

The 3rd International Conference on Artificial Intelligence in Education Technology (AIET 2022)

# **EXPLAINABILITY IN**

# **AUTOMATIC SHORT ANSWER GRADING**

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#### **AGENDA**



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Related Work	2
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# **INTRODUCTION**

#### **MOTIVATION: UN Sustainable Development Goal 4**







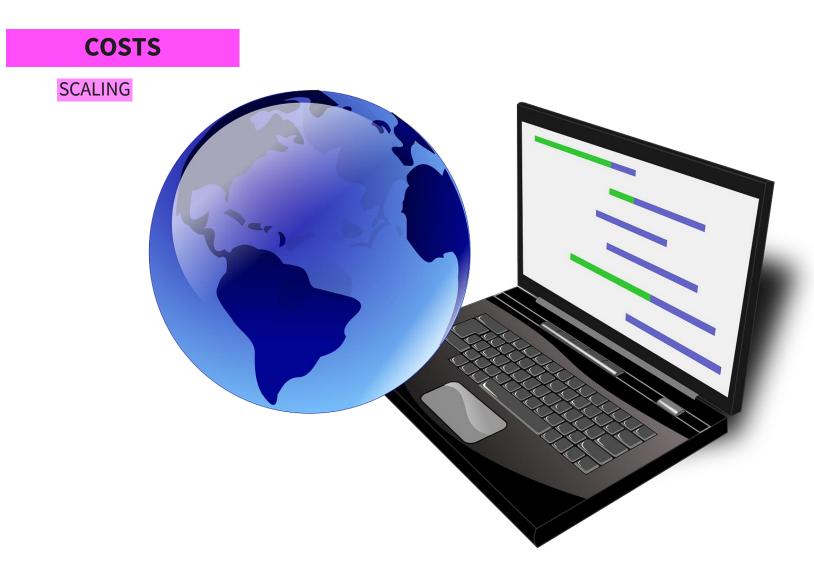


Image Sources: United Nations: Sustainable Development Goals: 17 Goals to Transform our World (2021); OpenClipart-Vectors/154119/Pixabay.

