

# Teaching Neural Networks to Say 'I Don't Know'

## Comparing Non-Bayesian Out of Distribution Data Detection Methods

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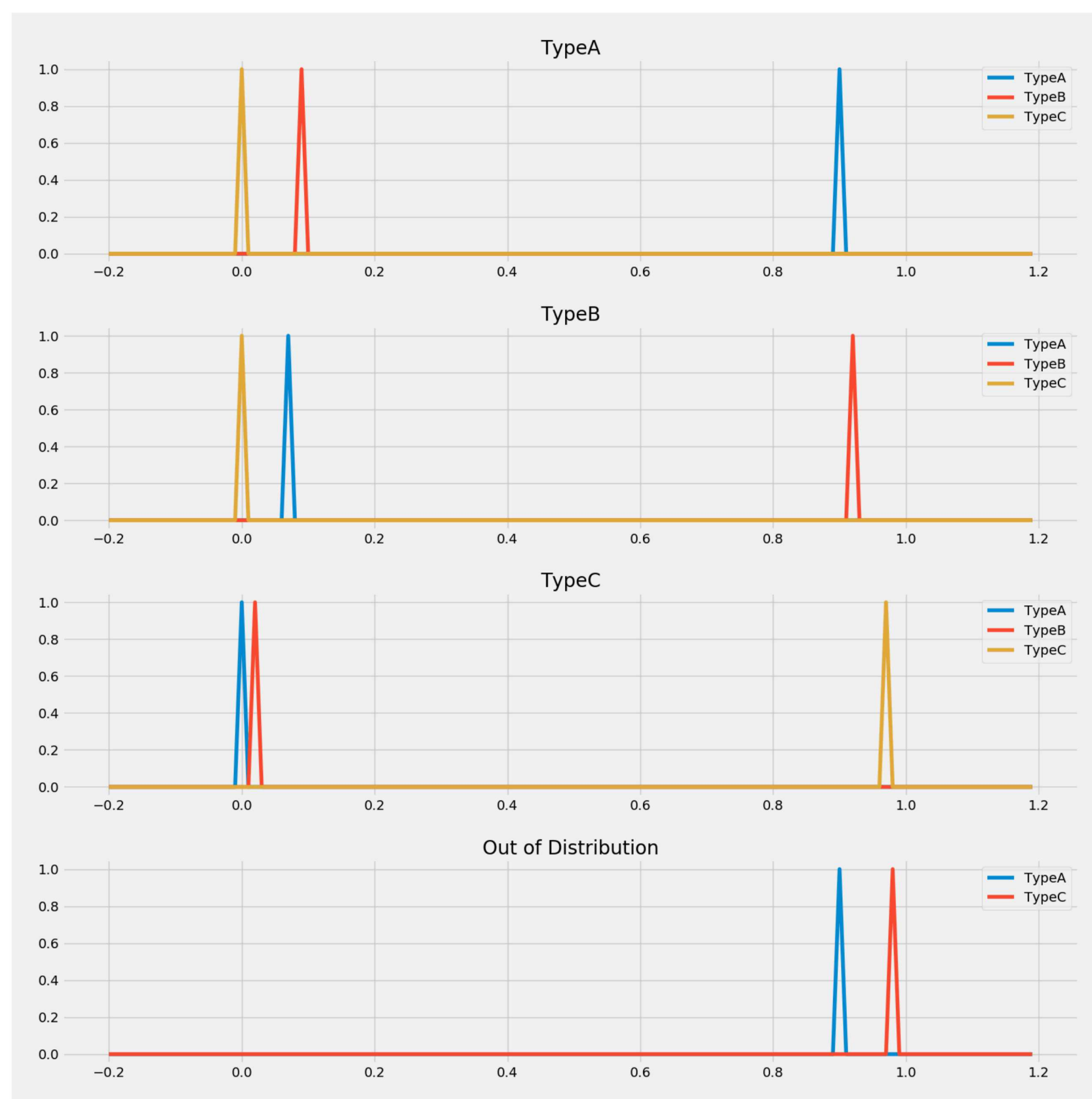
### Motivation

- Deep neural networks with softmax outputs make overconfident predictions, particularly for out of distribution data.
- As part of an Automatic Classification, Detection, and Geolocation (ACDG) project deliverable, non-Bayesian alternatives were compared using a proxy 1D dataset.

