPv2. We underline the best single-model and bold the best ensemble model.

| Model | Overall | Answer type |  |  |
| :--- | :---: | :---: | :---: | :---: |
|  |  | Yes/No | Number | Other |
| Eq.(17) | 54.01 | 73.02 | 43.15 | $\underline{47.02}$ |
| Eq.(18) | $\underline{54.55}$ | $\underline{74.03}$ | $\underline{49.16}$ | 45.82 |
| Top-3 | $\mathbf{5 6 . 7 4}$ | 77.70 | 48.56 | $\mathbf{4 8 . 0 1}$ |
| Top-4 | 56.41 | 77.34 | 48.05 | 47.74 |
| Top-5 | 56.54 | $\mathbf{7 7 . 4 8}$ | $\mathbf{4 8 . 7 2}$ | 47.71 |

Table S2: Comparison between VQA-CPv2 models in the paper and an ensemble of three models. We show the differences between the state-of-the-art [1] and our method in parenthesis.

| Question-type | LMH [1] | Eq.(18) | Eq.(17) | Ensemble |
| :--- | :---: | :---: | :---: | :---: |
| are there | 35.4 | $28.89(-6.51)$ | $32.22(-3.18)$ | $32.86(-2.54)$ |
| what brand | 26.75 | $27.35(0.6)$ | $21.84(-4.91)$ | $22.76(-3.99)$ |
| what room is | 92.57 | $93.22(0.65)$ | $92.66(0.09)$ | $93.95(1.38)$ |
| what color is | 77.18 | $75.53(-1.65)$ | $78.03(0.85)$ | $78.35(1.17)$ |
| is | 77.21 | $69.78(-7.43)$ | $78.74(1.53)$ | $75.13(-2.08)$ |
| are they | 66.78 | $62.0(-4.78)$ | $61.67(-5.11)$ | $67.33(0.55)$ |

