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**DETEKSI DAN SEGMENTASI IMUNOREAKTIVITAS KI-67 PADA GLIOMA DENGAN PENDEKATAN
DEEP LEARNING BERBASIS**

MASK REGION-BASED CONVOLUTIONAL NEURAL NETWORK (MASK R-CNN)

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