# Medically Informed Stable Diffusion (MISD)

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## Introduction

#### **Medical Interpretability**

#### Importance:

- Clinical Relevance
- Diagnostic Insights
- Ethical Considerations

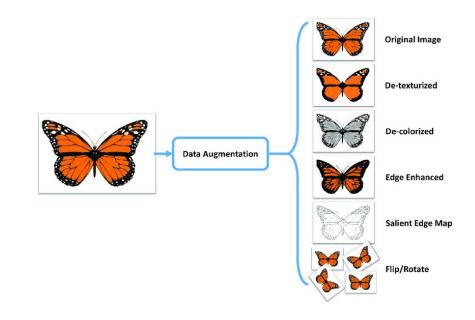
#### Challenges:

- Data Complexity
- Biological Realism
- Validation / Evaluation
- Incorporating Domain Knowledge



### **Data Augmentation for Medical Imaging**

- Data Transformation
  - Affine Transformation
  - Erasing Transformation
  - Elastic Transformation
  - Pixel-Level Transformation
- Generation of Artificial Data
  - Generative Adversarial Networks
  - Feature Mixing Methods
  - Model-based Modification
  - Reconstruction-based Methods



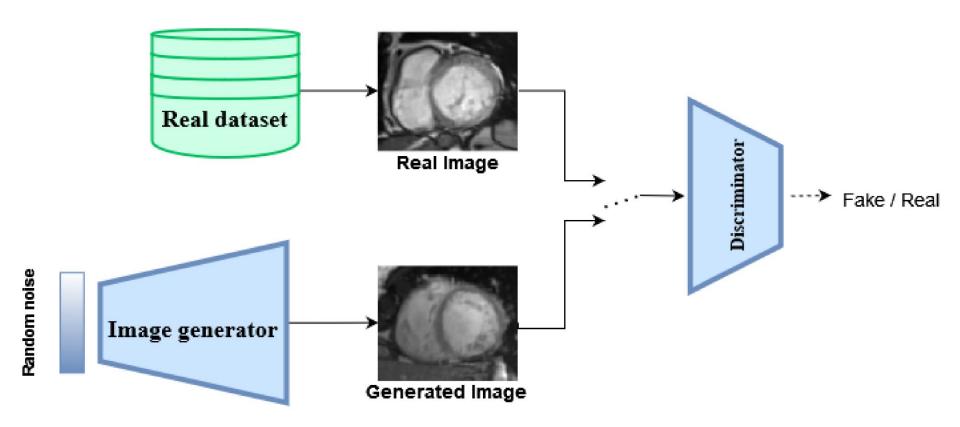
Gaussian Blur Crop **Ghosting Artefact** Spike Artefact Random Swap Original image Flip Gaussian Noise Affine transformation **Motion Artefact Biasfield Artefact** 

### **Image to Image with GAN**



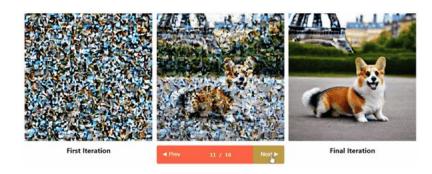


- Generative Adversarial Network
  - Generator vs Discriminator
- Source-Target Mapping
  - Capture characteristics of image collection
  - Translate characteristics into other images
- Synthetic Data
  - Cheaper
  - Cannot be re-identified



#### Image to Image with Stable Diffusion

- Diffusion Model
  - Generates realistic, high-resolution images
  - Class-conditional image synthesis
- Trainable with limited data
  - Requires 200 images to fine-tune
- Synthetic Data
  - Cheaper
  - Cannot be re-identified

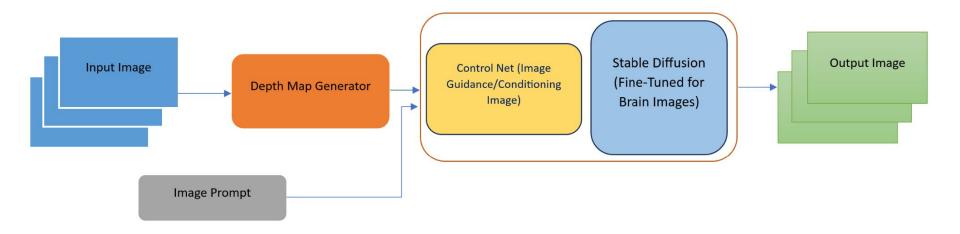


# **Proposal**

#### **Proof of Concept - Model Structure**

- 1. ControlNet
  - "Conditioning Image" guides image generation by adding constraints
- 2. StableDiffusion
  - Uses prompt to generate image
- 3. UNet
  - Image to image training to refine generated image
- 4. Textual Inversion
  - Trains text embeddings