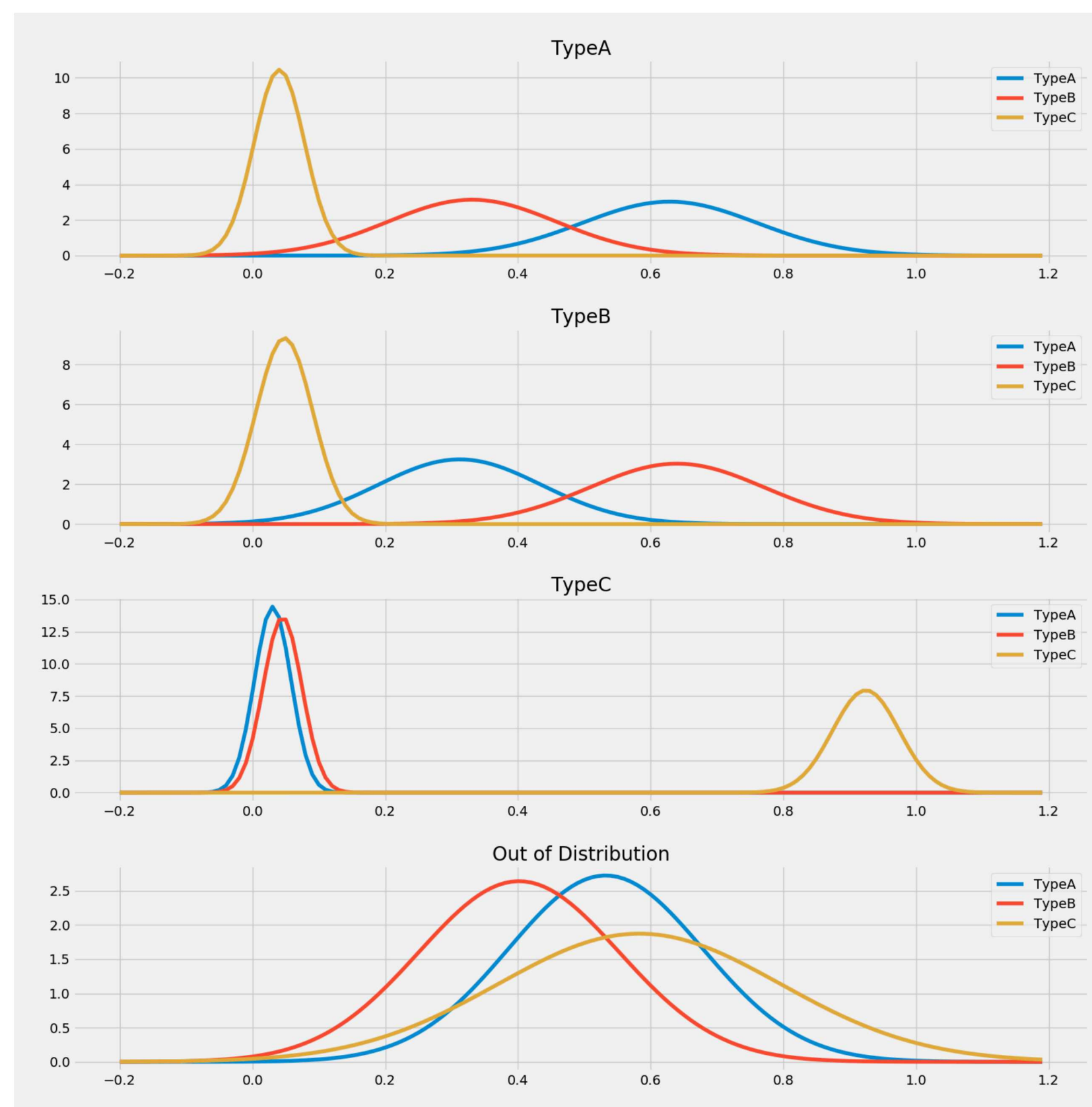
### Approach

- Trained and tested on three classes. Tested against an additional out of distribution class.
- Compared baseline convolutional model, an adaptive dropout method (concrete dropout), and a Dirichlet based loss method (evidentiary deep learning).

