

# **Brain Tumor Diagnosis System Using Vision Transformers**

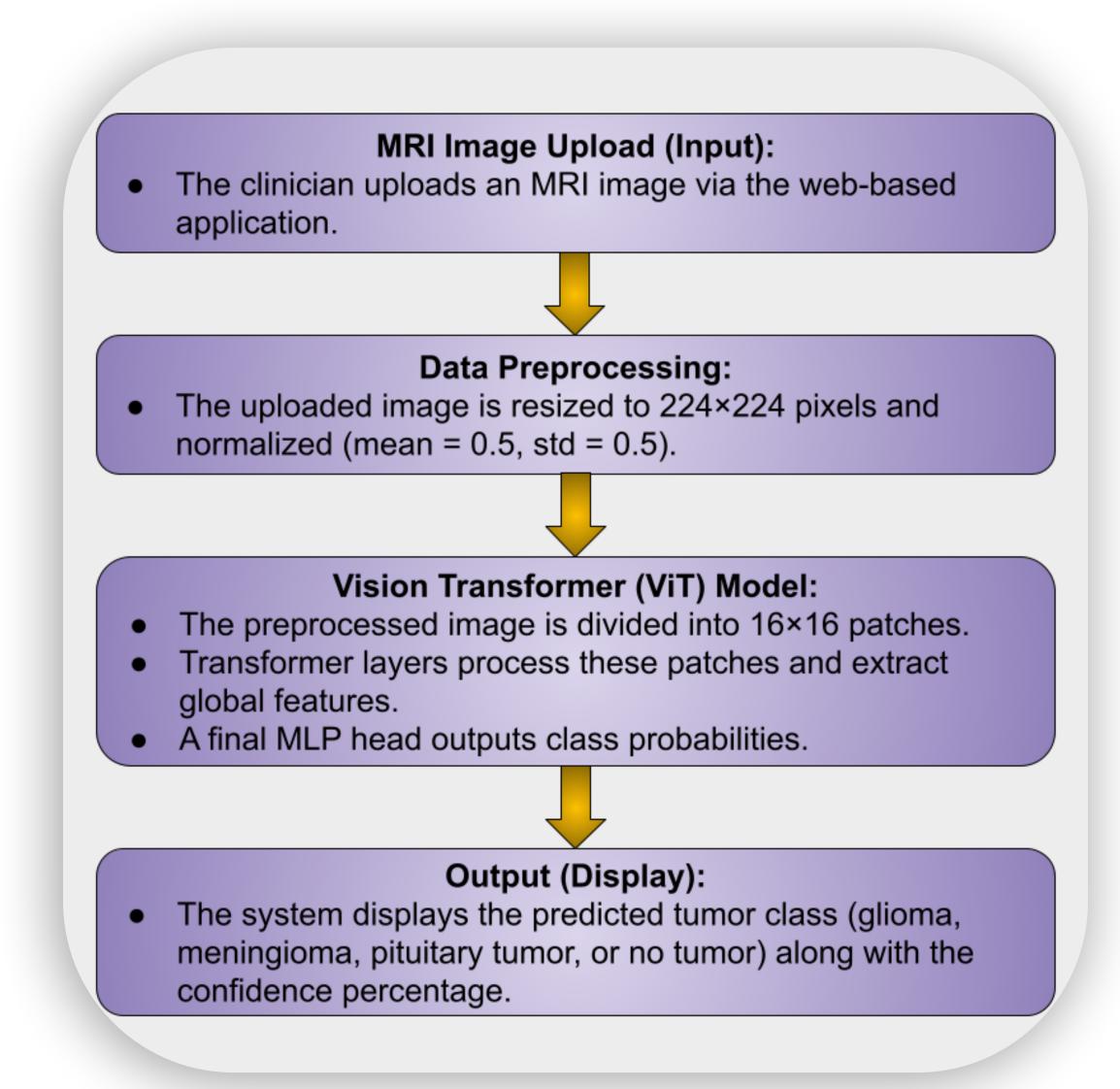
**Andy Achouche** 

# Background

Early diagnosis of brain tumors is critical, yet traditional workflows—spanning patient screening, lab tests, and MRI analysis—are often fragmented and slow. Recent deep learning advances, particularly Vision Transformers, offer a promising alternative for more accurate and scalable MRI image classification.

# **Research Question**

Can a Vision Transformer model accurately classify MRI images for brain tumor diagnosis and be effectively deployed via a web-based application to support clinical decision-making?