# **MOTIVATION: Challenges in Education**



**ACCELERATION** 



#### **AI IN EDUCATION: Potential**



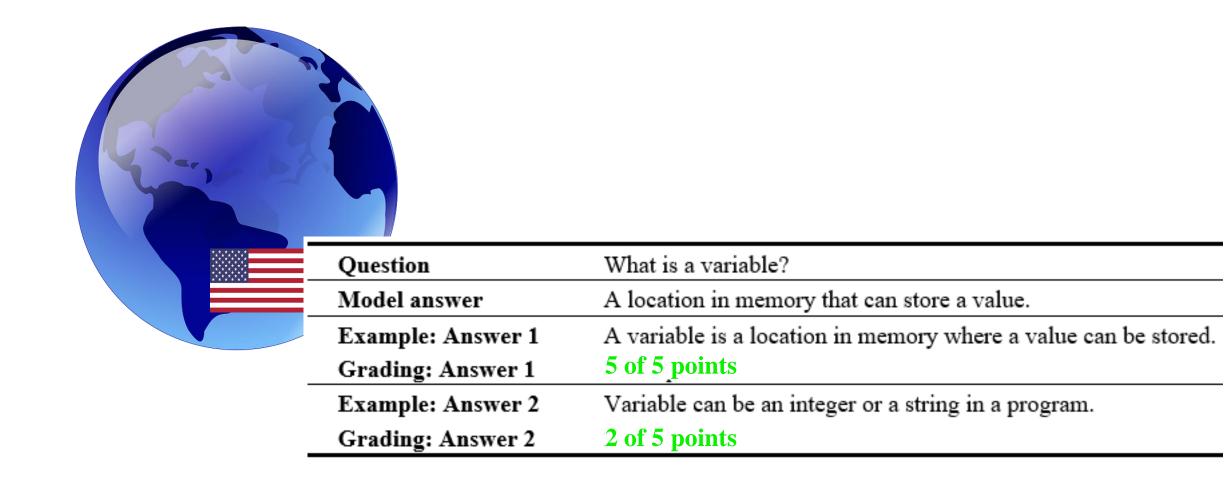
#### **AUTOMATION**

GRADING



#### **AUTO-GRADING**



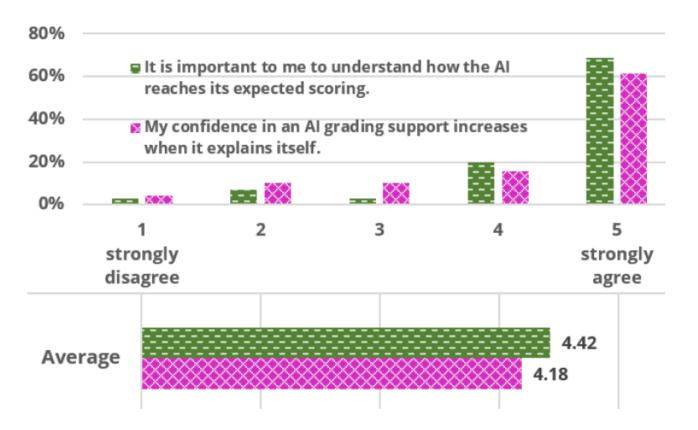


#### **AUTO-GRADING: Trust**



#### **STUDY**

**−70** professors, lecturers and teachers





# **RELATED WORK**

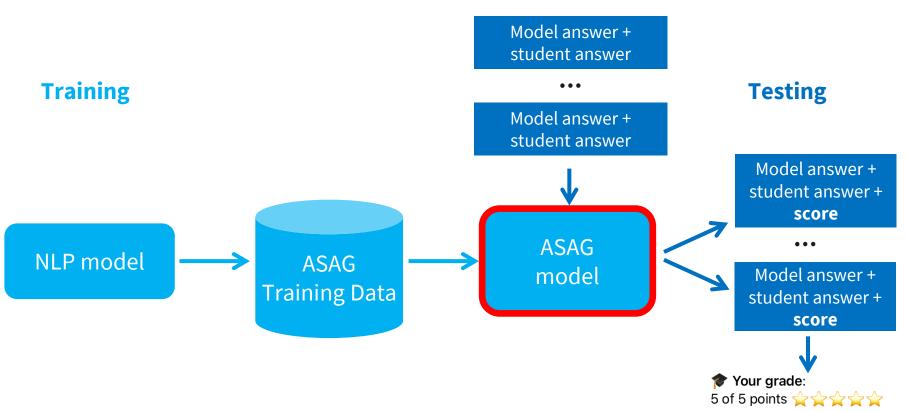
#### **RELATED WORK: AUTO-GRADING**



AUTOMATIC SHORT ANSWER GRADING

#### **Deep learning**

e.g., (Burrows et al., 2014; Camus & Filighera, 2020; Sawatzki et al., 2021; Schlippe & Sawatzki, 2021b)



Graphic Source: Custom Depiction.

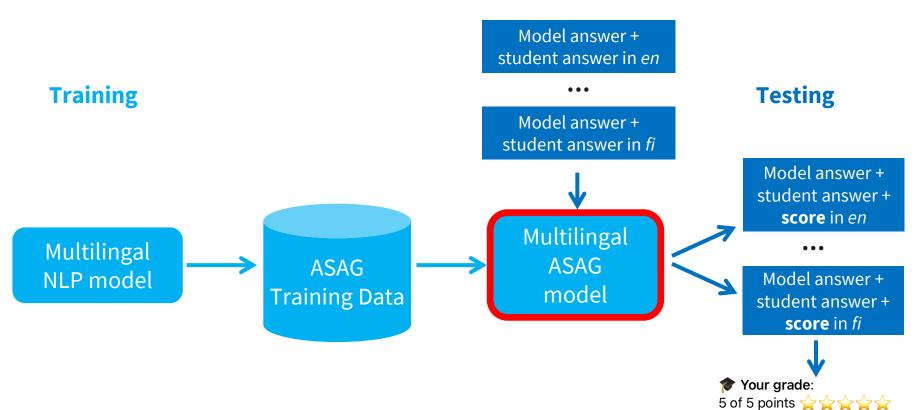
#### **RELATED WORK: CROSS-LINGUAL AUTO-GRADING**



AUTOMATIC SHORT ANSWER GRADING

**Deep learning** 

Cross-lingual
Automatic Short
Answer Grading



Graphic Source: Custom Depiction.

