

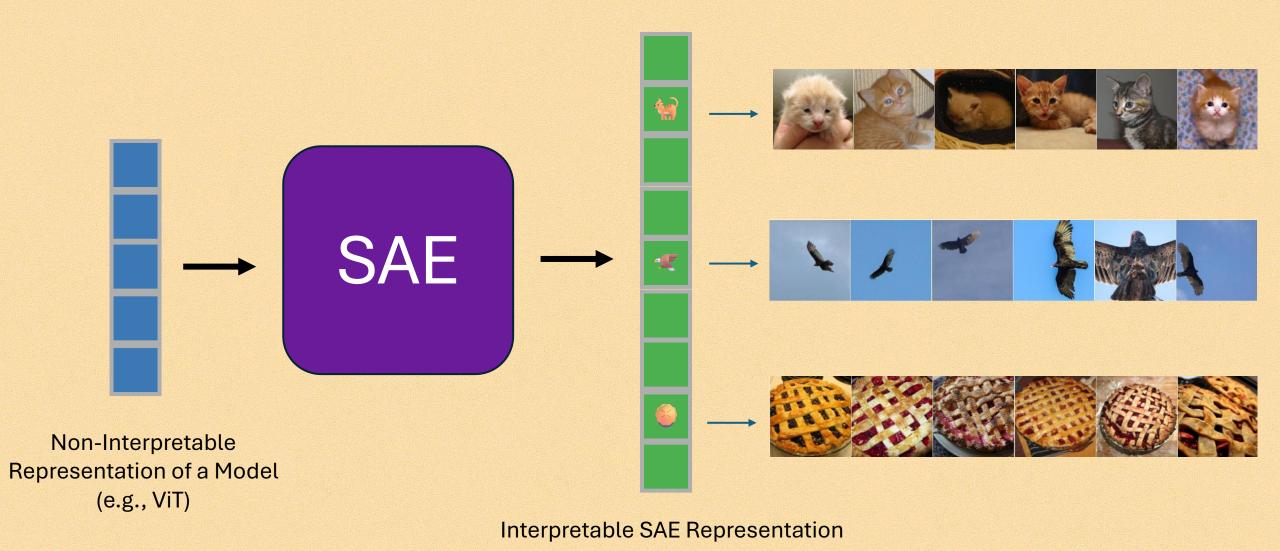


Concept-Aligned Sparse Autoencoders for Cross-Model and Cross-Modal **Interpretability** 

"Scaling Interpretability across models"

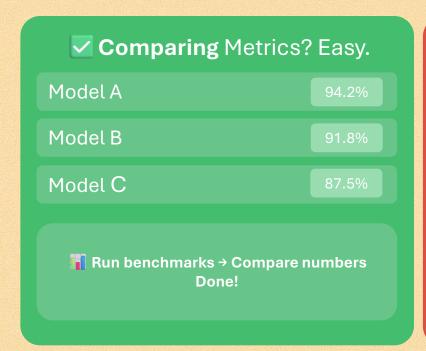
Ali Nasiri-Sarvi • Hassan Rivaz • Mahdi S. Hosseini Concordia University • Mila-Quebec Al Institute

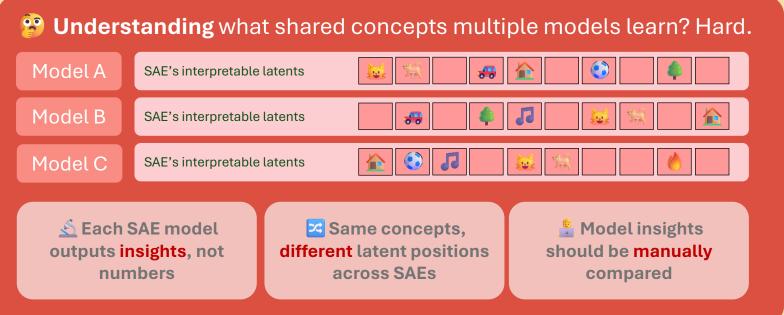
# **Sparse Autoencoders Make Neural Representations Interpretable**



## What if we want to understand the representation of multiple models?

SAEs Help... But Only One Model at a Time (comparing many gets expensive)







#### **Scaling Problem**

Thousands of neurons to analyze per model
Matching concepts by hand
Time-consuming, error-prone



#### **Expert Bottleneck**

Some domains require expert interpretation (e.g., medical)

High cost and slow progress



## O(n<sup>2</sup>) Time Consumption

Each model needs (n-1) comparisons

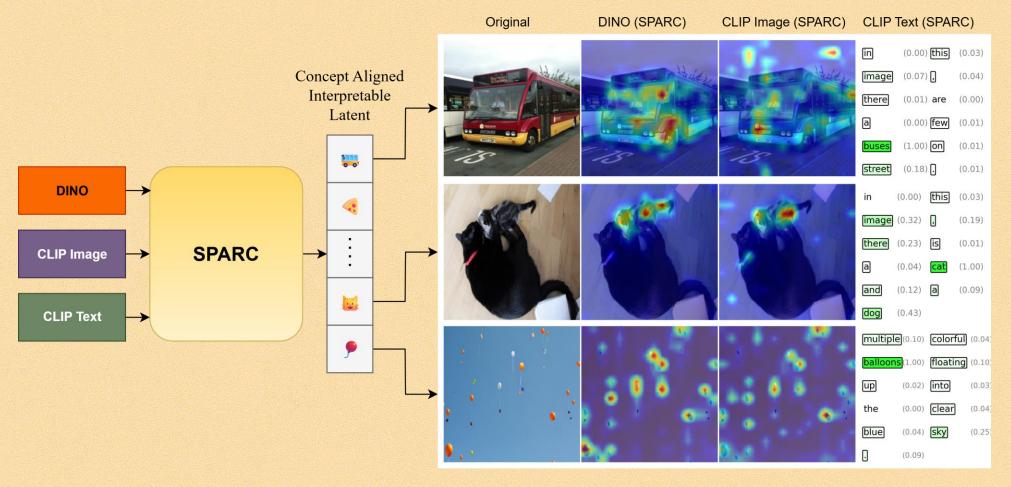
Manual concept matching

Complexity explodes with models

### **SPARC: A unified solution**

Sparse Autoencoders for Representation of Concepts

- Key idea: Build a single, shared interpretable latent space for multiple models
- Forces concept alignment directly through shared architecture
- Works across different model architectures and modalities (vision, language)



SPARC allows
direct comparison
across models and
modalities without
manual concept
matching

## **SPARC** aligns the learned SAE Concepts

Before SPARC: The same latent index encodes different concepts in each stream (welding, roosters, random captions) → no shared meaning.

DINO

CLIP-Image

**CLIP-Text** 



# After Applying SPARC



After SPARC: The same latent index now retrieves the same concept (kittens) in all streams.

DINO

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