[^0]| what number is | 3.28 | 4.22 (0.94) | 3.07 (-0.21) | 2.88 (-0.4) |
| :---: | :---: | :---: | :---: | :---: |
| what does the | 17.67 | 17.54 (-0.13) | 17.49 (-0.18) | 18.33 (0.66) |
| is this person | 73.15 | 68.21 (-4.94) | 74.89 (1.74) | 71.87 (-1.28) |
| is the | 36.83 | 40.32 (3.49) | 36.61 (-0.22) | 37.91 (1.08) |
| what is the man | 56.14 | 58.22 (2.08) | 57.21 (1.07) | 58.53 (2.39) |
| what kind of | 42.83 | 45.7 (2.87) | 44.26 (1.43) | 45.75 (2.92) |
| does this | 68.91 | 60.54 (-8.37) | 65.7 (-3.21) | 62.23 (-6.68) |
| is there a | 40.38 | 46.92 (6.54) | 49.62 (9.24) | 52.31 (11.93) |
| is he | 78.14 | 77.3 (-0.84) | 83.11 (4.97) | 78.97 (0.83) |
| what | 34.95 | 36.75 (1.8) | 35.03 (0.08) | 36.19 (1.24) |
| does the | 72.75 | 83.36 (10.61) | 75.84 (3.09) | 81.49 (8.74) |
| is the person | 72.8 | 78.0 (5.2) | 68.96 (-3.84) | 79.24 (6.44) |
| where is the | 26.15 | 28.07 (1.92) | 24.47 (-1.68) | 28.55 (2.4) |
| what animal is | 32.81 | 39.45 (6.64) | 36.83 (4.02) | 34.37 (1.56) |
| how | 17.38 | 19.79 (2.41) | 17.69 (0.31) | 18.82 (1.44) |
| what is the woman | 45.83 | 46.62 (0.79) | 47.0 (1.17) | 47.55 (1.72) |
| what is this | 59.05 | 60.47 (1.42) | 58.51 (-0.54) | 61.55 (2.5) |
| which | 31.68 | 32.79 (1.11) | 30.82 (-0.86) | 32.34 (0.66) |
| where are the | 35.22 | 33.85 (-1.37) | 32.83 (-2.39) | 36.11 (0.89) |
| are the | 44.31 | 42.94 (-1.37) | 44.31 (0.0) | 46.15 (1.84) |
| how many people are | 19.98 | 42.81 (22.83) | 35.98 (16.0) | 37.08 (17.1) |
| what is on the | 35.28 | 36.62 (1.34) | 36.82 (1.54) | 38.37 (3.09) |
| has | 68.72 | 73.53 (4.81) | 77.65 (8.93) | 81.7 (12.98) |
| was | 61.11 | 61.67 (0.56) | 60.28 (-0.83) | 63.89 (2.78) |
| what type of | 47.71 | 49.6 (1.89) | 48.14 (0.43) | 50.38 (2.67) |
| is this an | 58.79 | 48.83 (-9.96) | 53.9 (-4.89) | 46.67 (-12.12) |
| do | 37.75 | 39.25 (1.5) | 42.0 (4.25) | 39.5 (1.75) |
| can you | 64.0 | 81.84 (17.84) | 70.13 (6.13) | 80.9 (16.9) |
| who is | 25.08 | 25.19 (0.11) | 25.49 (0.41) | 25.17 (0.09) |
| are these | 48.59 | 51.77 (3.18) | 50.91 (2.32) | 51.95 (3.36) |
| do you | 68.97 | 82.41 (13.44) | 71.36 (2.39) | 81.88 (12.91) |
| what time | 20.36 | 24.13 (3.77) | 20.04 (-0.32) | 23.25 (2.89) |
| is the woman | 73.26 | 76.16 (2.9) | 84.11 (10.85) | 80.82 (7.56) |
| is this a | 85.34 | 90.17 (4.83) | 87.38 (2.04) | 92.25 (6.91) |
| what are the | 44.86 | 45.01 (0.15) | 45.27 (0.41) | 46.26 (1.4) |
| what color are the | 67.6 | 64.67 (-2.93) | 67.35 (-0.25) | 68.83 (1.23) |
| why | 18.0 | 17.98 (-0.02) | 18.37 (0.37) | 18.83 (0.83) |
| none of the above | 49.5 | 51.22 (1.72) | 50.64 (1.14) | 57.1 (7.6) |
| what is the person | 61.41 | 62.08 (0.67) | 61.52 (0.11) | 63.96 (2.55) |
| how many people are in | 48.16 | 52.78 (4.62) | 54.3 (6.14) | 58.36 (10.2) |
| is this | 81.59 | 84.46 (2.87) | 82.0 (0.41) | 88.46 (6.87) |
| why is the | 16.78 | 16.92 (0.14) | 17.7 (0.92) | 17.63 (0.85) |
| what is the color of the | 78.81 | 75.87 (-2.94) | 75.86 (-2.95) | 80.08 (1.27) |
| what is | 35.54 | 37.64 (2.1) | 36.58 (1.04) | 39.03 (3.49) |
| what are | 54.15 | 56.32 (2.17) | 54.03 (-0.12) | 55.57 (1.42) |
| is that a | 58.76 | 54.87 (-3.89) | 55.75 (-3.01) | 58.85 (0.09) |
| what is in the | 36.98 | 38.77 (1.79) | 39.35 (2.37) | 40.76 (3.78) |
| what sport is | 89.17 | 91.95 (2.78) | 91.02 (1.85) | 91.64 (2.47) |
| how many | 51.27 | 49.08 (-2.19) | 57.49 (6.22) | 56.32 (5.05) |
| what is the | 41.55 | 42.8 (1.25) | 41.18 (-0.37) | 44.11 (2.56) |
| is it | 68.75 | 71.49 (2.74) | 66.26 (-2.49) | 69.75 (1.0) |
| is the man | 53.54 | 50.3 (-3.24) | 52.13 (-1.41) | 52.56 (-0.98) |
| what is the name | 14.76 | 15.82 (1.06) | 14.37 (-0.39) | 15.95 (1.19) |
| is there | 42.0 | 22.4 (-19.6) | 37.2 (-4.8) | 36.0 (-6.0) |
| what color is the | 63.27 | 64.85 (1.58) | 64.29 (1.02) | 67.89 (4.62) |
| what color | 70.19 | 70.5 (0.31) | 70.44 (0.25) | 73.08 (2.89) |
| are | 45.37 | 45.83 (0.46) | 59.44 (14.07) | 58.58 (13.21) |
| Overall | 52.05 | 54.01 (1.96) | 54.55 (2.5) | 56.74 (4.96) |

