



# Emory CXR An Embeddings Study

**Thoracic Rangers**



# What?



- ◆ An interactive visualization tool for guiding subgroup analysis.
  - Comparing how different models interpret CXR embeddings.
  - Understand outlier regions

# Methods

- Dataset: Emory CXR
- The models:
  - RaD-DINO
  - MedGemma
  - CheXagent
  - MedImageInsight
  - BiomedCLIP
- Visualization Tool: Embedding Atlas



# Process

- + Step 1 - Exploratory Data Analysis on CXR all metadata and labels
- + Step 2 - Select 100k samples for each embedding model, same SOP for direct comparison
- + Step 3 - Dimensionality reduction of the embeddings.
- + Step 4 - Generate 2D Visualization of Embeddings using UMAP with Cosine Similarity
- + Step 5 - Interactive Real-time Clustering (Embedding Atlas)
- + Step 6- Example usage: Sample images from each cluster to provide insights and guide subgroup Analysis.



# EDA - EMORY CXR



- The Emory CXR dataset contains Chest X-Ray (CXR) images embeddings

Metadata: (2430209, 19)

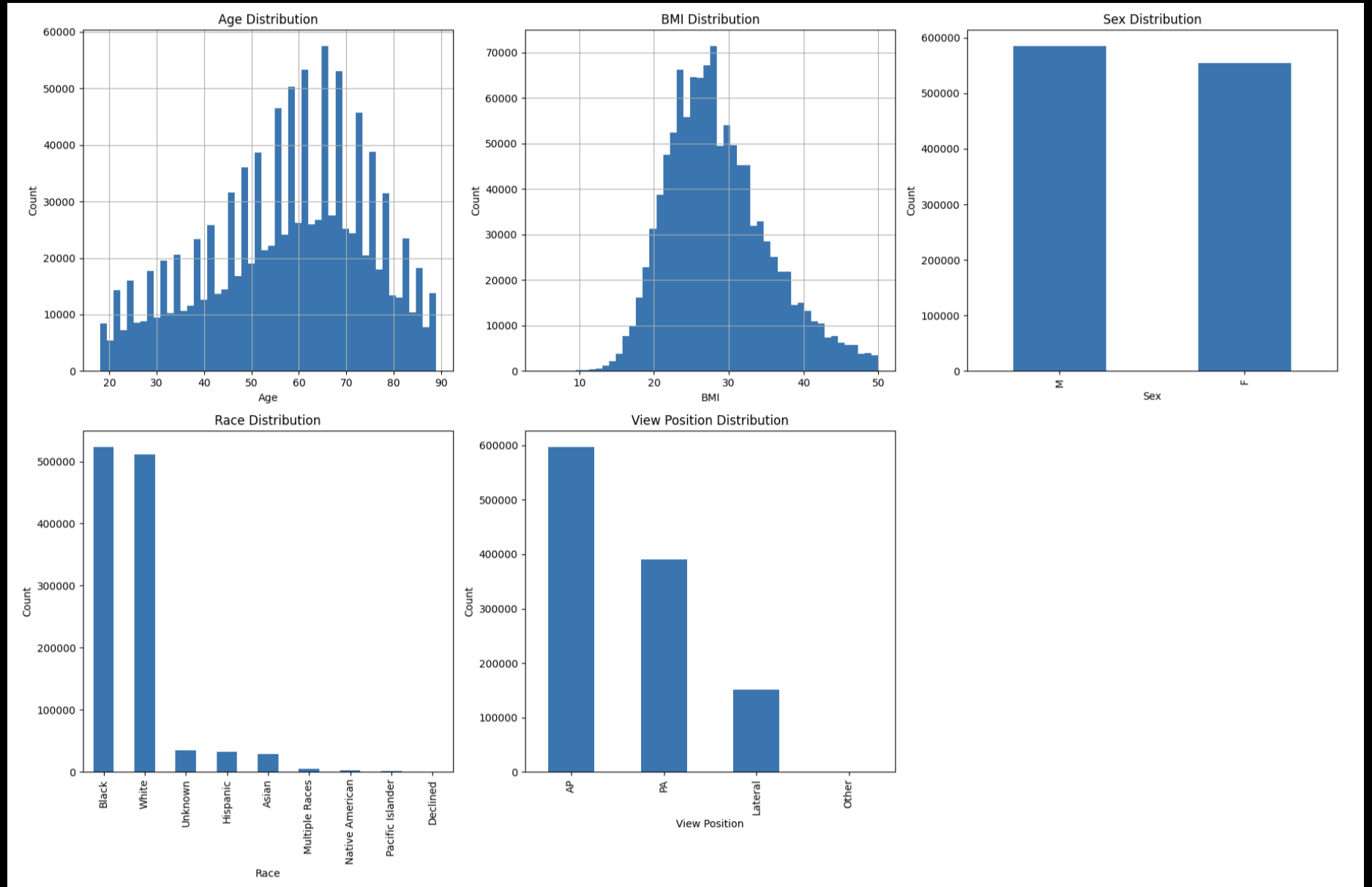
Findings: (1323308, 15)

