CORNELL NYCTECH **Class-Balanced Loss Based on Effective Number of Samples** Yin Cui^{1,2}, Menglin Jia¹, Tsung-Yi Lin³, Yang Song⁴, Serge Belongie^{1,2} Google Al ¹Cornell University ²Cornell Tech ³Google Brain ⁴Alphabet Inc.

Motivation

The World is Long-Tailed • Distribution of images per category in iNaturalist 2017:



• The more data, the better. However, as the number of samples increases, the marginal benefit a model can extract from the data diminishes.



Effective Number of Samples Data Sampling as Random Covering • To measure data overlap, we associate each sample with a small neighboring region of unit volume 1. Assume the volume of all possible data is N. **Definition** (Effective Number). The *effective number* of Overlapped (r samples is the expected volume of samples. **Proposition** (Effective Number). $E_n = (1-\beta^n)/(1-\beta)$, () All possible data (N) where $\beta = (N-1)/N$. Previously sampled data **Implication** (Asymptotic Properties). $E_n = 1$ if $\beta = 0$ Newly sampled data (1) $(N = 1). E_n \to n \text{ as } \beta \to 1 \ (N \to \infty).$ **Class-Balanced Loss** Class-Balanced Softmax Loss: $CB_{\text{softmax}}(\mathbf{z}, y) = -\frac{\left(1 - \beta\right)}{1 - \beta^{n_y}} \log\left(\frac{\exp(z_y)}{\sum_{j=1}^C \exp(z_j)}\right)$ Class-Balanced Sigmoid Loss: $\beta = 0$ $\beta = 0.9$ 10^{-3} $\beta = 0.99$ $\beta = 0.999$ $CB_{sigmoid}(\mathbf{z}, y) = -$ **Class-Balanced Focal Loss:** $CB_{focal}(\mathbf{z}, y) = -\left[\frac{1-\beta}{1-\beta^{n_y}}\right]\sum_{i=1}^{C} (1-p_i^t)^{\gamma} \log(p_i^t) \qquad z_i^t = \begin{cases} z_i, & \text{if } i = y. \\ -z_i, & \text{otherwise.} \end{cases} \quad p_i^t = \text{sigmoid}(z_i^t)$ Dataset # Classes Dataset Name Imbalance Long-Tailed CIFAR-10 10.00 - 200.00 10 Long-Tailed CIFAR-100 10.00 - 200.00 100 435.44 5,089 iNaturalist 2017 500.00 8,142 iNaturalist 2018 1.78 1,000 ILSVRC 2012

• Imbalance: the number of training samples in the largest class divided by the smallest class.







Classification Error Rate of ResNet-32 on CIFAR

Dataset Name	Long-Tailed CIFAR-10						Long-Tailed CIFAR-100					
Imbalance	200	100	50	20	10	1	200	100	50	20	10	1
Softmax	34.32	29.64	25.19	17.77	13.61	6.61	65.16	61.68	56.15	48.86	44.29	29.07
Sigmoid	34.51	29.55	23.84	16.40	12.97	6.36	64.39	61.22	55.85	48.57	44.73	28.39
Focal ($\gamma = 0.5$)	36.00	29.77	23.28	17.11	13.19	6.75	65.00	61.31	55.88	48.90	44.30	28.55
Focal ($\gamma = 1.0$)	34.71	29.62	23.29	17.24	13.34	6.60	64.38	61.59	55.68	48.05	44.22	28.85
Focal ($\gamma = 2.0$)	35.12	30.41	23.48	16.77	13.68	6.61	65.25	61.61	56.30	48.98	45.00	28.52
Class-Balanced	31.11	25.43	20.73	15.64	12.51	6.36*	63.77	60.40	54.68	47.41	42.01	28.39*
Loss Type	SM	Focal	Focal	SM	SGM	SGM	Focal	Focal	SGM	Focal	Focal	SGM
eta	0.9999	0.9999	0.9999	0.9999	0.9999	-	0.9	0.9	0.99	0.99	0.999	-
γ	-	1.0	2.0	-	-	-	1.0	1.0	-	0.5	0.5	-
$\frac{1000}{2} = \frac{1000}{2} = 1$							Long-Tailed	ng-Tailed CIFAR-100 (Imbalance Factor = 50)				



Classification Error Rate on ImageNet and iNaturalist

				iNatura	list 2017	iNaturalist 2018		ILSVRC 2012		
Network	Loss	β	γ	Input Size	Top-1	Top-5	Top-1	Top-5	Top-1	Top-5
ResNet-50	Softmax	7 - 87	-	224×224	45.38	22.67	42.86	21.31	23.92	7.03
ResNet-101	Softmax	-	-	224×224	42.57	20.42	39.47	18.86	22.65	6.47
ResNet-152	Softmax	-	-	224×224	41.42	19.47	38.61	18.07	21.68	5.92
ResNet-50	CB Focal	0.999	0.5	224×224	41.92	20.92	38.88	18.97	22.71	6.72
ResNet-101	CB Focal	0.999	0.5	224×224	39.06	18.96	36.12	17.18	21.57	5.91
ResNet-152	CB Focal	0.999	0.5	224×224	38.06	18.42	35.21	16.34	20.87	5.61
ResNet-50	CB Focal	0.999	0.5	320×320	38.16	18.28	35.84	16.85	21.99	6.27
ResNet-101	CB Focal	0.999	0.5	320×320	34.96	15.90	32.02	14.27	20.25	5.34
ResNet-152	CB Focal	0.999	0.5	320×320	33.73	14.96	30.95	13.54	19.72	4.97







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Experiments





ResNet-50 Training Curves