### RELATED WORK: AUTO-GRADING

	multi+	mono						
	en	de	nl	jp	zh	fi	6	
en	0.45	0.61	0.64	0.68	0.63	0.63	0.43	0.43
ceb	0.70	0.73	0.72	0.68	0.72	0.71	0.63	-
sv	0.63	0.67	0.68	0.73	0.72	0.68	0.48	-
de	0.64	0.51	0.67	0.70	0.70	0.65	0.46	0.45
fr	0.61	0.66	0.64	0.67	0.70	0.67	0.54	-
nl	0.62	0.64	0.52	0.70	0.73	0.67	0.45	0.47
ru	0.68	0.73	0.83	0.74	0.75	0.78	0.52	-
it	0.62	0.65	0.72	0.71	0.73	0.70	0.52	-
es	0.61	0.68	0.76	0.68	0.72	0.65	0.49	-
pl	0.62	0.71	0.77	0.69	0.72	0.68	0.51	-
vi	0.71	0.72	0.84	0.77	0.73	0.71	0.52	-
jp	0.66	0.70	0.73	0.49	0.63	0.71	0.44	0.53
zh	0.63	0.71	0.77	0.69	0.50	0.79	0.41	0.44
ar	0.72	0.78	0.85	0.78	0.76	0.76	0.59	-
uk	0.65	0.70	0.82	0.73	0.73	0.75	0.54	-
pt	0.59	0.67	0.75	0.69	0.73	0.69	0.50	-
fa	0.64	0.66	0.71	0.67	0.70	0.69	0.56	0
ca	0.64	0.70	0.74	0.70	0.76	0.67	0.53	
ST	0.69	0.81	0.83	0.76	0.79	0.86	0.56	
id	0.66	0.68	0.69	0.70	0.79	0.63	0.49	- 4
no	0.63	0.69	0.65	0.75	0.71	0.69	0.45	-
ko	0.70	0.70	0.76	0.66	0.66	0.67	0.58	-
fi	0.69	0.79	0.77	0.77	0.73	0.52	0.47	0.45
hu	0.69	0.76	0.81	0.72	0.76	0.69	0.54	-
cs	0.62	0.77	0.82	0.72	0.78	0.71	0.51	-
sh	0.66	0.77	0.79	0.74	0.78	0.79	0.53	-