Cardiomegaly: (2486502, 4)

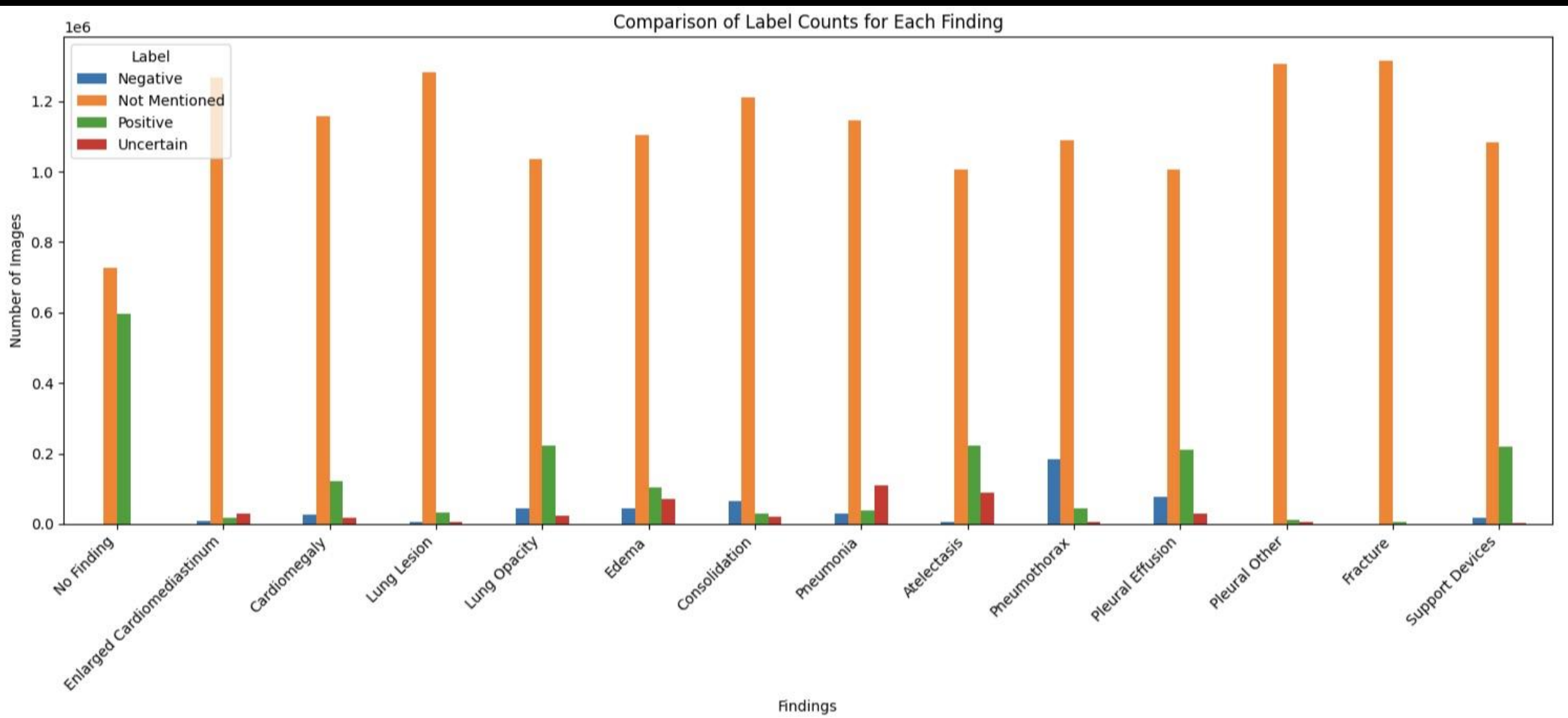
Manufacturer: (1853067, 4)