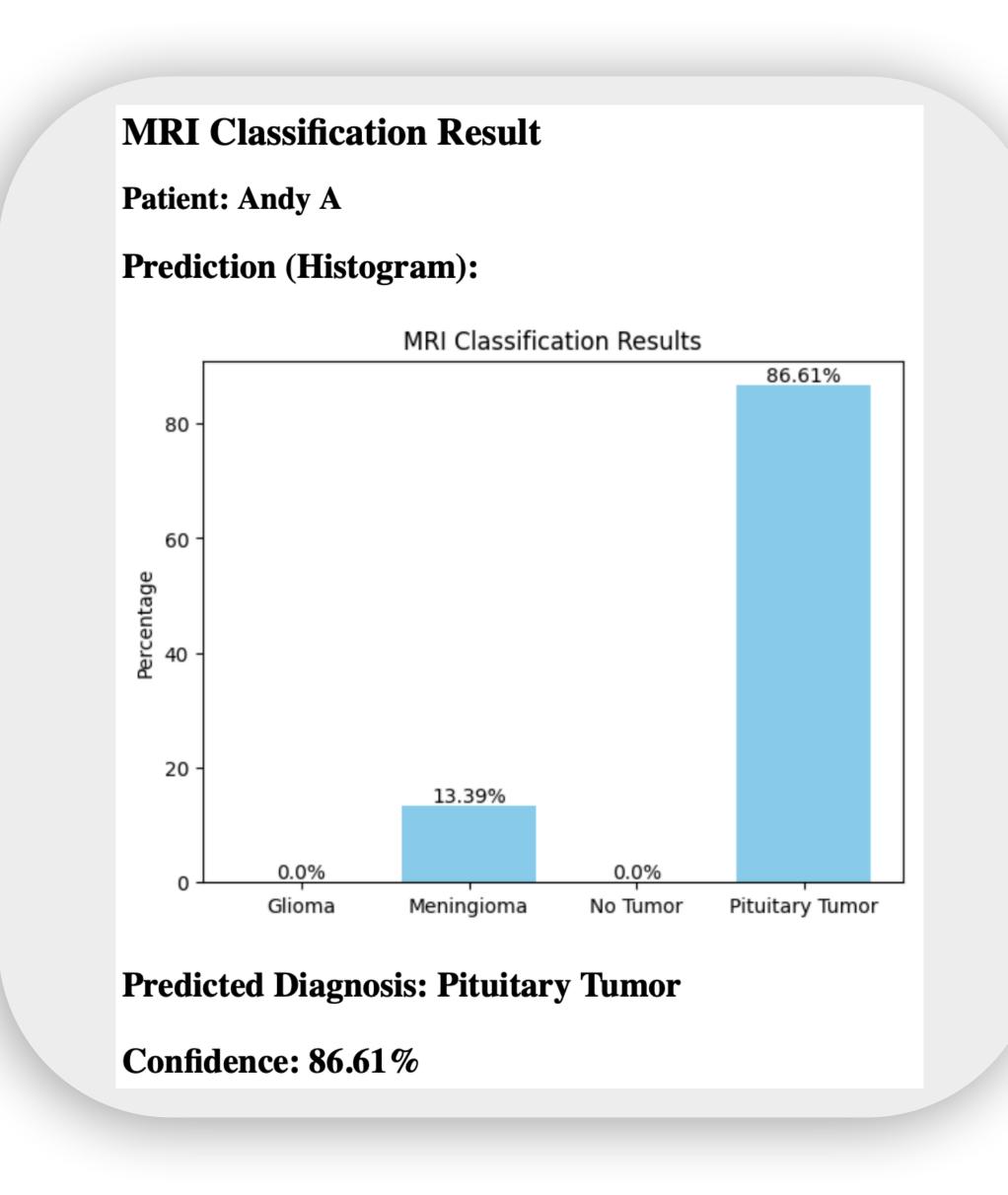
# **Implications**

- Enhanced Diagnostic Accuracy:
  Earlier and more accurate tumor
  detection improves patient outcomes.
- Streamlined Clinical Workflow: Realtime, automated MRI analysis reduces delays and supports faster decisionmaking.
- Increased Efficiency: Automation minimizes radiologist workload and reduces misclassification rates.
- Scalable Al Framework: Establishes a robust platform for future integration of advanced diagnostic tools.

## **Materials & Methods**

- Programming: Python 3.10, PyTorch (vit-pytorch), Jupyter Notebook
- Hardware: Apple Silicon (M3 Pro with GPU/MPS), AWS EC2 (p3.2xlarge)
- Web & Deployment: Flask, Docker, Gunicorn
- Libraries: scikit-learn, Matplotlib





### References

- Dosovitskiy, A., Beyer, L., Kolesnikov, A., et al. (2021). An Image is Worth 16x16 Words: Transformers for Image Recognition at Scale. ICLR 2021.
- Vaswani, A., Shazeer, N., Parmar, N., et al. (2017). Attention is All You Need. NeurIPS 2017.
- Nickparvar, M. (2021, September 24).
   Brain Tumor MRI dataset. Kaggle.
- AWS. (n.d.). Deployments on an EC2/ On-Premises compute platform - AWS CodeDeploy.

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