## Measurements of Three-Level Hierarchical Structure in the Outliers in the Spectrum of Deepnet Hessians Appendix

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number of parameters
C
                                                number of classes
                                                training examples
n
                                                training examples per class
n_c
x_{i,c}
                                               i-th example in c-th class
                                               one hot vector corresponding to the c-th class
\theta \in \mathbb{R}^p
                                               concatenation of all the parameters
 f(x_{i,c};\theta) \in \mathbb{R}^C
                                               logits (predictions prior to softmax) of x_{i,c}
 f_{c'}(x_{i,c};\theta) \in \mathbb{R}
                                                c'-th logit of x_{i,c}
\frac{\partial f(x_{i,c};\theta)}{\partial \theta} \in \mathbb{R}^{C \times p}
\frac{\partial f_{c'}(x_{i,c};\theta)}{\partial \theta} \in \mathbb{R}^{p}
p(x_{i,c};\theta) \in \mathbb{R}^{C}
                                               logit derivatives of x_{i,c}
                                                c'-th logit derivative of x_{i,c}
                                                Softmax(f(x_{i,c};\theta))
p_{c'}(x_{i,c};\theta) \in \mathbb{R}
                                                c'-th entry of Softmax(f(x_{i,c};\theta))
                                                \sqrt{p_{c'}(x_{i,c};\theta)}(y_{c'}-p(x_{i,c};\theta))^T \frac{\partial f(x_{i,c};\theta)}{\partial \theta}
\delta_{i,c,c'} \in \mathbb{R}^p
\delta_{c,c'} \in \mathbb{R}^p
                                                \text{Ave}_i \left\{ \delta_{i,c,c'} \right\}
                                               Ave<sub>i</sub> \{(\delta_{i,c,c'} - \delta_{c,c'})(\delta_{i,c,c'} - \delta_{c,c'})^T\}
\Sigma_{c,c'} \in \mathbb{R}^{p \times p}
\delta_c \in \mathbb{R}^p
                                                Ave_{c'\neq c}\{\delta_{c,c'}\}
                                              Ave<sub>c'\neq c</sub> \{\(\delta_{c,c'} - \delta_c\)\)\(\delta_{c,c'} - \delta_c\)\(\delta_{c,c'} - \delta_c\)\(\delta_{c,c'} - \delta_c\)\)\(T) Ave<sub>c</sub> \{\delta_{c,c}\delta_c^T\}\((C - 1)\) Ave<sub>c</sub> \{\delta_c\delta_c^T\}\((C - 1)\) Ave<sub>c</sub> \{\Sigma_c\delta_c\}\)
\Sigma_c \in \mathbb{R}^{p \times p}
G_0 \in \mathbb{R}^{p \times p}
G_1 \in \mathbb{R}^{p \times p}
G_2 \in \mathbb{R}^{p \times p}
 G_3 \in \mathbb{R}^{p \times p}
                                                \frac{1}{C}\sum_{c,c'}\sum_{c,c'}
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Table 1: Summary of notations.

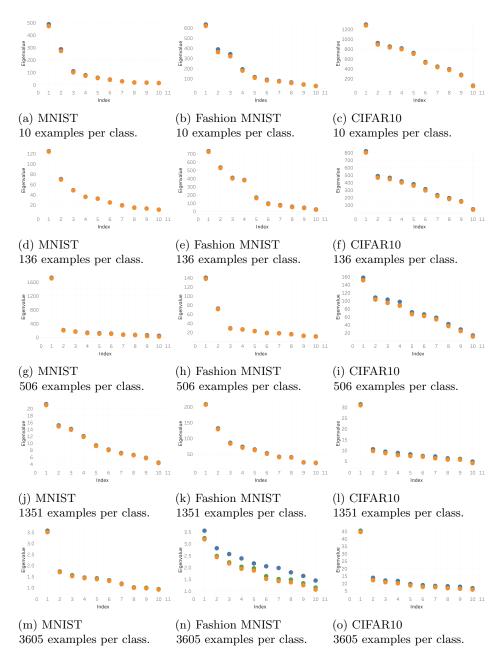


Figure 1: Scree plots of  $G_1$ ,  $G_{1+2}$  and G for the VGG11 architecture. Each column of panels corresponds to a different dataset, and each row to a different sample size. Each panel plots the top-C eigenvalues of  $G_1$  in orange,  $G_{1+2}$  in green and G in blue. The top eigenvalues in G – which correspond to the outliers in the approximated spectrum of G – were computed using the Lowrank Deflation procedure. For every  $1 \le c \le C$ , we have  $\lambda_c(G) \ge \lambda_c(G_{1+2}) \ge \lambda_c(G_1)$ . Moreover,  $\lambda_c(G_{1+2})$  and  $\lambda_c(G_1)$  are usually very close.