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Exp-normalize trick

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This trick is the very close cousin of the infamous log-sum-exp trick (scipy.misc.logsumexp (http://docs.scipy.org/doc/scipy/reference/generated/scipy.misc.logsumexp.html)), Supposed you'd like to evaluate a probability distribution $m{\pi}$ parametrized by a vector $m{x} \in \mathbb{R}^n$ as follows:

$$\pi_i = rac{\exp(x_i)}{\sum_{j=1}^n \exp(x_j)}$$

The exp-normalize trick leverages the following identity to avoid numerical overflow. For any $b\in\mathbb{R}$,

$$\pi_i = rac{\exp(x_i - b) \exp(b)}{\sum_{j=1}^n \exp(x_j - b) \exp(b)} = rac{\exp(x_i - b)}{\sum_{j=1}^n \exp(x_j - b)}$$

 $\pi_i = \frac{\exp(x_i - b) \exp(b)}{\sum_{j=1}^n \exp(x_j - b) \exp(b)} = \frac{\exp(x_i - b)}{\sum_{j=1}^n \exp(x_j - b)}$ In other words, the π is shift-invariant. A reasonable choice is $b = \max_{i=1}^n x_i$. With this choice, overflow due to \exp is impossible—the largest number exponentiated after shifting is 0.

Exp-normalize v. log-sum-exp

If what you want to remain in log-space, that is, compute $\log(\pi)$, you should use logsumexp. However, if π is your goal, then exp-normalize trick is for you! Since it avoids additional calls to exp, which would be required if using log-sum-exp and more importantly exp-normalize is more numerically stable!

Log-sum-exp for computing the log-distibution

$$\log \pi_i = x_i - \operatorname{logsumexp}(\boldsymbol{x})$$

where

$$ext{logsumexp}(oldsymbol{x}) = b + ext{log} \sum_{j=1}^n \exp(x_j - b)$$

Typically with the same choice for b as above.

Numerically-stable sigmoid function

The sigmoid function can be computed with the exp-normalize trick in order to avoid numerical overflow. In the case of $\operatorname{sigmoid}(x)$, we have a distribution with unnormalized log probabilities [x,0], where we are only interested in the probability of the first event. From the exp-normalize identity, we know that the distributions [x,0] and [0,-x] are equivalent (to see why, plug in $b = \max(0, x)$). This is why sigmoid is often expressed in one of two equivalent ways:

$$sigmoid(x) = 1/(1 + exp(-x)) = exp(x)/(exp(x) + 1)$$

Interestingly, each version covers an extreme case: $x=\infty$ and $x=-\infty$, respectively. Below is some python code which implements the trick:

```
def sigmoid(x):
 "Numerically-stable sigmoid function."
 if x >= 0:
     z = exp(-x)
     return 1 / (1 + z)
 else:
     # if x is less than zero then z will be small, denom can't be
     # zero because it's 1+z.
     z = exp(x)
     return z / (1 + z)
```

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Tim Vieira — Thanks for the pointer!

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