# Automatic Generation of MCQs Using Transformer Mode

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**Abstract***: Multiple Choice questions is the go to assessment technique used for competitive entrance exams, company's aptitude screening process and for the assessment exams of various institutions and universities. The rise of the online paradigm due to the pandemic has also changed the way universities and institutions conduct their assessment exams\* Hence, MCQs play an important role in assessment of skill and knowledge in different domains and situations\* However, it becomes very difficult for a human to generate large amount of MCQs along with good quality distractors in limited time. This problem can be solved using state of the art natural language processing and deep learning techniques. Our work is an attempt to generate multiple choice based questions which can be used for assessment in an exam setting. This paper proposes a system which uses T5 transformer for question generation and other state of the art deep learning technique for distractor generation generate multiple choice questions which resembles the human way of questioning.*

**Keywords:** Natural Language Processing, MCQs, T5 transformer, Deep learning, Distractors, Question generation

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