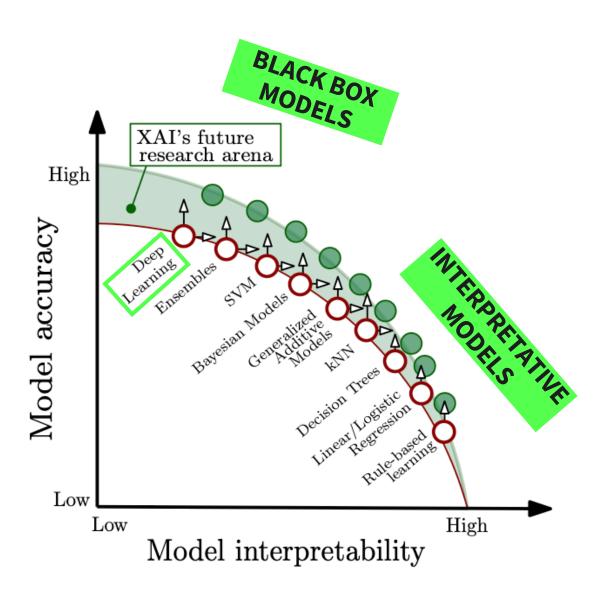


Mean Absolute Error out of 5 points

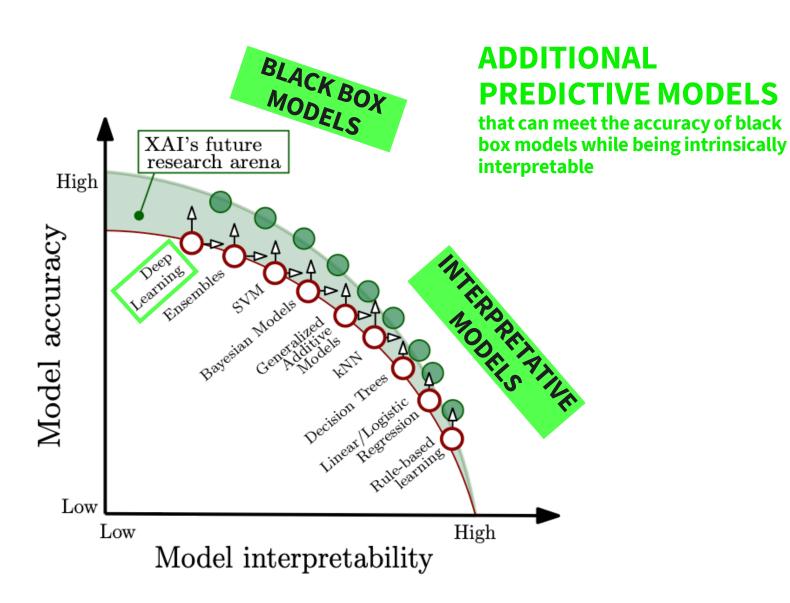
**0.75 POINTS** 

**HUMAN GRADER VARIABILITY** 









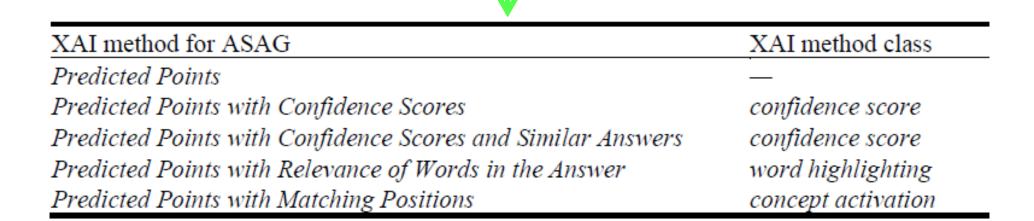
Source: Barredo Arrieta et al. (2020).



XAI method class	Description
confidence score	Certainty of a model's prediction is made interpretable and
	inspectable (van der Waa et al., 2020)
word highlighting	Words are color marked to indicate their relevance towards
	the classification (Ribeiro, Singh & Guestrin, 2016)
concept activation	High level human concepts are used to explain a
	classification (Kim at al., 2018)



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# **EXPLAINABILITY IN**

# **AUTOMATIC SHORT ANSWER GRADING**



AI PREDICTION

2 points

points



AI PREDICTION	CONFIDENCE		
2 points	LOW	92%	HIGH
<b>EXPLANATION</b> 11 of 12 comparable That corresponds to	answers were also rat 92%.	ed with 2 poi	nts by humans.

confidence



AI PREDICTION	CONFIDENCE				
2 points	LOW 83% H				
<b>EXPLANATION</b> 10 of 12 comparable answers were also rated with 2 points by humans. That corresponds to 83%.					
EXAMPLE OF SIMILAR ANSWERS  An animal that lowers or raises its temperature depending on its environment.  Score: 2 points					

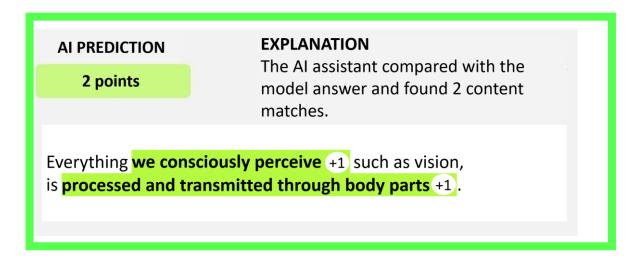
confidence + similar answers



AI PREDICTION	<b>EXPLANATION</b> Highlighted in color, you can see how relevant the individual words were for the AI.				
2 points					
A <mark>kind</mark> of precipitation consisting of small pieces of ice.					
Irrelevant Very relevant					

