

MIT-IBM Effective Human-AI Teams via Learned Natural Language Rules and Onboarding

NeurIPS 2023 Spotlight

arxiv.org/pdf/2311.01007.pdf

github.com/clinicalml/onboarding human ai



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Humans Interacting with AI

Setting: you are learning to code, so you ask an AI for help

- Should you trust the Al's code?
- Try to solve it on your own?
- Edit the AI code to make it work?

You

Please solve this coding problem in Python: given two integers a and b, return the sum of the two integers without using the operators + and -., write concise code without examples

ChatGPT

You can solve this problem concisely using Python by employing bit manipulation. Here's a concise implementation:

python	🖞 Copy code
<pre>def get_sum(a, b): mask = 0xFFFFFFFF # A mask to limit the result to 32 bits while b: a, b = (a ^ b) & mask, ((a & b) << 1) & mask return a if a <= 0x7FFFFFFF else ~(a ^ mask)</pre>	
# Example usage:	
# a = 3	
# b = 2	
<pre># result = get_sum(a, b)</pre>	
<pre># print(result)</pre>	

This code uses a loop to simulate the addition process without using the `+` or `-` operators. It ensures the result is a 32-bit integer by applying a mask and handles negative results correctly. Uncomment the example usage to test the function with specific values.

Humans Interacting with AI

Setting: you are driving but it's low visibility. Al dashboard is helping you see.

- Do you stop since AI is telling you it's a red light?
- Pull over until visibility improves?

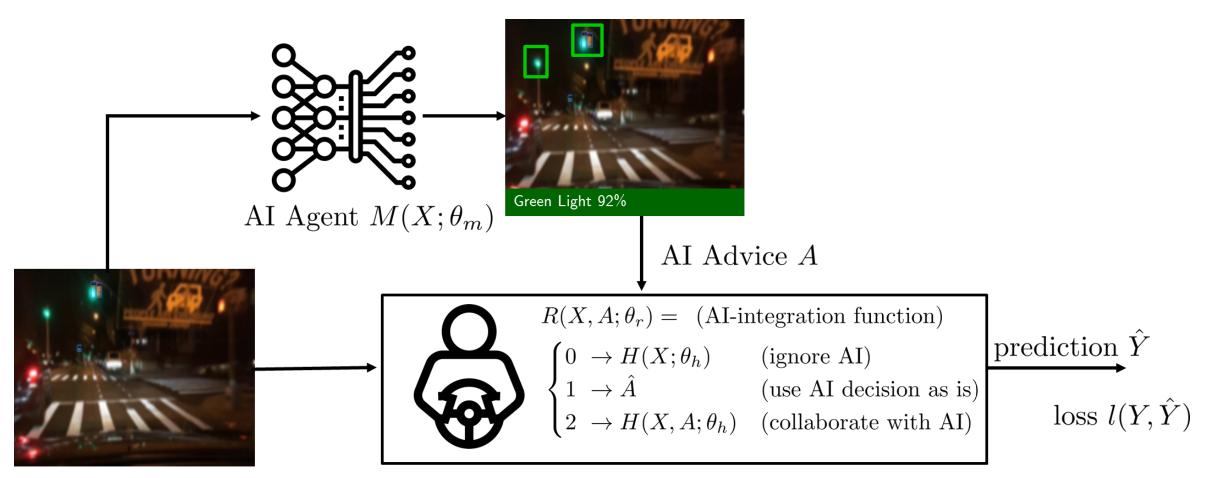


Image modified from https://www.popsci.com/technology/new-audi-tech-provides-traffic-light-updates/

Central Problem:

- How do you know when to:
- rely on the AI,
- collaborate with the AI or
- ignore its suggestion?

AI-Assisted Decision Making



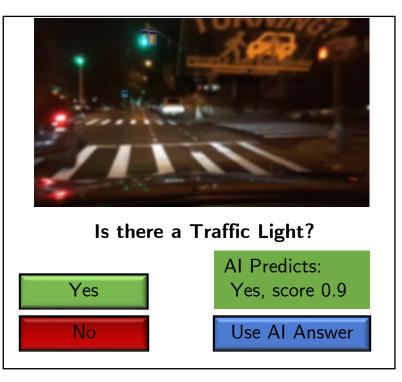
Human $H(X, A; \theta_h)$

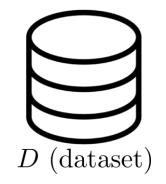
Input Context XGround Truth Y

Proposed IntegrAl Approach:

- Learn from data the optimal Alintegration function
- Teach the human the optimal way to collaborate