## 2 Qualitative Evaluation

In this section, we qualitatively study success and failure cases using the VQA-CPv2 dataset. In Fig. 2 of the main paper, we show the proportions of information in each modality. We note that our regularization permits higher utilization of the visual modality. At the end of the document, we present success (see Fig. S1,S16) and failure cases (see Fig. S17.S32.

## 3 Experimental Settings

In the following, for each experiment, we describe its settings.

### 3.1 Colored MNIST

Recent works introduce different variants to Colored MNIST [2, 3, 4]. The main difference is the way bias is incorporated into the MNIST dataset. For instance, in [2] the authors use $\sigma=0.1$ and evaluate their method on the gray-scale test set while in [4] the authors use $\sigma=0.02$ and evaluate on a uniformly-colored test set. We follow the experimental settings introduced in REPAIR [2].

Model: We use LeNet [5] to encode the shape, then we concatenate the color representation to the last linear layer of LeNet and pass it through another linear layer with a bias term for classification. We use cross-entropy loss.
Hyper-parameters considered: We consider $\lambda$ in the range of $1 \mathrm{e}-9$ to $1 \mathrm{e}-11$. We report our results with $\lambda=1 \mathrm{e}-10$. Note, the values of the regularization term (i.e., the functional Fisher information) are approximately 1 e 9 . Therefore, the $\lambda$ hyperparameter acts as a normalizer that scales down the bias term to the $\left[0,10^{2}\right]$ range.

Computing infrastructure used: We use a single RTX2080Ti GPU. The average runtime for each epoch is one minute. The model converges in 35 epochs.

### 3.2 VQA-CPv2

Visual question answering (VQA) requires to answer a given question-image pair. VQA-CPv2 [6] is a re-shuffle of VQAv2 [7] that alleviates dataset priors.

Model: We add our regularization to the current state-of-the-art [1]. The current state-of-the-art model is based on "Bottom-Up and Top-Down Attention for Image Captioning and Visual Question Answering" [8].
Hyper-parameters considered: We consider $\lambda$ in the range of $1 \mathrm{e}-10$ to $1 \mathrm{e}-20$. We report our results with $\lambda=4 \mathrm{e}-17$. Note, the values of the regularization term (i.e., the functional Fisher information) are approximately 1 e 15 . Therefore, the $\lambda$ hyperparameter acts as a normalizer that scales down the bias term to the $\left[0,10^{2}\right]$ range.

Computing infrastructure used: We use two RTX2080Ti GPUs. The average runtime for each epoch is 5 minutes. The model converges in 15 epochs.

### 3.3 SocialIQ

SocialIQ [9] input data is constructed from a tuple of a video, a question about the social situation in the video, and an answer to that question. Tiven the tuple, the task is to predict whether the answer is correct or not.
Model: The textual inputs are encoded with BERT [10]. The visual input is encoded using a VGG16 [11]. From BERT we extract the last hidden layer's representation. From VGG16, we extract the layer before the final fully-connected representation of four frames from the video (picked uniformly over the video). Then, we forward all representations through a linear layer. Subsequently, the concatenation of the representations is passed through an MLP consisting of two linear layers for classification. We use binary cross-entropy loss. We use a ReLU [12] activation between all layers and dropout [13] with a dropout rate of 0.2.

