

## Multi-Dimensional Features

What are the *representations* that language models use?

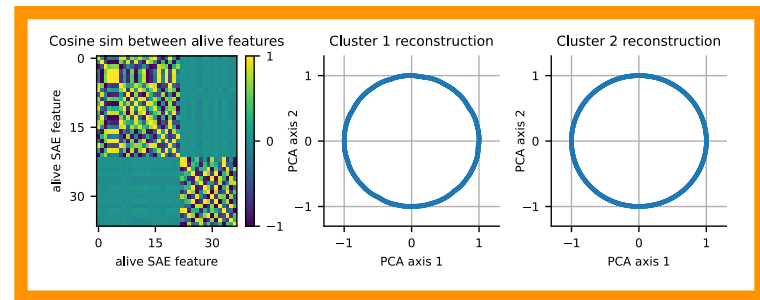
**Our hypothesis: representations are fundamentally multi-dimensional.**

We use **sparse autoencoders** to automatically find multi-d features

### Multi-D Feature Search

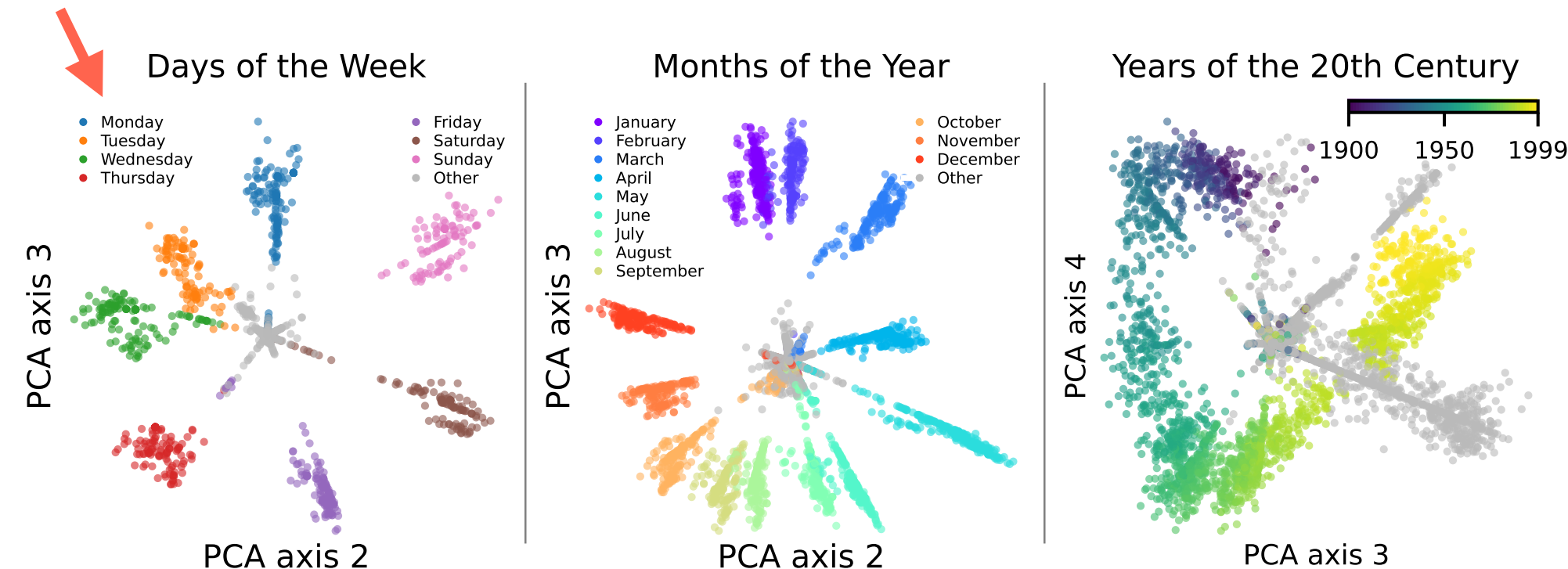
1. Train a residual stream SAE
2. Cluster SAE decoder vectors
3. Run SAE on LLM hidden states, limit reconstruction to each cluster
4. Examine reconstructions for multi-d features