SDOH: (159946, 388)

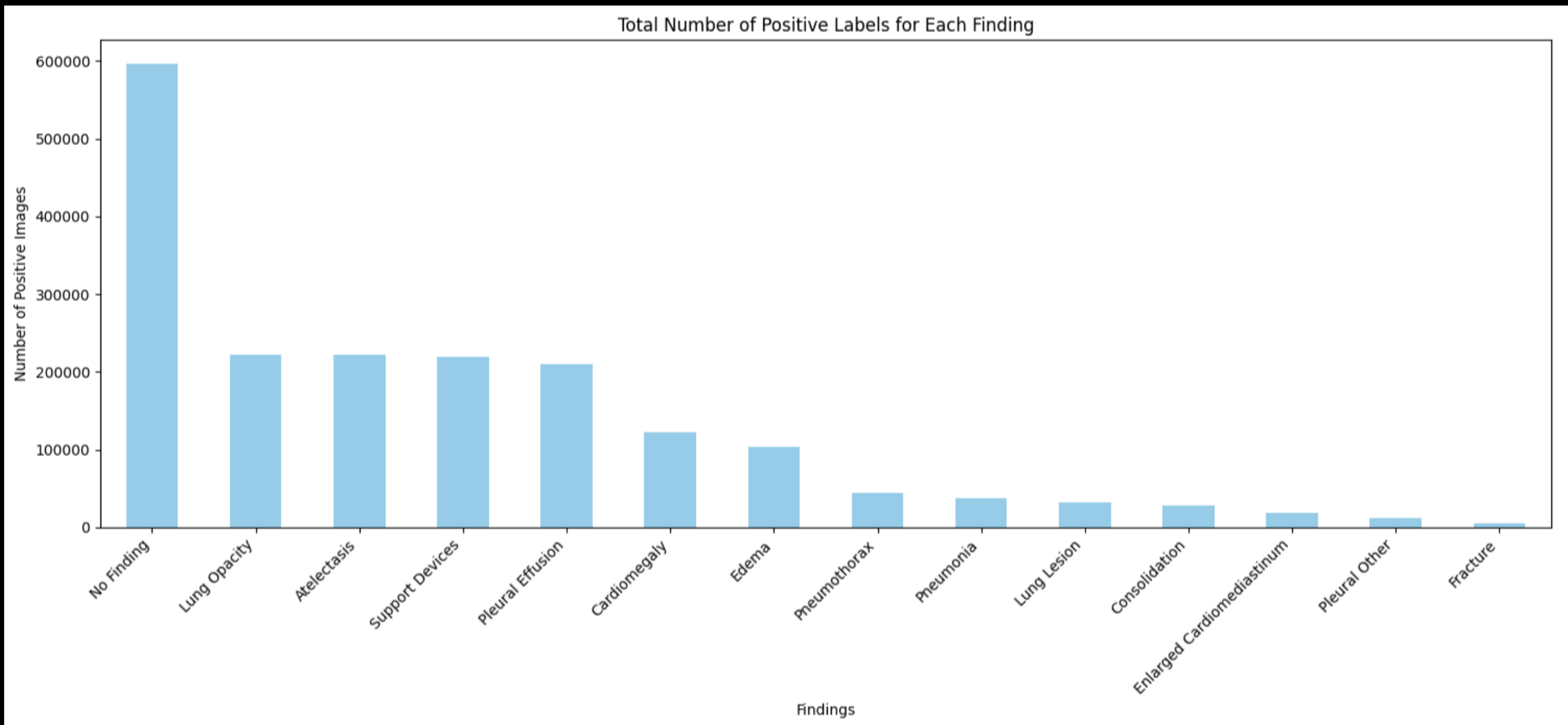
# EDA - EMORY CXR



# EDA - EMORY CXR

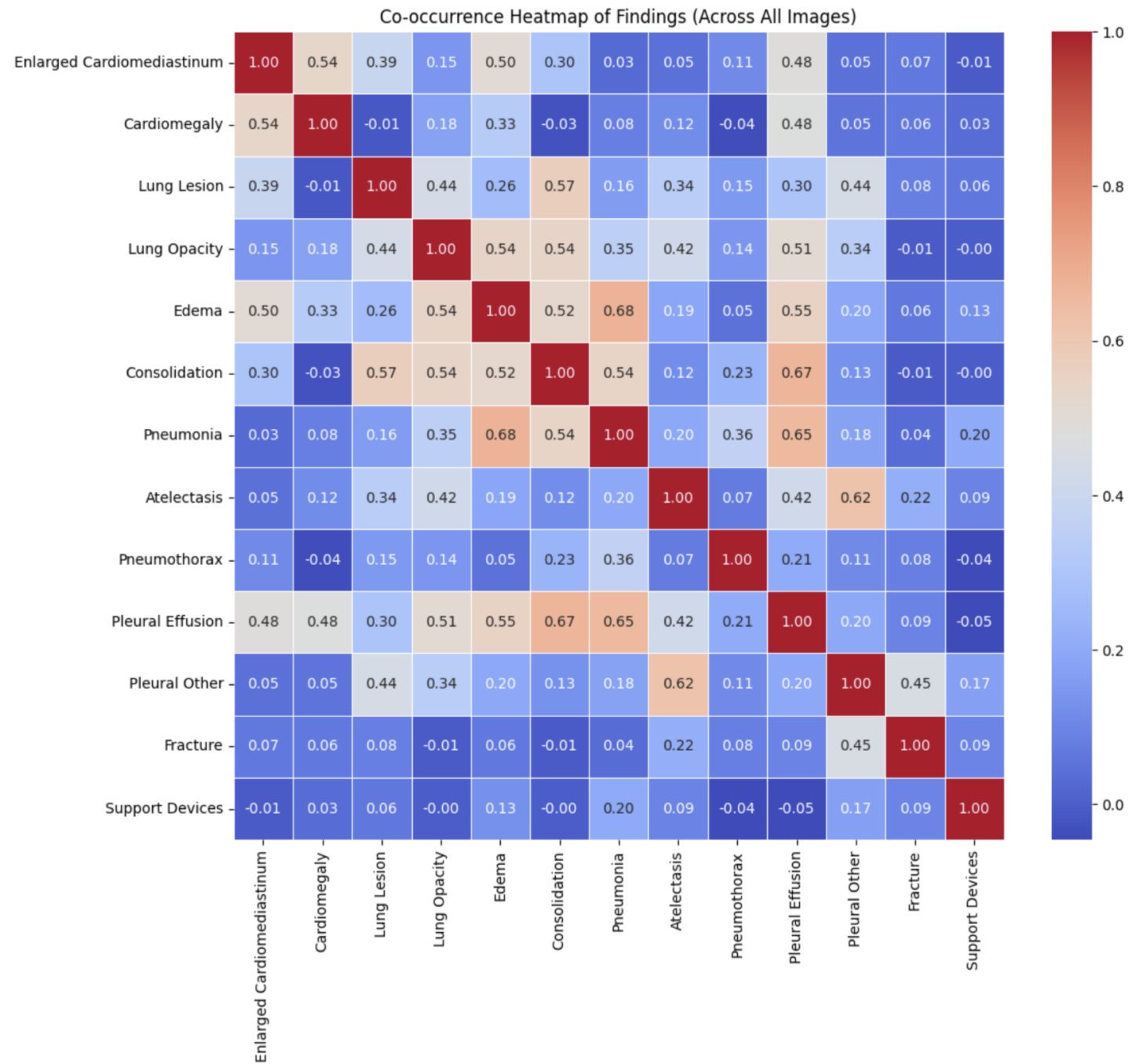