Hyper-parameters considered: We consider $\lambda$ in the range of $1 \mathrm{e}-9$ to $1 \mathrm{e}-11$. We report our results with $\lambda=3 \mathrm{e}-10$. Note, the values of the regularization term (i.e., the functional Fisher information) are approximately 1 e 9 . Therefore, the $\lambda$ hyperparameter acts as a normalizer that scales down the bias term to the $\left[0,10^{2}\right]$ range.

Computing infrastructure used: For the baseline, we use a single RTX2080Ti GPU, the average runtime for each epoch is one minute. The model converges after 20 epochs.

### 3.4 Dogs and Cats

In this task, we classify whether a given image shows a dog or a cat. This task was first presented as a Kaggle competition ${ }^{2}$. The authors of "learning not to learn" [4] introduce a modified version. They collected one split to hold bright dogs and dark cats (TB1) and another split that contains the opposite, i.e., dark dogs and bright cats.
Model: We follow the setting of Kim et al. [4]: The authors use a ResNet-18 [14] that was pre-trained on ImageNet [15]. We fine-tuning the last fully-connected layer.

Hyper-parameters considered: In this dataset, we consider $\lambda$ in the range of $1 \mathrm{e}-10$ to $1 \mathrm{e}-14$. We report our results with $\lambda=2 \mathrm{e}-10$. Note, the values of the regularization term (i.e., the functional Fisher information) are approximately 1 e 9 . Therefore, the $\lambda$ hyperparameter acts as a normalizer that scales down the bias term to the $\left[0,10^{2}\right]$ range.
Computing infrastructure used: For the baseline, we use five RTX2080Ti GPUs. The average runtime for each epoch is 3 minutes. The model converges after 5 epochs.

## 4 Code

We use the PyTorch [16] framework to conduct all of our experiments. We also provide a Python package which we wrote to calculate the regularization term. Further explanations of how to integrate our regularization to any multi-modal problem are in the README file.

[^1]

Question type: what is the man


Question: Is anyone watching the baseball game?

GT: no, Ours: yes
 playing with?
GT: frisbee, Ours: frisbee

Question type: is


GT: no, Ours: no


Question: What is the man flying in the air?
GT: nothing, Ours: nothing


GT: no, Ours: no

Figuze S1


Question type: what number is


GT: 12, Ours: 12


GT: no number, Ours: 0

Question type: what sport is


GT: frisbee, Ours: frisbee


GT: baseball, Ours: baseball

Question type: is this person

appropriate protective gear for skateboarding
GT: no, Ours: no

GT: no, Ours: no


GT: no, Ours: no


Question type: what is the name


GT: horse, Ours: horse


Question: What is the name
of the clothing type she is
wearing?
GT: dress, Ours: dress

Question type: how many

on the grass?
GT: 5, Ours: 5


GT: no, Ours: no

here?
GT: 1, Ours: 1

Question type: does this
 be in the United States?

GT: no, Ours: no


Question: What is the name of the piece of furniture the stuffed animal is sitting on? GT: chair, Ours: chair


GT: 1, Ours: 1


GT: no, Ours: no

Question type: is there a


Question: Is there a man
or woman on the train?
GT: man, Ours: man


GT: dog, Ours: dog

on this beach?
GT: yes, Ours: yes


Question: What room of a
house would you find all of these items?
GT: kitchen, Ours: kitchen


Question: Is there a male
or female on the right side?
GT: male, Ours: male

Question type: is that a


GT: electric, Ours: electric

Question type: can you


GT: yes, Ours: yes

Question type: what


GT: winter, Ours: winter


GT: fork, Ours: fork
 at this intersection?
GT: yes, Ours: yes


Question: What structure is in the background?

GT: fence, Ours: fence

Question type: does the


GT: yes, Ours: yes


Question: Does the fruit appear to be ripe?
GT: yes, Ours: yes

Question type: is the person


GT: yes, Ours: yes

ra is beautiful?
GT: yes, Ours: yes


GT: yes, Ours: yes

Question type: do you


GT: yes, Ours: yes

Question type: where is the


GT: beach, Ours: beach


GT: wall, Ours: on wall


GT: kitchen, Ours: kitchen


GT: human, Ours: human


Question: How will the man get back to shore?