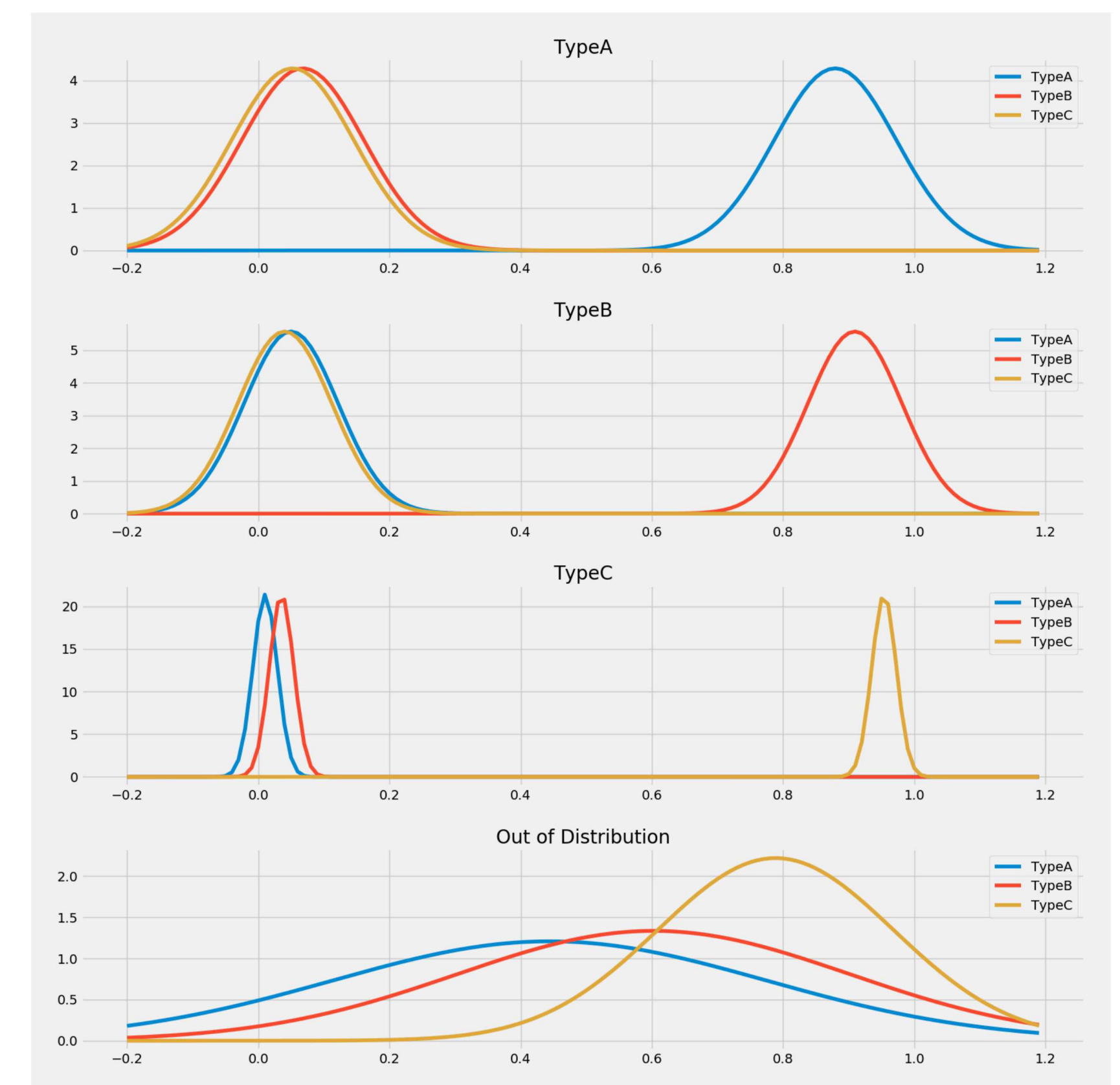
### Baseline Softmax



### Concrete Dropout

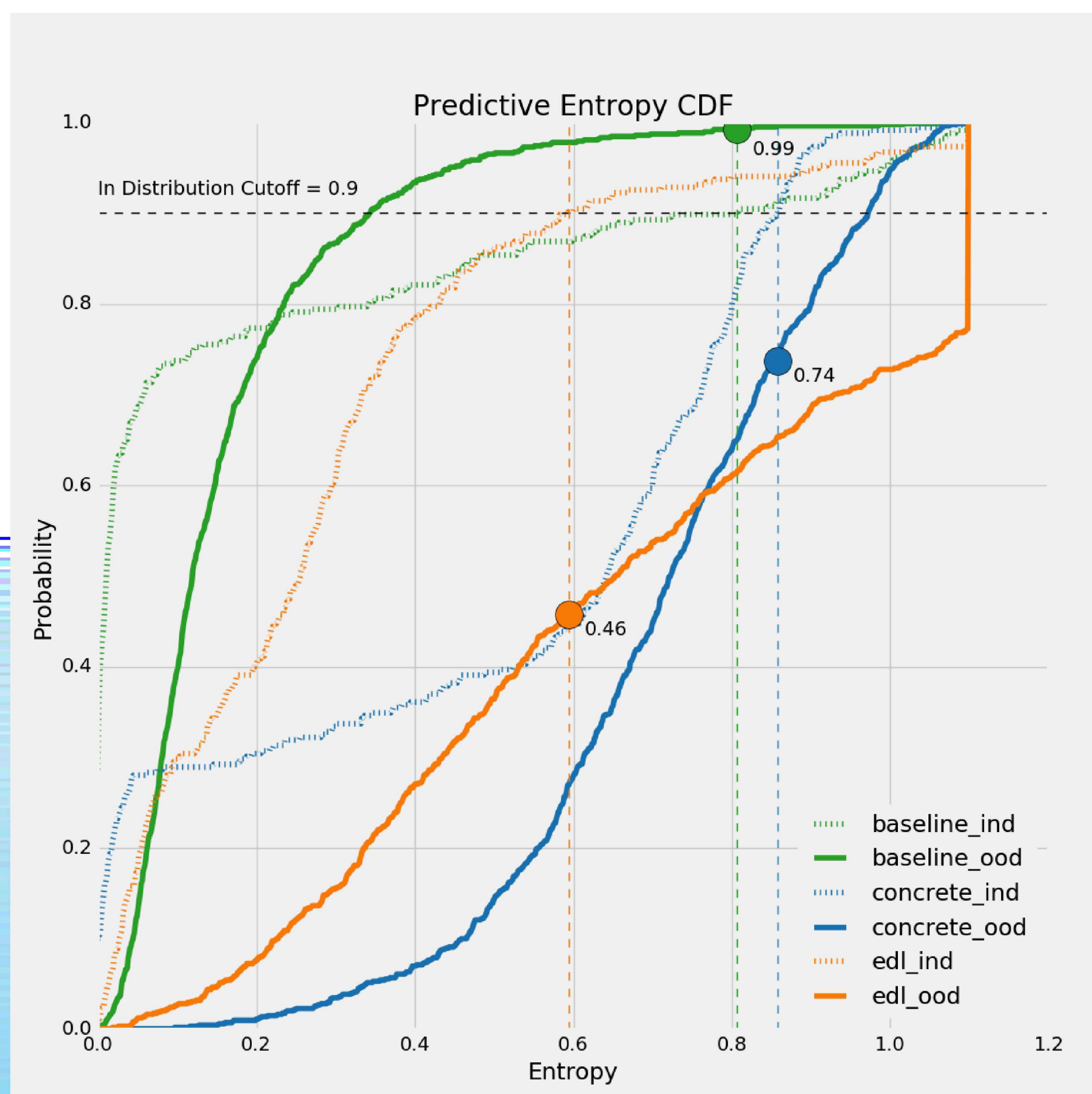


### Evidentiary Deep Learning



- Baseline softmax produces point estimate outputs. Concrete dropout and evidentiary deep learning methods are capable of producing uncertainty estimates.
- Baseline method produces much more confident output for all classes than either concrete dropout or evidentiary deep learning.

### Results



- Predictive entropy is a good measure of how well out of distribution data is detected. Higher entropy (more uncertainty) is better.

- The farther apart the in and out of distribution data entropies are, the more effective an entropy cutoff will be in detecting anomalies.

