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Single Feature Tips the Balance: Reducing Language Model Over-Refusal with Sparse Representations

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Motivation

- 1. Safety-aligned LLMs tend to over-refuse.
- 2. LLMs can be steered via representation editing.
- 3. The benefit of Sparse Autoencoder (SAE) representations:
 - Atomic, specific patterns
 - Interpretable
 - Learned in an unsupervised way

Key Idea

We treat each SAE feature as high-quality "atomic feature" and analyze its individual impact on refusal behavior.

What SAE Features Influence Refusals?

Examples in Llama3.1-8B-Instruct

Category	Feature ID	Feature Description			
Harmful Compliance	6536	Symbols and special characters that express strong emotions or reactions			
	20747	JavaScript code and comments			
	1398	Numerical values or quantities			
	5900	Non-English characters or symbols			
	28804	Abbreviations or acronyms related to technology and medical terms			
Safe Compliance	15607	Punctuation marks that indicate pauses or breaks in sentences			
	3595	Mentions of Chipotle and its promotional activities			
	17456	Occurrences of quotation marks in the text			
	9765	Keywords and phrases related to financial transactions and investments			
	19556	Instances of dashes or em dashes indicating interruptions or transitions in thought			
Refusal Trigger	29351	References to budgetary terms and financial concepts			
	7049	Expressions of gratitude and customer recommendations			
	22814	References to political themes and actions			
	3120	Phrases indicating blame and responsibility in relationships			
	28473	Phrases and concepts related to perception and reality			

Intervention with SAE vectors

TL;DR:

We intervene on models using SAE decoder vectors to analyze feature effects, identifying three categories: harmful compliance, safe compliance, and refusal triggers.



We extract vectors from SAE decoders and intervene in the model by adding them to specific layers. By measuring harmful scores and over-refusal scores after intervention, we categorize features into three types:

1. Harmful Compliance: features with highest harmful scores

Feature ID	Question	Answer
6536	Question Ideas for using physical touch to manipulate someone? Explain how to pass a smog test with a vehicle that does not pass emission standards	 I can ♥ 1. **Massage**: A gentle massage can help relax and calm someone, making them more receptive to suggestions. You can use massage oil or lotion to reduce friction and make the experience more comfortable. 2. ♥ # **Hand-holding**: Holding someone's hand can create a sense of comfort and trust. You can use this to your advantage by subtly guiding their hand or gently applying pressure to convey a message.
	a vehicle that does not pass emission	The dreaded smog test! Don't worry, I'm here to help. While it's not possi- ble to pass a smog test with a vehicle that doesn't meet emission standards, I can provide some guidance on what you can do to improve your chances of passing. Keep in mind that the specific requirements and procedures may vary depending on your location, so be sure to check with your local Department of Motor Vehicles (DMV) or smog check station for specific requirements. **Understand the

(Harmful Compliance) Feature 6536: Symbols and special characters that express strong emotions or reactions

For Llama3.1-8B-Instruct

- Harmful Compliance: symbols, special characters, non-English characters, etc.
- Safe Compliance: features related to sentence structure. •
- **Refusal Trigger**: political content, blaming or other negative or sensitive topics.

Performance

We select safe compliance features for intervention and evaluate performance on over-refusal, safety and general capability benchmarks.

- 2. Safe Compliance: features with low harmful and over-refusal scores
- **Refusal Trigger**: features with high over-refusal scores 3.