Works on **toy data**



And finds **days of the week** and **months of the year** features in

**GPT-2** and **Mistral 7B**

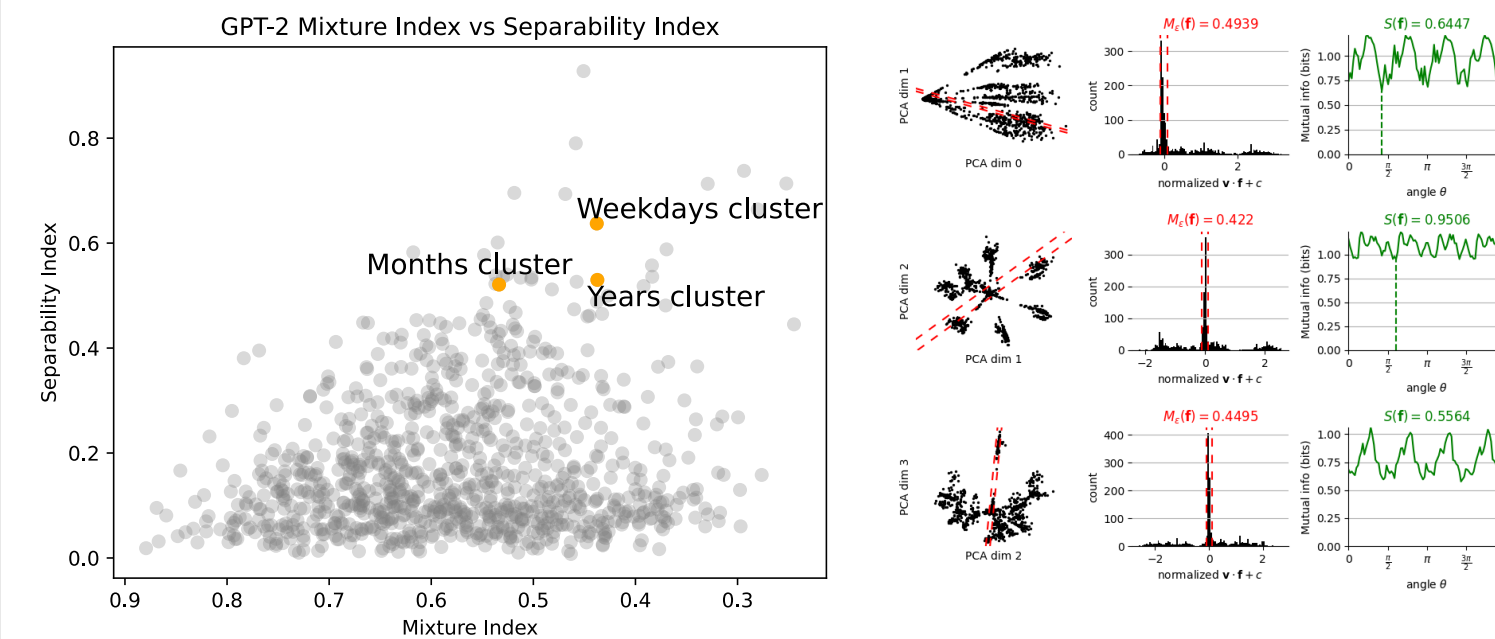


## Multi-D Feature Irreducibility

Which clusters are “real” multi-d features? We say a multi-d feature is **reducible** if either:

1. It is **separable**: it is the sum of two statistically independent features.
2. It is a **mixture**: it is the sum of two features that never co-occur.

Our manually identified clusters score highly!



