Uncertainty Sampling Cheatsheet

When a Supervised Machine Learning model makes a prediction, it often gives a confidence in that prediction. If the model is uncertain (low confidence), then human feedback can help. Getting human feedback when a model is uncertain is a type of *Active Learning* known as *Uncertainty Sampling*.

This cheatsheet has four common ways to calculate uncertainty, with examples, equations and python code.



The predictions are a probability distribution (**x**), meaning that every prediction is between 0 and 1 and the predictions add to 1. y_{1}^{*} is the most confident, y_{2}^{*} is the second most confident, etc. for **n** predicted labels.

This example can be expressed as a NumPy array: prob_dist = np.array([0.0321, 0.6439, 0.0871, 0.2369])

Least Confidence: difference between the most confident prediction and 100% confidence



```
most_conf = np.nanmax(prob_dist)
num_labels = prob_dist.size
numerator = (num_labels * (1 - most_conf))
denominator = (num_labels - 1)
```

least_conf = numerator / denominator

Margin of Confidence: difference between the top two most confident predictions



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



Entropy: difference between all predictions, as defined by information theory



Robert Munro. Human-in-the-Loop Machine Learning, *Manning Publications*. <u>http://bit.ly/huml_book</u> See the book for more details on each method and for more sophisticated problems like sequence models and semantic segmentation, plus other sampling strategies like Diversity Sampling. <u>robertmunro.com | @WWRob</u>