relevance of words





points + matching positions





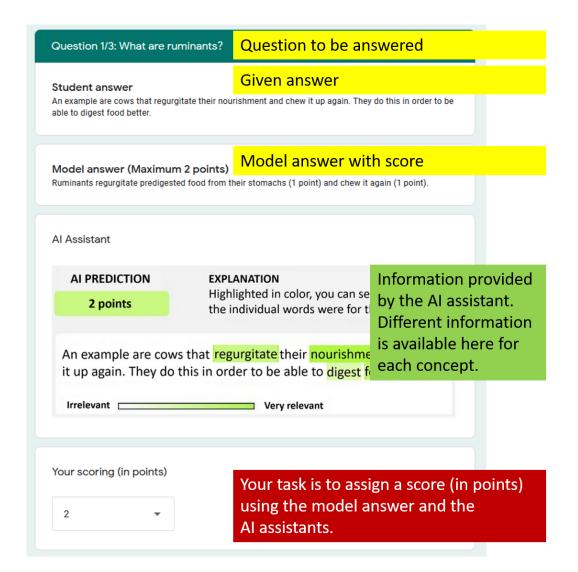
# **USER STUDY**

### **AUTO-GRADING: Explainability**

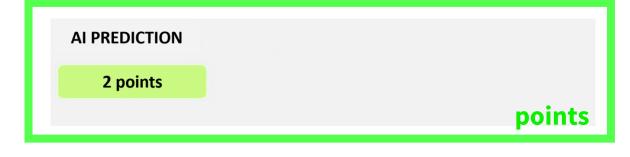


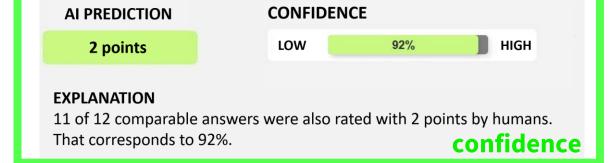
#### **STUDY**

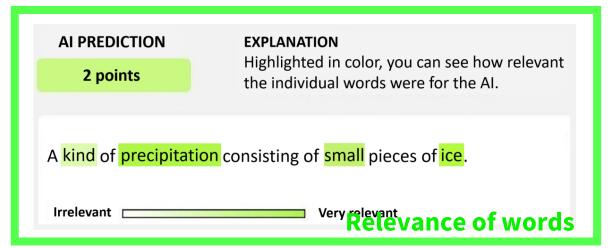
- 5 XAI methods
- **70** professors, lecturers and teachers
- **9** aspects evaluated
  - √ trust, ✓ informative content, ✓ speed, ✓ consistency & fairness, ✓ fun,
  - ✓ comprehensibility, ✓ applicability, ✓ use in exam preparation, ✓ in general

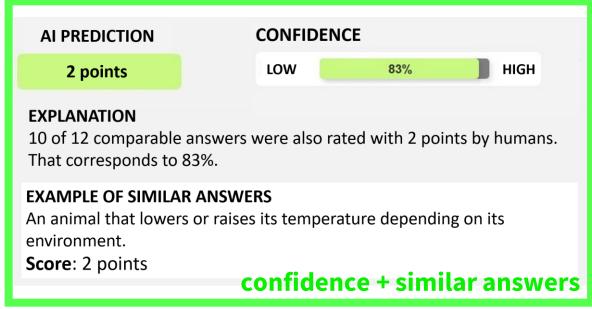


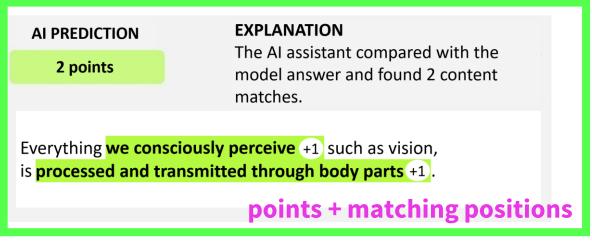








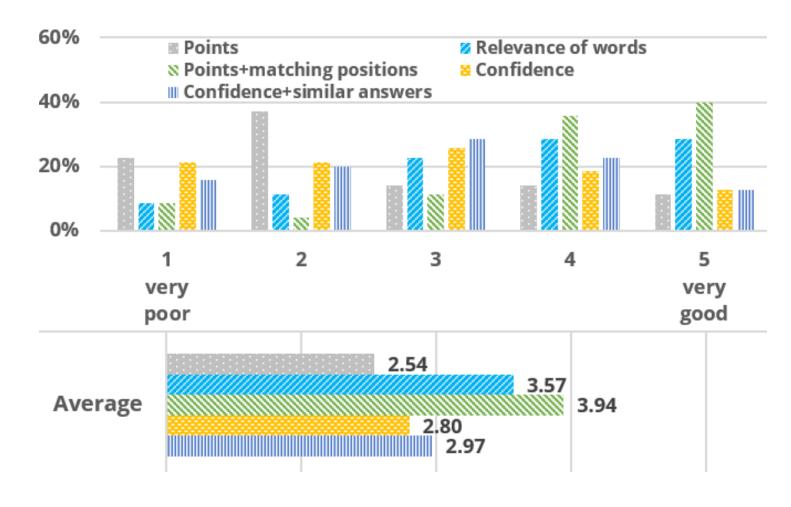




- √ trust, ✓ informative content, ✓ speed, ✓ consistency & fairness, ✓ fun,
- √ comprehensibility, √ applicability, √ use in exam preparation, √ in general



What do you think of the concepts?







# **CONCLUSION & FUTURE WORK**

#### **CONCLUSION AND FUTURE WORK**



#### Conclusion

- Investigated and evaluated different methods for explainability in automatic short answer grading
- Survey of over 70 professors, lecturers and teachers
- Important for them to understand how the AI reaches its scoring and their confidence in an AI grading support increases when it explains itself
- Displaying the predicted points together with matches between student answer and model answer is rated better than the other tested XAI methods.
- Evaluated aspects: trust, informative content, speed, consistency and fairness, fun, comprehensibility, applicability, use in exam preparation, and in general.

#### **Furture Work**

- Analyze the use of our XAI
   methods in interactive training
   programs to prepare students
   optimally for exams
- Direct interpretation of the complex ASAG models could be also investigated



# THANK YOU

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