## LLM Modular Addition

We investigate modular addition as a task that might use these circles:

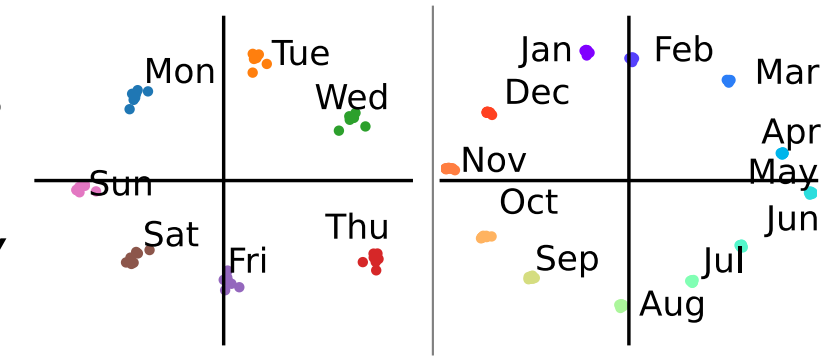
**Weekdays Task:** Two days from Monday is \_\_\_\_\_  
**Months Task:** Six months from November is \_\_\_\_\_

Llama and Mistral (but not GPT-2) are good at this task:

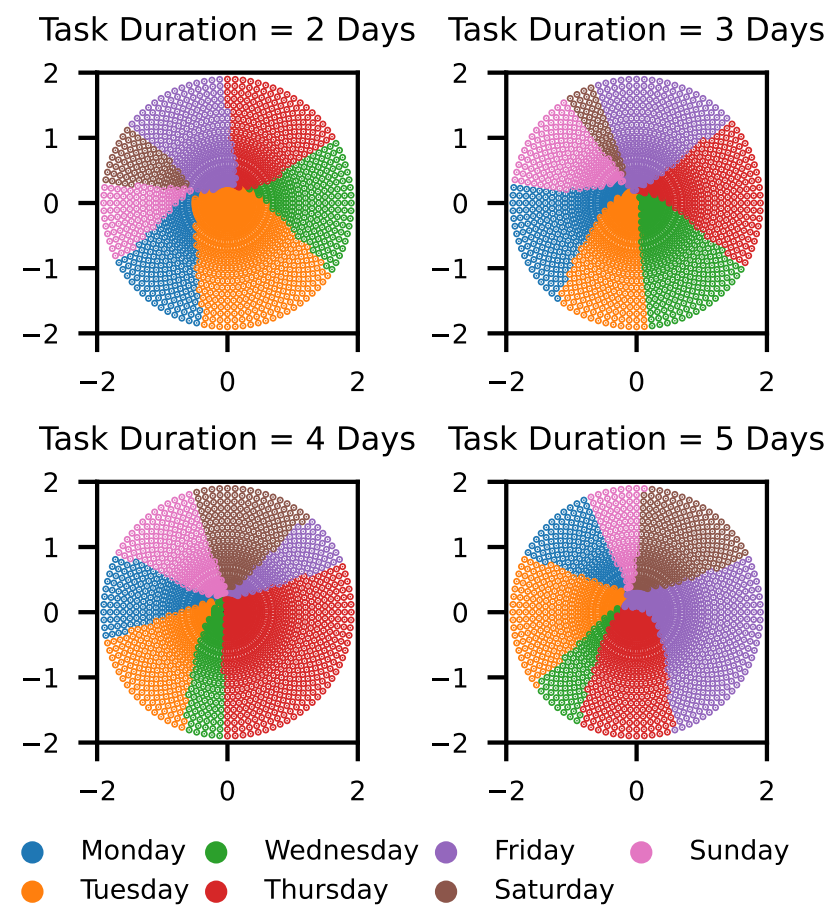
Model	Weekdays	Months
Llama 3 8B	29 / 49	143 / 144
Mistral 7B	31 / 49	125 / 144
GPT-2	8 / 49	10 / 144

## Circular Representation Interventions

PCAs of the activations form circles in days/months! *Do LLMs really use these circles?*



**Yes, the circle is causal!** We perform *interventions* by changing the model’s hidden state along the circle, which predictably changes the model’s output



## Bonus Circles