# EDA - EMORY CXR



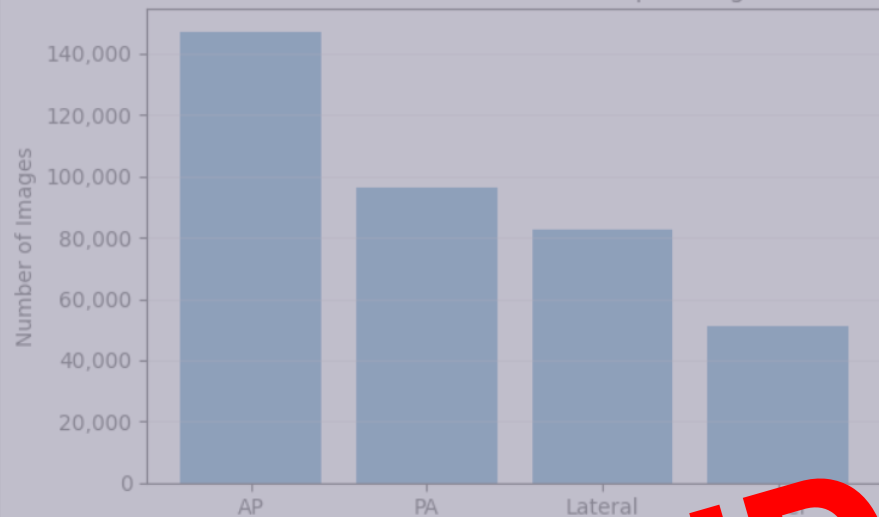


# EDA - EMORY CXR



# EDA - MIMIC CXR

View Position Distribution (Unique Images)



