**INDIANA UNIVERSITY** 

## Abstract

We explore methods for improving the accuracy of medical image decisions by aggregating repeated decisions. Novices (undergraduates) and experts (medical professionals) made classification decisions (cancer vs. not cancer) and confidence judgments on cell images, viewing and classifying each image twice. We show that the maximum confidence slating algorithm, which uses the most confident response for an image as the final response, improves performance for novices and experts at the individual level. We then show that aggregation algorithms based on confidence weighting scale to larger groups of participants, with the performance of groups of novices reaching that of individual experts.



Figure 1. Panel A illustrates the structure of Experiment 1a. Participants first completed a brief training phase before the main experiment. In the main task, two responses were collected for each participant on each of the 300 images. In the first pass (shown in yellow), participants were presented with the image of a cell and were asked the question 'Is this a blast?' (i.e. Is this cancerous?). In the second pass (shown in orange), the same image was rotated and participants were asked the question 'Is this a non-blast?' (i.e. Is this non-cancerous?). Each set of main trials was preceded by practice trials. Experiment 1b was similar to Experiment 1a except that the same question was asked in both of the main blocks (i.e., 'Is this a blast?') and images were not rotated. Experiment 2 was a shorter version of Experiment 1a and was designed for expert participants. Panel B shows the two parts of every trial in the main task In the first part, participants decided whether a cell was a blast or not. In the second part, participants indicated their confidence on a scale of 50-100, where 50 was 'guessing' and 100 was 'certain correct'.

# The Wisdom of the Confident Crowd in Medical Image Decision-making

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# **Results: Combining two responses**

Table 1. The mean performance of each algorithm for Experiments 1a, 1b, and 2.

Exp. 1a	Exp. 1b	Exp. 2
Novice (Reframing)	Novice (No Reframing)	Expert (Reframing)
66.1%	66.5%	71.6%
67.4%*	67.4%*	73.7%*
66.7%	66.5%	72.4%
70.0%*	70.0%*	78.5%*
	Exp. 1a Novice (Reframing) 66.1% 67.4%* 66.7% 70.0%*	Exp. 1aExp. 1bNovice (Reframing)Novice (No Reframing)66.1%66.5%67.4%*67.4%*66.7%66.5%70.0%*70.0%*

*Note.* \*Significant improvement as compared to average response with p<0.0001

# **Results: Combining Multiple Responses**



Figure 2. Panels A, B and C plot the accuracy obtained by applying all of our Wisdom of the Crowd algorithms to data from Experiments 1a, 1b, and 2 respectively. The legend for Panel B, C and D is identical to the one is Panel A. We observe that as the responses are pooled from more participants, the accuracy increases. Panel D compares the best performing algorithm, Confidence Weighted Voting with Both Responses, for novice participants from Experiment 1a and expert participants from Experiment 2. We restrict the responses from Experiment 1a to the stimuli that were used in the main trials of Experiment 2 and apply the algorithm. We observe that the performance of the expert participants is greater than that of the novice participants. We also observe that 5 novice participants can match the performance of 1 expert.



# **Results: Comparison of the Algorithms**

Table 2. Results of the different Wisdom of the Crowd (WoC) algorithms when groups of 7 decision makers are considered.

Algorithm	Exp. 1a	Exp. 1b	Exp. 2
	Novice	Novice	Expert
	Reframing	No Reframing	Reframing
First Response (1 Person) <sup>a</sup>	66.5%	67.0%	73.0%
Simple Voting - First Resp.	77.6%	78.7%	86.2%
Simple Voting - Second Resp.	77.5%	77.5%	84.0%
Simple Voting - Both Resp.	80.7%	80.0%	86.9%
Conf. Weight Voting - First Resp.	78.3%	79.2%	86.7%
Conf. Weight Voting - First Resp Log Odds Weight	79.4%	80.4%	88.6%
Conf. Weight Voting - Both Resp.	81.8%	81.0%	87.9%
Conf. Weight Voting - Both Resp Log Odds Weight	81.3%	81.0%	88.3%
Conf. Slated Voting	80.2%	79.3%	85.5%
Conf. Slated Voting - Log Odds Weight	79.9%	79.8%	85.7%
<i>Note.</i> The best performing algorithm for each experiment is in bo <sup>7</sup> The average accuracy using only the first response for a group siz	old. 	ovided as a baselin	e comparison.

- same question again.

- performance of experts.

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### Conclusions

• Maximum confidence slating - within improves performance for Novices and Experts. • Maximum confidence slating - between is more effective than maximum confidence slating within. Using responses from two people is more useful than asking the same person the

 Aggregating decisions from a large number of individuals dramatically improves performance. Repeated decision making and confidence weighting improves performance at the group level. This is especially larger when the frame of the question is changed.

It is better to average confidence judgments than slating to the more confident decision. Aggregating decisions from a small number of novices (i.e. 5 novices) matches the

### Acknowledgments

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