# **Summary**

This report aimed to find the best metric for estimating the quality of questions based on the problem statement of the NeurIPS competition. In preprocessing, I added confidence, grouped, and dealt with N/A values. In ordering and evaluation, I tried several metrics, including confidence and correct answer ratio. The results showed that the quality of the question is quite related to confidence and correct answer ratio, and their product metric produced the best results.

#### Introduction

The NeurIPS competition (and SSU ML class 2023) presented a challenge: how can we estimate the quality of questions? In response, this report presents a methodology for identifying the best metric for assessing question quality.

### **Methods**

During the preprocessing stage, I added confidence, grouped, and handled missing data. The 'answer\_metadata\_task\_3\_4.csv' dataset was merged to use confidence metadata. Dataframe grouping was then done by QuestionId to estimate each question's quality. To deal with the missing data, I initially dropped rows with N/A values, but the resulting metrics were unsatisfactory. To address this, I filled in the missing values with the mean value.

In the ordering and evaluation stage, several metrics were experimented with, including confidence, correct answer ratio, and their product. The results showed that the product of confidence and correct answer ratio was the most effective metric for determining question quality. The report suggests that lower confidence and correct answer ratio are better indicators of question quality.

You can view all of my works in GitHub(https://github.com/tuxedcat/question\_quality\_assessment).

## Results

Metric	evaluated
Confidence ascending	0.76
Confidence descending	0.36
Correct answer ratio ascending	0.76
Correct answer ratio descending	0.44
Confidence * CorrectRatio ascending	0.8
Confidence * CorrectRatio descending	0.4

# **Discussion**

The results indicate that the quality of questions is highly related to both confidence and correct answer ratio. By multiplying these two metrics, the product metric produced the best estimate of question quality. These findings have practical implications for various domains where question quality is important, such as education and online forums.

# Conclusion

In conclusion, this report presents a comprehensive methodology for estimating the quality of questions using several metrics. The results showed that the product of confidence and correct answer ratio produces the best estimate of question quality. These findings have practical implications for question quality assessment in various domains, including education, online forums, and question-answering platforms. Future work can extend this methodology by (cross-)entropy, incorporating additional features and employing more sophisticated machine learning algorithms.

# References

https://proceedings.mlr.press/v133/wang21a.html https://pandas.pydata.org/docs/