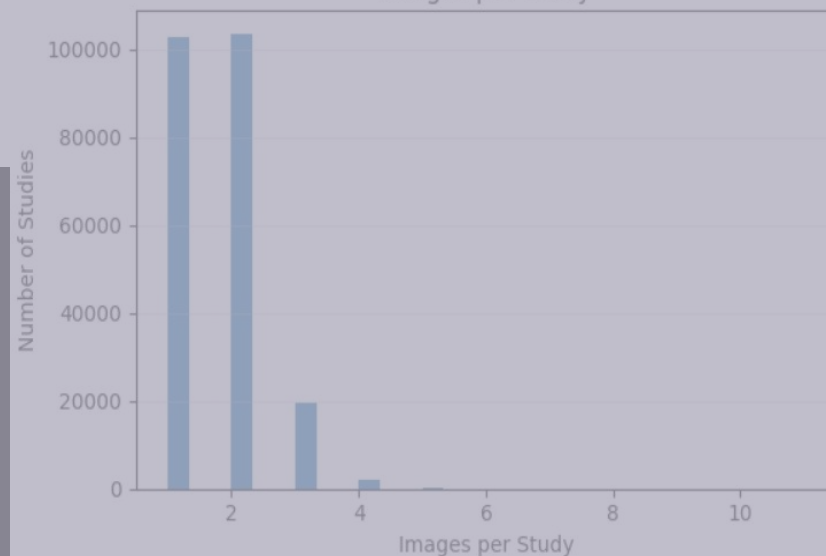
Images per Year



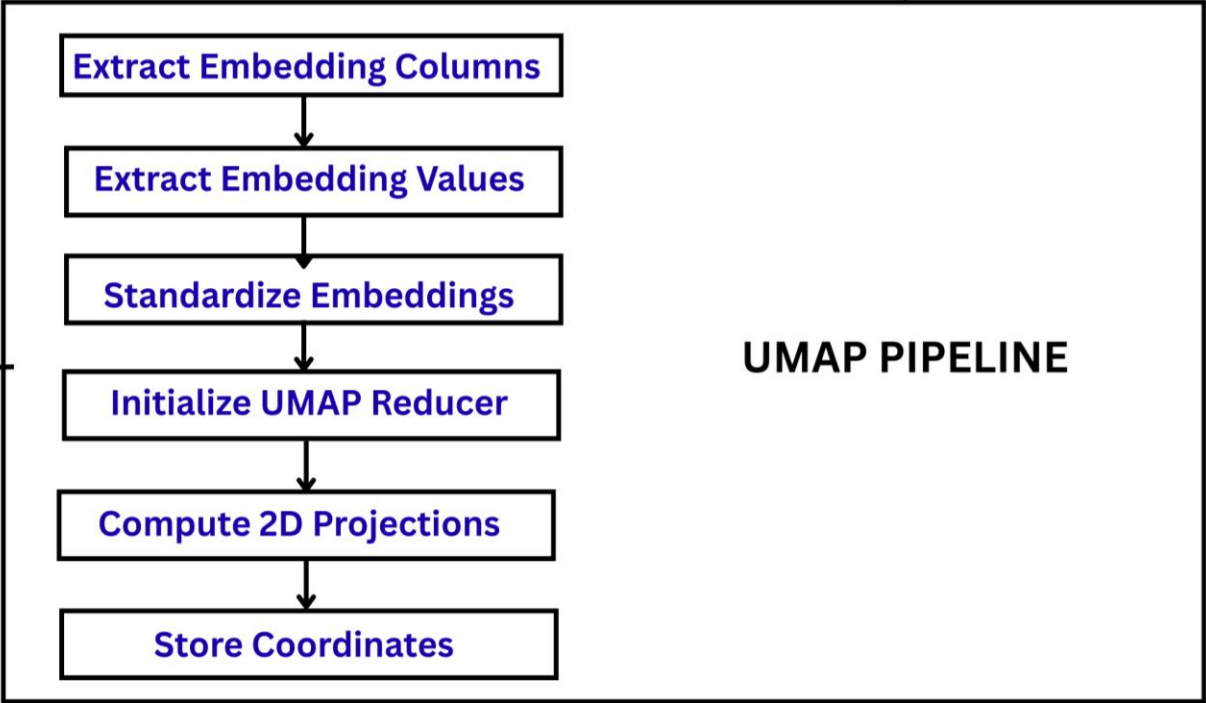
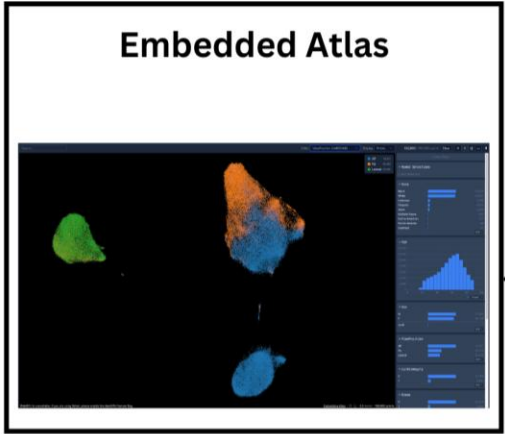