Step 1: Human Data Collection Understand Human performance and prior reliance on AI



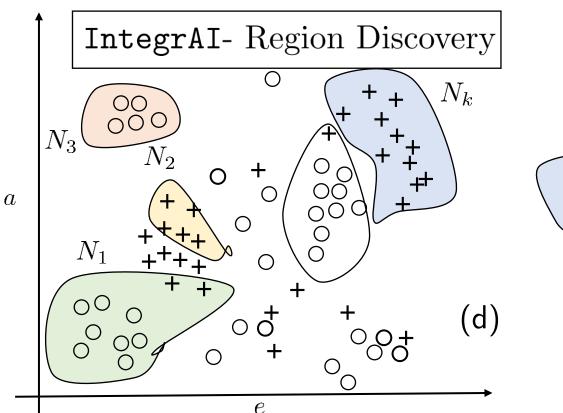


- x_i (task), y_i (answer)
- a_i (AI), h_i (human)
- t_i (text description), e_i (embedding)
- r_i (used AI)

Assume the human has a prior integration function R_{0} , our goal is:

- Understand human performance without AI H(X)
- Understand the human prior R₀

Step 2: Discover Regions Find regions of data where human collaborates incorrectly with AI



Goal of teaching is to correct the prior: e.g. if human trusted the AI, in a region where AI is incorrect, find that region and inform the human!

Region is denoted by a **center** c and a **radius** gamma

Step 2: Discover Regions Find regions of data where human collaborates incorrectly with AI

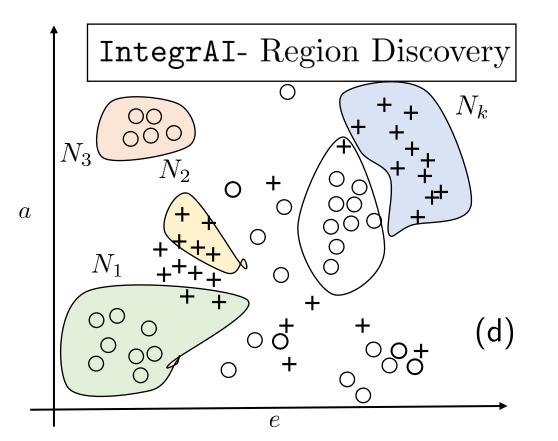
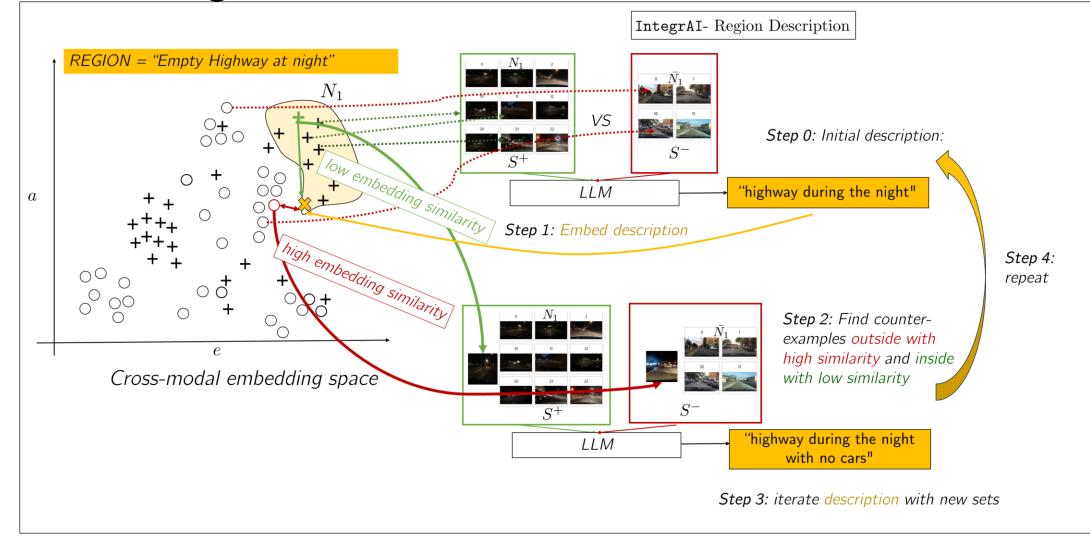


Table 3: Error (\downarrow) on the test set (in %) of the human-AI system when following integrators resulting from different region discovery methods with 10 regions on the different non-synthetic datasets.