GT: boat, Ours: boat


Question: Is this a double decker bus?

GT: yes, Ours: yes


GT: human, Ours: human

Question type: how


GT: bright, Ours: bright

Question type: is this a


Question: Is this a bar?

GT: yes, Ours: yes

Question type: none of the above


Question: Did the wave get larger after this image was taken?

GT: yes, Ours: yes


Question: Would you consider the cyclist in yellow an individual?

GT: yes, Ours: yes


GT: yes, Ours: yes


Question type: where are the


GT: street, Ours: street


GT: field, Ours: field

Question type: are the


Question type: how many people are


Question: How many people are standing up?

GT: 2, Ours: 2


Question: How many people are wearing glasses?

GT: 2, Ours: 2


Question: How many people are on the bike?

GT: 2, Ours: 2


GT: snow, Ours: snow


Question type: has


GT: yes, Ours: yes


Question: Has the food been

GT: yes, Ours: yes

Question type: was


GT: summer, Ours: summer


GT: home, Ours: home

Question type: what type of


GT: wooden, Ours: wooden


Question: What type of building is the couple standing in front

GT: church, Ours: church


GT: living room, Ours: living room

## Question type: do



Question type: what is the person


Question: What is the person squatting called?

GT: catcher, Ours: catcher


Question: What is the person skiing on?

GT: snow, Ours: snow

Question type: who is


GT: cat, Ours: cat

Question type: are these


GT: alive, Ours: alive

of this car?
GT: ford, Ours: ford


GT: chopsticks, Ours: chopsticks
T:


GT: umbrella, Ours: umbrella


GT: women, Ours: women

Question type: what is the woman


GT: skateboard, Ours: skateboard


GT: horse racing, Ours: horse racing

Question type: what are


GT: elephant, Ours: elephant


GT: motorcycle, Ours: motorcycles

Question type: what time


GT: night, Ours: night


GT: fall, Ours: fall

Question type: is the woman


GT: no, Ours: no


GT: no, Ours: no


GT: uphill, Ours: uphill
 position behind the batter?

GT: catcher, Ours: catcher

Question type: what are the


Question: What are the colors
of the chairs?
GT: brown, Ours: brown


Question: What are the military men cutting?

GT: cake, Ours: cake


Question: What is the gir
wearing on her head?
GT: hat, Ours: hat


GT: standing, Ours: standing

Question type: what color are the


GT: white, Ours: white


Question: Is he wearing a helmet?

GT: no, Ours: no
 man's pants?

GT: white, Ours: white

Question type: is he


GT: no, Ours: no


GT: no, Ours: no
 slightly squatting?

T: skateboarding, Ours: skateboarding


GT: resting, Ours: resting


GT: drinking, Ours: drinking

Question type: what is this


Question type: how many people are in


GT: 1, Ours: 1

in the photo?
GT: 3, Ours: 3

Question type: what color


GT: white, Ours: white


GT: black, Ours: black

photo?
GT: 1, Ours: 1


GT: white, Ours: white



Question: Is this picture in full color?

GT: yes, Ours: yes

Question type: why is the


GT: cutting cake, Ours: cutting cake


GT: shade, Ours: shade


GT: safety, Ours: safety

Question type: what is the color of the


GT: yellow, Ours: yellow


Question: What is zoomed in
on in the picture?
GT: keyboard, Ours: keyboard


Question: What is the color of the woman's shirt?

GT: white, Ours: white

Question type: what is


Question: What is keeping the vehicle from falling?

GT: kickstand, Ours: kickstand


GT: snow, Ours: snow


GT: no, Ours: no
 structure?
GT: no, Ours: yes

Question type: what is in the


GT: dog, Ours: dog


Question: What is in the

GT: fruit, Ours: fruit

Question type: what does the


GT: ski poles, Ours: ski poles

have on her head?