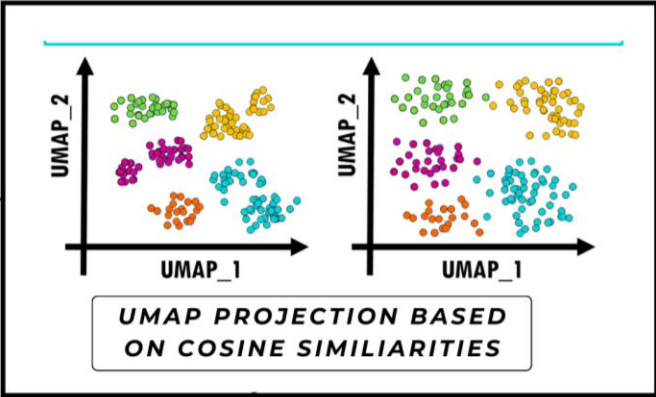
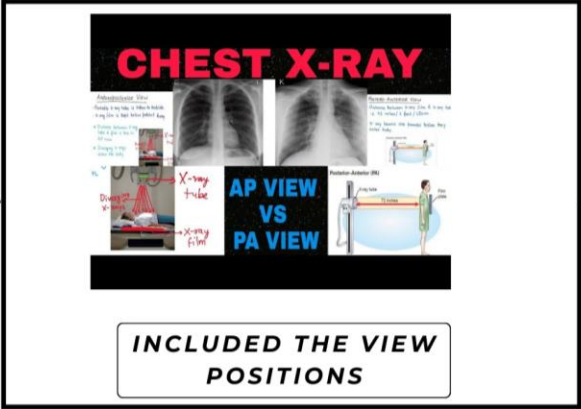
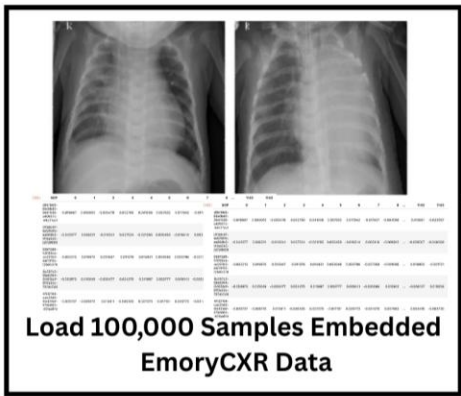
CheXpert Counts — Including Missing