	BDD	MMLU	DynaSent	MS-COCO
IntegrAI (ours)	$\textbf{17.8} \pm \textbf{0.2}$	$\textbf{45.3} \pm \textbf{0.3}$	20.2 ± 0.3	$\textbf{22.6} \pm \textbf{0.4}$
DOMINO [EVS+21]	18.9 ± 0.4	48.1 ± 0.2	20.0 ± 0.2	22.7 ± 0.4
K-MEANS [RLC ⁺ 22]	19.0 ± 0.5	$\textbf{45.3} \pm \textbf{0.3}$	20.0 ± 0.2	23.2 ± 0.1
DoubleGreedy [MSS22]	18.9 ± 0.1	46.1 ± 0.6	20.0 ± 0.2	23.8 ± 0.4

Step 3: Describe Regions Describe regions in text with an LLM

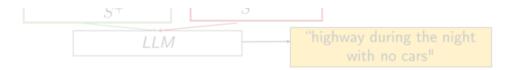


Step 3: Describe Regions Describe regions in text with an LLM



Table 5: Evaluation of our region description algorithm (Algorithm 2) on selected subsets of MS-COCO where the different algorithms try to describe a set of images that all contain a given object. For example, a region may be defined by images containing the object "apple". Then we compare the descriptions resulting from the different algorithms to the description "apple".

	best-caption	SEAL	IntegrAI $(S^- = \emptyset)$	IntegrAI $(m=0)$	IntegrAI $(m=5)$	IntegrAI $(m=10)$
METEOR	$ 12.9 \pm 1.9$	9.16 ± 1.89	24.3 ± 3.3	25.4 ± 3.2	$\bf 26.1 \pm 3.3$	25.4 ± 3.3
sent-sim	39.8 ± 1.9	44.1 ± 2.5	65.1 ± 3.2	67.0 ± 3.1	66.0 ± 3.2	68.0 ± 3.3
ROUGE	5.81 ± 1.2	0.0 ± 0.0	25.6 ± 4.9	32.6 ± 5.4	27.9 ± 5.1	35.6 ± 5.5
SPICE	12.7 ± 1.9	7.53 ± 2.3	41.1 ± 5.8	43.8 ± 5.8	45.2 ± 5.8	45.2 ± 5.8



Step 3: iterate description with new sets

Step 4: Teach Regions to Human In onboarding stage, teach human each region

Region 1:

Ignore AI when it's a "highway during the night" **Region 2:**
 1
 2

 1
 2

 1
 12

 1
 12

 1
 12

 1
 12

 1
 12

 1
 12

 1
 12

 1
 12

 1
 12

 1
 12

 1
 12

Use AI when it's a "city during the day"



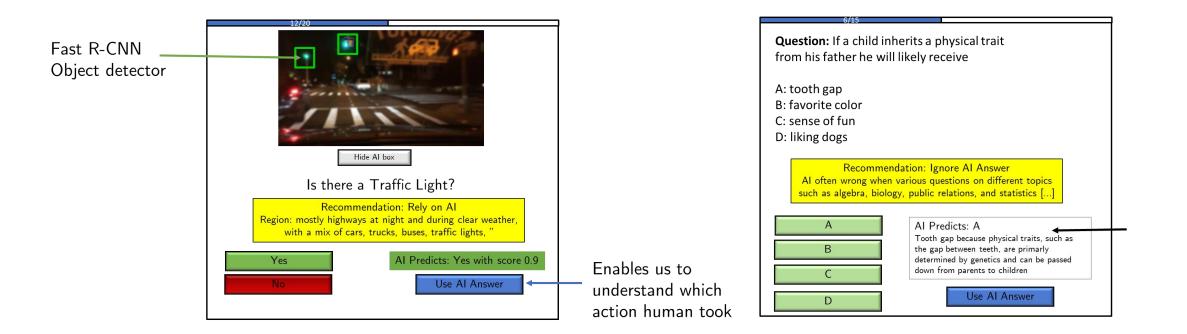
. . .

Region k:

Teaching Phase: for each region

- **Step 1:** human predicts on example in region
- **Step 2**: human receives feedback about their correctness and the Ais correctness
- **Step 3:** (from example to region) human is shown the region description and a gallery of examples from the region

(any example the human gets wrong are re-shown)



Task: Given a blurry image, human must figure out if the light is red or green (proxy for moving forward)

Task: Given a multiple-choice question, human must figure out the right answer.

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Accuracy: without onboarding 77.2, with onboarding 82.6 (significant increase)

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Accuracy: without onboarding 75.0, with onboarding 74.4, why?

ChatGPT explanations often inform human whether to use its answer: - 'Unfortunately, the options provided do not provide a clear answer to [...]

- 'The answer cannot be provided with the given information [...]