GT: hair, Ours: hair

Question type: what kind of


GT: wood, Ours: wood


GT: chicken, Ours: chicken


GT: bananas, Ours: bananas

Question type: is it

fall?
GT: winter, Ours: winter

night?
GT: day, Ours: day

Question type: is the man


GT: snowboarding, Ours: snowboarding
 or catching the frisbee?

GT: catching, Ours: catching

Question type: is there


GT: water, Ours: water

or grass?
GT: grass, Ours: grass


GT: shorts, Ours: shorts
 cola on the door of the refrigerator?
GT: beer, Ours: beer

Question type: what color is the


GT: blue, Ours: blue


Question: WHAT COLOR is the
sky?
GT: blue, Ours: blue


Question: What color is the

GT: blue, Ours: blue


GT: yellow, Ours: green

automobiles?
GT: no, Ours: no
 glasses or wine glasses?

GT: wine, Ours: wine


GT: white, Ours: white

Question type: are


GT: yes, Ours: no

Question type: are there


GT: broccoli, Ours: broccoli


GT: white, Ours: white


GT: yes, Ours: no


Figure S16

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 motorcycle?

GT: ducati, Ours: yamaha

Question type: what room is


GT: office, Ours: dining room


GT: store, Ours: living


Question: What brand of computer
is the partially closed one?
GT: apple, Ours: dell


GT: office, Ours: living room

Question type: what is the man


GT: skis, Ours: ski poles


Question: What is the man doing to the man in white?
GT: walking, Ours: catching

Question type: is

GT: no, Ours: yes



GT: no, Ours: yes



Question: Is today the dogs birthday?

GT: no, Ours: birthday


Question type: what number is


$$
\text { GT: 9, Ours: } 38
$$



Question: What number is on
the back of the red shirt?
GT: 28, Ours: 0

Question type: what sport is


GT: football, Ours: frisbee

Question type: is this person


GT: no, Ours: yes


Question: Is this person and athlete?

GT: no, Ours: yes


Question: Is this person waterskiing?

GT: no, Ours: yes


Question type: what is the name


Question type: how many


GT: 1, Ours: 2

Question type: does this
 sport?
GT: no, Ours: yes


GT: no, Ours: yes

 a bathroom?

GT: no, Ours: yes

Question type: is there a
 street or a dirt trail in this photo?
GT: paved, Ours: road


Question: Is there a painting
or a mirror in the frame?
GT: mirror, Ours: picture


GT: international, Ours: no

Question type: is that a


Question: Is that a limb or stick?

GT: limb, Ours: rope


GT: yes, Ours: no


Question: What shape is the
red sign?
GT: star, Ours: octagon


GT: india, Ours: usa


GT: golden retriever, Ours: terrier

Question type: does the
 the color of his pants?

GT: yes, Ours: no

Question type: is the person


Question: Is the person in the image a kid or a midget?
GT: kid, Ours: child


Question: Is the person using
this kitchen organized?

GT: yes, Ours: no

Question type: do you


Question: Do you see a serving knife?

GT: yes, Ours: no

ice cream?
GT: yes, Ours: no

Question type: where is the


GT: parked, Ours: on road


Question: Where is the train?

GT: station, Ours: on tracks


Question: Where is the man Question: Where is the man
and woman riding an elephant?

GT: forest, Ours: outside


## Question type: how



GT: hour, Ours: 1 hour
GT: 25, Ours: fast

Question type: is this a

GT: yes, Ours: no



Wii controller?
GT: yes, Ours: wii
Ques



Question: How old is the elephant?

GT: old, Ours: young

Question type: none of the above


GT: yes, Ours: snow


GT: stomach, Ours: head


GT: yes, Ours: no


Question type: where are the


GT: zoo, Ours: outside


GT: field, Ours: park


GT: street, Ours: parking lot

Question type: are the


GT: dirty, Ours: clean


Question type: how many people are


GT: 2, Ours: 4


GT: 2, Ours: 3


GT: 7, Ours: 4

Question type: what is on the
 suitcase?
GT: jacket, Ours: clothes


GT: yes, Ours: no


GT: toasted, Ours: fried


GT: sticker, Ours: clock

Question type: has


Question: Has the wine been opened?