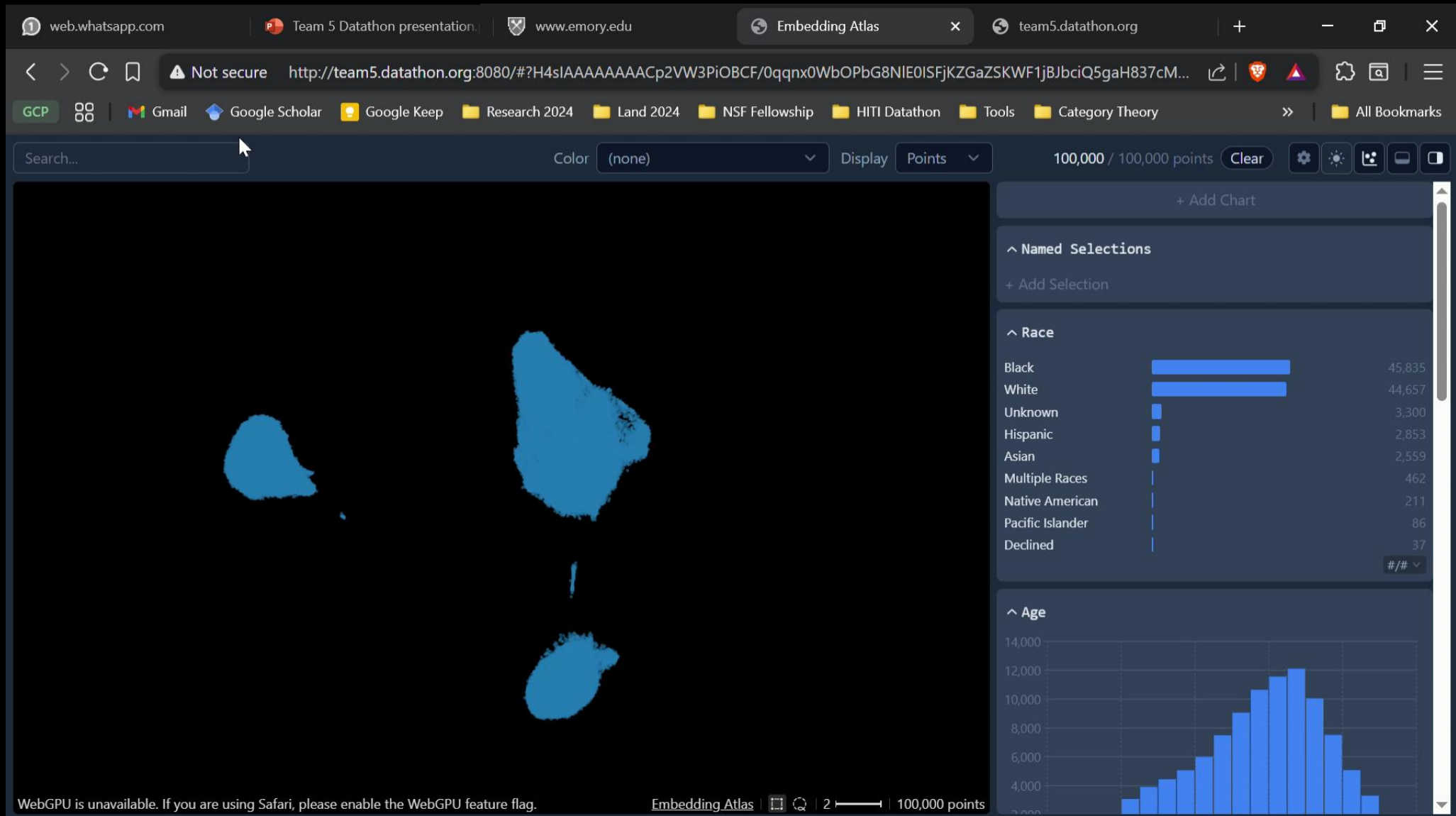


Images per Study



# DEMO PIPELINE







## Our Limitations - Design

- Sampling method may not be accurate representative of the full dataset.
- Quantification and comparison of embeddings is a difficult unsolved problem.
- The tool still requires a lot of expansion to interpret/analyze the embedding atlas.



# Our Limitations - Practical

- Time consumed identifying team dynamics and fully understanding our data and tools.
- Resource limitations for fine-tuning the embeddings on foundation models.



# What is the way forward?



- Compare clustering when a single feature is mis-labelled.
- Comparing different CXR datasets on a same embedding model.
- How well models trained on each set of embeddings classify a single outcome (e.g. pneumonia)





THORACIC RANGERS



# Appendix

# Model Classification



Model	Input/Output	Use Case	Clinical Relevance
MedGemma	Image + Text	Radiology Reports	Improve Workflow Increase Efficiency
BioMedCLIP	Image + Text	Radiology Reports	Improve Workflow Increase Efficiency Knowledge Transfer
RaD-DINO	Image	Classification	Encoding
MedImageInsight	Image	Classification	Encoding
CheXpert	Image + Text	Labels – Embeddings	Training Models