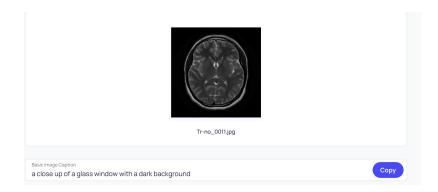
## **Diagram**

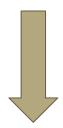


## **Proof of Concept**

#### **Prompt Engineering**

- Training images require custom captions
- image-to-caption generators are not so good
- Must engineer effective & unique captions





- Intricate brain MRI reveals neural networks in stunning detail
- 2 CT scan showcases brain structures with high-resolution clarity
- B Detailed MRI maps brain activity patterns for neurological insights
- CT imaging unveils brain anatomy in precise cross-sectional views
- 5 Brain MRI offers unparalleled insights into neural connectivity
- 6 CT scan captures intricate brain structures in detailed cross-sections

#### **Stable Diffusion: prompt only**

"brain CT scan"

- Abstract concept of 'brain'
- Cinematographic
- Not medically accurate
- Undesired details



No Style

Default - Cinematographic

#### **ControlNet:** image-guided with prompt

"brain CT scan"

- No concept of 'brain'
- Follows depth-map
- Not medically accurate
- Undesired details



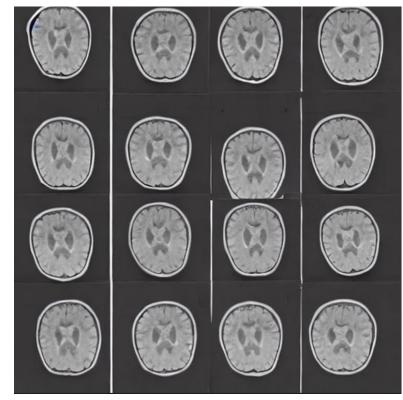
Original

ControlNet

#### Fine Tuned: pretrained model with prompt only

"brain CT scan"

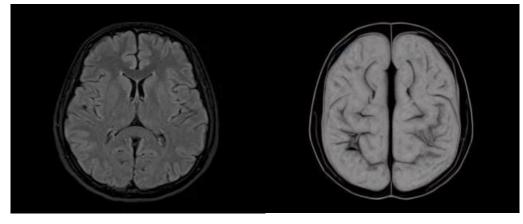
- Fine-tuned concept of "brain"
- Ability to create multiple images with slight variations in the internal structure while maintaining overall integrity
- Needs more guidance in terms of more finer details for better image generation



#### Fine Tuned + ControlNet: image and prompt

"brain CT"

- Fine-tuned concept of "brain"
- Guided image generation
- Enforces constraints



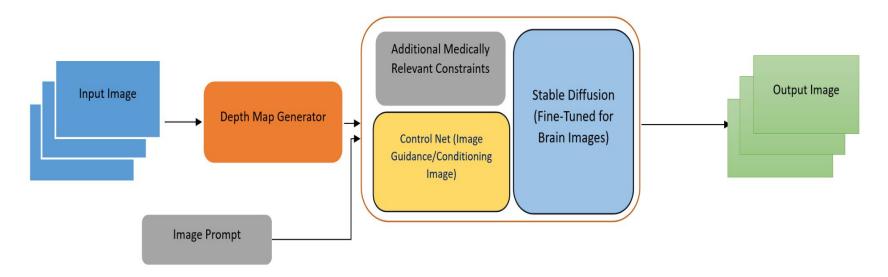
Original

Fine Tuned + ControlNet

# **Next Steps**

#### **Biology-Informed Neural Network**

- Add constraints to optimization algorithm
- Constraints inform model how to properly create images



## **Conclusion**

## **Thank You**

#### References

- https://hai.stanford.edu/news/could-stable-diffusion-solve-gap-medical-i maging-data
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