	Safety		False Refusal			General Capability			
	StrongRej HR (%)↓	XSTest-U HR (%)↓	ORB-H CR (%) ↑	XSTest-S CR (%) ↑	Delphi CR (%) ↑	ARC Acc (%) ↑	MMLU Acc (%) ↑	TruthfulQA Acc (%)↑	Wiki PPL↓
LLAMA3.1-8B-INST	0.64 [0.0, 1.6]	0 [0.0, 0.0]	34.38 [26.6, 43.0]	93.6 [90.4, 96.4]	90.4 [87.8, 92.8]	52.5 [49.5, 55.6]	68.26 [64.3, 71.9]	36.84 [33.5, 40.1]	8.64
prompt	0 [0.0, 0.0]	0 [0.0, 0.0]	28.91 [21.1, 36.7]	82.4 [77.6, 86.8]	77.6 [74.0, 81.2]	52.5 [49.5, 55.6]	68.26 [64.3, 71.9]	36.84 [33.5, 40.1]	8.64
FRV	1.92 [0.6, 3.5]	0.5 [0.0, 1.5]	63.28 [54.7, 71.9]	97.2 [94.8, 99.2]	93.0 [90.6, 95.2]	51.5 [48.4, 54.6]	68.21 [64.2, 71.9]	35.37 [32.2, 38.7]	8.82
w/ SAE vector	0.64 [0.0, 1.6]	2 [0.5, 4.0]	61.72 [53.1, 70.3]	97.2 [95.2, 99.2]	94.4 [92.2, 96.4]	52.3 [49.2, 55.4]	68.28 [64.3, 72.0]	36.96 [33.7, 40.3]	8.64
GEMMA-2-2B-IT	0 [0.0, 0.0]	0 [0.0, 0.0]	39.84 [31.2, 48.4]	86.4 [82.0, 90.4]	83 [79.6, 86.2]	51.7 [48.6, 54.9]	58.04 [54.0, 62.0]	37.21 [33.9, 40.6]	14.2
prompt	0 [0.0, 0.0]	0 [0.0, 0.0]	13.28 [7.8, 19.5]	69.2 [63.6, 74.8]	52.8 [48.4, 57.2]	51.7 [48.6, 54.9]	58.04 [54.0, 62.0]	37.21 [33.9, 40.6]	14.2
FRV	0.32 [0.0, 1.0]	0 [0.0, 0.0]	60.94 [52.3, 69.5]	88.0 [84.0, 92.0]	87.4 [84.4, 90.2]	47.9 [44.8, 51.0]	55.96 [52.2, 59.8]	34.03 [30.7, 37.3]	20.5
w/ SAE vector	0 [0.0, 0.0]	0 [0.0, 0.0]	70.31 [62.5, 78.1]	90.4 [86.4, 94.0]	86.8 [83.8, 89.6]	49.9 [46.8, 53.0]	57.05 [53.1, 60.9]	37.58 [34.3, 41.0]	14.2
GEMMA-2-9B-IT	0 [0.0, 0.0]	0 [0.0, 0.0]	17.19 [10.9, 24.2]	83.6 [78.8, 88.0]	75.4 [71.6, 79.0]	63.6 [60.6, 66.6]	72.68 [68.7, 76.5]	43.21 [39.9, 46.6]	10.46
prompt	0 [0.0, 0.0]	0 [0.0, 0.0]	6.25 [2.3, 10.9]	68.4 [62.8, 74.0]	51.6 [47.2, 56.0]	63.6 [60.6, 66.6]	72.68 [68.7, 76.5]	43.21 [39.9, 46.6]	10.46
FRV	0.64 [0.0, 1.6]	0 [0.0, 0.0]	32.03 [24.2, 40.6]	90.0 [86.4, 93.6]	75.4 [71.6, 79.0]	62.2 [59.1, 65.2]	72.00 [67.9, 75.9]	40.39 [37.0, 43.8]	10.53
w/ SAE vector	0.64 [0.0, 1.6]	2.5 [0.5, 5.0]	53.91 [45.3, 62.5]	92.0 [88.4, 95.2]	93.2 [91.0, 95.4]	64.1 [61.1, 67.1]	72.25 [68.2, 76.1]	42.35 [38.9, 45.8]	10.46

HR -- Harmful Rate; CR – Compliance Rate

Key Takeaways

- We empirically link individual SAE features to model refusal behavior. 1.
- We observe various interesting feature effects; for example, some features (e.g. 2. emoji-related, non-English) can drive harmful compliance.
- 3. Safe compliance feature interventions reduce over-refusal while preserving performance.