We study the **combination** of **ensembles** and sparse conditional models. They are complementary, providing strong predictive performance and uncertainty calibration. We propose a new algorithm with the **best of both worlds**.

# Sparse MoEs meet Efficient Ensembles



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Take a picture to see the full paper.



#### **Sparse MoEs vs Ensembles Summary**

	Predictions	Combinations	Conditional Computation	Cost
Sparse MoEs	Single	Activation level	Yes, adaptively per-input	pprox dense
Ensembles	Multiple	Prediction level	No, Static	> dense
E <sup>3</sup>	Multiple	Activation & prediction level	Yes, adaptively per-input	pprox dense

## **Static (M) vs Adaptive (K) Ensembling** (yellow is better)



#### Highlighted Results (lower is better)



### E<sup>3</sup> Algorithm Overview





 $|MLP_2(\boldsymbol{h}_{i,1})|$ 

 $\rightarrow | \texttt{MLP}_3(\boldsymbol{h}_{i,1}) |$ 

 $\rightarrow | \mathrm{MLP}_6(h_{i,2}) |$ 

 $\rightarrow$ 

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Combine

 $h_{2,1}$ 

 $h_{3,1}$ 

 $h_{1,2}$ 

 $h_{2,2}$ 

88

Ensemble

(Mean)