GT: yes, Ours: no

Question type: was


Question type: what type of


GT: rain, Ours: rainy


GT: selfie, Ours: color


GT: dog, Ours: brown


Question type: what is the person


GT: atv, Ours: motorcycle


GT: skateboarder, Ours: girl


GT: shorts, Ours: wetsuit

Question type: who is


GT: people, Ours: surfer

Question type: are these


GT: street, Ours: ground


GT: no one, Ours: elephant


GT: vegetables, Ours: fruit

Question type: what is the woman


GT: brushing hair, Ours: eating


GT: masks, Ours: sunglasses

Question type: what are


GT: nothing, Ours: sunglasses


GT: brushing teeth, Ours: cutting

doing?
GT: listening to music, Ours: talking


GT: looking down, Ours: eating

Question type: what time


GT: 1:55, Ours: 3:00

Question type: is the woman


GT: no, Ours: yes


GT: no, Ours: yes


GT: no, Ours: yes


GT: bunny, Ours: baby

Question type: what are the


GT: signs, Ours: nothing

Question type: what color are the


GT: white, Ours: gray


Question: Is he a beginner surfer?

GT: no, Ours: yes

letters on the board?
GT: white, Ours: silver

Question type: is he


Question: Is he bunting or
swinging?
GT: bunting, Ours: baseball
 players' shoes?

GT: white, Ours: blue


Question: Is he right handed?

GT: no, Ours: yes


Question type: what is this


GT: farmer, Ours: police


GT: stove, Ours: kitchen


GT: food, Ours: bus

Question type: how many people are in

in the picture?
GT: 10, Ours: 15

Question type: what color


GT: white, Ours: pink


Question: What color are this person's sunglasses?

GT: white, Ours: black


Question: What color stands out
in this picture?
GT: red, Ours: pink

Question type: is this


GT: yes, Ours: no


Question: Is this area desert-like or lush and green?

GT: lush and green, Ours: trees


GT: yes, Ours: no

Question type: why is the


Question: Why is the picture
funny?
GT: no, Ours: yes
 face blurred?

GT: its moving, Ours: tired

Question type: what is the color of the

of the vehicles?
GT: green and yellow, Ours: yellow


Question: What is clear in the background?

GT: window, Ours: train

Question type: what is
 of the catcher's hat?
GT: red, Ours: blue


Question: What is below the mirror?

GT: fireplace, Ours: tv


GT: white, Ours: pink


GT: trash can, Ours: desk


GT: no, Ours: yes

Question type: what is in the


GT: bottle, Ours: vase


GT: nothing, Ours: flowers

Question type: what does the


GT: melrose, Ours: clock
of the vehicle hold?

Question type: what kind of


GT: jet ski, Ours: boat


GT: horse and cow, Ours: cows


GT: no, Ours: yes


Question: What is in the

GT: oil, Ours: beer


GT: slow, Ours: nothing


GT: khaki, Ours: jeans

Question type: is it

cold where this photo was taken?
GT: warm, Ours: cold


Question: Is it cold or
warm?
GT: warm, Ours: cold

Question type: is the man


GT: woman, Ours: man


Question: Is the man standing straight or leaning on the surfboard?

GT: leaning, Ours: neither

Question type: is there


Question: Is there an even number of donuts or an odd
number?

GT: odd, Ours: 0

Question type: what color is the


GT: red, Ours: white


GT: gray, Ours: green


Question: What color is the
GT: gray, Ours: brown


GT: red, Ours: blonde


GT: yes, Ours: no



GT: red, Ours: brown

Question type: are


GT: no, Ours: coffee

## Question type: are there



Figure S32


GT: red and black, Ours: red


GT: yes, Ours: unknown



[^0]:    *itaigat@technion.ac.il

[^1]:    $2^{2}$ https://www.kaggle.com/c/dogs-vs-cats

