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Proof of Theorem

I. THE SOLUTION FOR (7)

The constrained optimization problem of the MGB is:

$$\mathbf{g}^*(t) = \min_{\mathbf{g}(t)} \frac{1}{2} ||\hat{\mathbf{g}}_k(t) - \mathbf{g}(t)||_2^2$$

$$s.t. \quad \tilde{\mathbf{G}}(t)^{\mathrm{T}} \mathbf{g}(t) \ge 0,$$
(1)

where k is the ordinal number of the complex language, $\hat{\mathbf{g}}_k(t) = w_k(t+1)\mathbf{g}_k(t)$ is the normalized gradient, $\mathbf{g}^*(t)$ is the balancing gradient, and $\tilde{\mathbf{G}}(t)$ is the set of the rest normalized gradients. We can obtain the lagrangian function of (1), which is shown as follows:

$$L(\mathbf{g}(t), \boldsymbol{\gamma}) = \frac{1}{2} \mathbf{g}(t)^{\mathrm{T}} \mathbf{g}(t) - \hat{\mathbf{g}}_{k}(t)^{\mathrm{T}} \mathbf{g}(t) - \boldsymbol{\gamma} \tilde{\mathbf{G}}(t)^{\mathrm{T}} \mathbf{g}(t),$$
(2)

where $L(\mathbf{g}(t), \gamma)$ is the lagrangian function and γ is the lagrange multiplier.

The lagrangian dual form of (1) is shown as follows:

$$f(\gamma) = \min_{\mathbf{g}(t)} L(\mathbf{g}(t), \gamma)$$
s.t. $\gamma \ge 0$, (3)

where $f(\gamma)$ is the lagrangian dual function.

We next find $\mathbf{g}^*(t)$ that minimizes (2) by setting the derivative of (2) to zero:

$$\frac{L(\partial \mathbf{g}(t), \boldsymbol{\gamma})}{\partial \mathbf{g}(t)} = \mathbf{g}(t)^{\mathrm{T}} - \hat{\mathbf{g}}_{k}(t)^{\mathrm{T}} - \boldsymbol{\gamma}\tilde{\mathbf{G}}(t)^{\mathrm{T}} = 0,$$

$$\Longrightarrow \mathbf{g}^{*}(t) = \hat{\mathbf{g}}_{k}(t) + \tilde{\mathbf{G}}(t)\boldsymbol{\gamma}^{\mathrm{T}}.$$
(4)

We further put $\mathbf{g}^*(t)$ in (3), and it can be written as:

$$f(\gamma) = \frac{1}{2} (\hat{\mathbf{g}}_{k}(t)^{\mathrm{T}} \hat{\mathbf{g}}_{k}(t) + 2\gamma \tilde{\mathbf{G}}(t)^{\mathrm{T}} \hat{\mathbf{g}}_{k}(t) + \gamma \tilde{\mathbf{G}}(t)^{\mathrm{T}} \tilde{\mathbf{G}}(t)\gamma^{\mathrm{T}})$$

$$- \hat{\mathbf{g}}_{k}(t)^{\mathrm{T}} \hat{\mathbf{g}}_{k}(t) - 2\gamma \tilde{\mathbf{G}}(t)^{\mathrm{T}} \hat{\mathbf{g}}_{k}(t) - \gamma \tilde{\mathbf{G}}(t)^{\mathrm{T}} \tilde{\mathbf{G}}(t)\gamma^{\mathrm{T}})$$

$$= -\frac{1}{2} \hat{\mathbf{g}}_{k}(t)^{\mathrm{T}} \hat{\mathbf{g}}_{k}(t) - \gamma \tilde{\mathbf{G}}(t)^{\mathrm{T}} \hat{\mathbf{g}}_{k}(t) - \frac{1}{2}\gamma \tilde{\mathbf{G}}(t)\gamma^{\mathrm{T}}$$
(5)

The solution $\gamma^* = \max_{\gamma;\gamma>0} f(\gamma)$ to the dual form is obtained by:

$$\frac{\partial f(\boldsymbol{\gamma})}{\partial \boldsymbol{\gamma}} = -\hat{\mathbf{g}}_k(t)^{\mathrm{T}} \tilde{\mathbf{G}}(t) - \boldsymbol{\gamma} \tilde{\mathbf{G}}(t)^{\mathrm{T}} \tilde{\mathbf{G}}(t) = 0,$$

$$\Longrightarrow \boldsymbol{\gamma}^* = -\frac{\hat{\mathbf{g}}(t)^{\mathrm{T}} \tilde{\mathbf{G}}(t)}{\tilde{\mathbf{G}}(t)^{\mathrm{T}} \tilde{\mathbf{G}}(t)}.$$
(6)

Finally, because γ^* must be grater than zeros, we obtain the balancing gradient by putting γ^* in (4):

$$\mathbf{g}^*(t) = \hat{\mathbf{g}}(t) - \tilde{\mathbf{G}}(t) \left[\frac{\hat{\mathbf{g}}(t)^{\mathrm{T}} \tilde{\mathbf{G}}(t)}{\tilde{\mathbf{G}}(t)^{\mathrm{T}} \tilde{\mathbf{G}}(t)} \right]^{\mathrm{T}}.$$
 (7)