

PyTorch Active Learning Library

Robert Munro, Human-in-the-Loop Machine Learning

An Active Learning library in Pytorch, implementing major techniques to sample items for annotation based on: model's confusion; gaps in the model's knowledge; and distributional properties of the data and target domain.

Machine Learning Knowledge Quadrant:

	Knowns	Unknowns
Known	Confident Model Predictions (Known Knowns)	Non-Confident Model Predictions (Known Unknowns)
×	Current Model State	Uncertainty Sampling
Unknown	Latent Information in Related Models (Unknown Knowns) Transfer Learning	Gaps in Model Knowledge (Unknown Unknowns) Diversity Sampling
_		

Examples:

The heatmaps show the differences between the Uncertainty Sampling methods on a three-label dataset. The methods use a probability distribution like: prob = torch.tensor([0.032, 0.643, 0.087, 0.236])



most conf = torch.max(prob) labs = prob.numel() numerator = (labs * (1 - most conf))denominator = (labs - 1)

Least Confidence: *difference*

between the most confident

n (1 -
$$P_{\theta}(y_{1}^{*} | \mathbf{x}))$$

<u>n</u> - 1



prediction and 100% confidence

Margin of Confidence:

difference between the top two most confident predictions

1 - (
$$P_{\theta}(y_{1}^{*} | \mathbf{x}) - P_{\theta}(y_{2}^{*} | \mathbf{x}))$$

Ratio of Confidence: ratio between the top two most confident predictions

$$\frac{P_{\theta}(y_{2}^{*} \mid \mathbf{x})}{P_{\theta}(y_{1}^{*} \mid \mathbf{x})}$$

Entropy: the difference between all predictions, as $-\sum_{y}$ defined by information theory

$$P_{\theta}(y \mid \mathbf{x}) \log_{2} P_{\theta}(y \mid \mathbf{x}))$$
$$\log_{2}(n)$$



least conf = numerator / denominator

prob, = torch.sort(prob, descending=True) difference = (prob.data[0] - prob.data[1])

margin_conf = 1 - difference

prob, _ = torch.sort(prob, descending=True) ratio_conf = (prob.data[1] / prob.data[0])

prbslogs = prob * torch.log2(prob) numerator = 0 - np.sum(prbslogs) denominator = math.log2(prob.numel())

entropy = numerator / denominator

Model-based Outliers: sampling for low activation in logits

Advanced Active Learning: combining multiple Active Learning techniques and incorporating Unsupervised Machine Learning, Domain Adaptation, and

and hidden layers to identify gaps in the model's knowledge





Code is open source & written to accompany:

Human-in-the-Loop Machine Learning, Robert Munro, Manning Publications. http://bit.ly/huml book