Appendix 2: Convergence Analysis of ALM-ANAD Algorithm

The convergence of ALM has been thoroughly studied [22,23], so the convergence of the present ALM-ANAD algorithm relies on the convergence of ANAD algorithm. By extending Dai and Fletcher's proof [20], we establish the convergence results of ANAD algorithm.

Firstly, we impose the following assumptions on the function $\phi(x, y)$. Assumption1 There exists L > 0, such that at any given y, and for all x and \tilde{x} ,

$$\|\nabla_1 \phi(x, y) - \nabla_1 \phi(\tilde{x}, y)\|_2 \le L \|x - \tilde{x}\|_2.$$
(1)

Assumption 2 The function $\phi(x, y)$ is bounded below, i.e., there exists C > 0 such that for all (x, y),

$$\phi(x,y) \ge -C. \tag{2}$$

The convergence result of ANAD algorithm relies on the following lemma. For notational simplicity, let us define $\phi_k(\cdot) = \phi(\cdot, y_k)$ and $\nabla \phi_k(\cdot) = \nabla_1 \phi(\cdot, y_k)$. **Lemma 1** Let α_k satisfies condition (??) in ANAD algorithm, for all k > 0, we have

$$\alpha_k \ge \min\left\{1, \frac{2(1-\delta)\theta_1}{L} \frac{|\nabla \phi_k(x_k)^T d_k|}{\|d_k\|^2}\right\}$$
(3)

Proof. In fact, if $\alpha = 1$ satisfies condition (??), then we have $\alpha_k = 1$. Otherwise, there exists $\rho \in [\theta_1, \theta_2]$ for which $\frac{\alpha_k}{\rho} > 0$ fails to satisfy condition (??), it follows that

$$\phi_k(x_k + \frac{\alpha_k}{\rho}d_k) > \phi_r + \delta \frac{\alpha_k}{\rho} \nabla \phi_k(x_k)^T d_k.$$
(4)

On the other hand, by the mean-value theorem and Lipschitz condition, we have

$$\phi_k(x_k + \frac{\alpha_k}{\rho}d_k) - \phi_k(x_k) = \int_0^{\frac{\alpha_k}{\rho}} \langle \nabla \phi_k(x_k + td_k) - \nabla \phi_k(x_k), d_k \rangle dt + \frac{\alpha_k}{\rho} \nabla \phi_k(x_k)^T d_k$$

$$\leq \frac{L}{2} \left(\frac{\alpha_k}{\rho}\right)^2 ||d_k||^2 + \frac{\alpha_k}{\rho} \nabla \phi_k(x_k)^T d_k.$$
(5)

Combing the above two inequalities, we can find that (3) holds.

Then, the convergence theorem result of the ANAD algorithm can be described as follows.

Theorem 1 Let $\{(x_k, y_k)\}$ be a sequence generated by the ANAD algorithm. Then any accumulation point of the sequence $\{(x_k, y_k)\}$ is a stationary point, that is,

$$\begin{cases} \lim_{k \to \infty} \inf \|\nabla_1 \phi(x_k, y_k)\|_2 = 0, \\ \nabla_2 \phi(x_k, y_k) \ni 0. \end{cases}$$
(6)

Proof. We need to establish the two relationships given in equation (6). As the definition $y_k = \arg \min_{y} \phi(x_k, y)$ in the ANAD algorithm, $0 \in \nabla_2 \phi(x_k, y_k)$ always holds. Thus, it

suffices to show that the second equation holds in (6). By contradiction, for all k > 0, we have $\langle \nabla \phi_k(x_k), d_k \rangle \leq -\epsilon$ for some $\epsilon > 0$. In this case, we can conclude that there exists an infinite subsequence I such that for $k_i \in I$, the values of ϕ_r on iterations k_i are strictly monotonically decreasing. Let $\phi_r^{k_i}$ denote the value ϕ_r on iteration k_i , we have

$$\phi_{k_i}(x_{k_i+1}) \le \phi_r^{k_i} + \delta \alpha_{k_i} \left\langle \nabla \phi_{k_i}(x_{k_i}), d_{k_i} \right\rangle \le \phi_r^{k_0} + \delta \sum_{j=k_0, j \in I}^{k_i} \alpha_j \left\langle \nabla \phi_j(x_j), d_j \right\rangle \tag{7}$$

Furthermore, from the definition of d_k and Lemma 1, we have $\langle \nabla \phi_k(x_k), d_k \rangle \leq -\frac{1}{\rho_{\max}} \|d_k\|^2$ and $\alpha_k \geq \min\left\{1, \frac{2(1-\delta)\theta_1}{L\rho_{\max}}\right\}$. So, we can derive that

$$\phi_r^{k_0} - \phi_{k_i}(x_{k_i}) \ge \delta \sum_{j=k_0, j \in I}^{k_i} \alpha_j \left\langle \nabla \phi_j(x_j), d_j \right\rangle \ge \epsilon \delta \sum_{j=k_0, j \in I}^{k_i} \min\left\{ 1, \frac{2(1-\delta)\theta_1}{L\rho_{\max}} \right\}$$
(8)

Since $\phi(x, y)$ is bounded below, let $i \to \infty$, we get $\infty > \phi_r^{k_0} - \phi_{k_i}(x_{k_i}) \to \infty$. This is a contradiction. Hence, $\lim_{k\to\infty} \inf \|\nabla_1 \phi(x_k, y_k)\|_2 = 0$, which completes the proof.

Finally, as indicated before, the convergence of ALM-ANAD algorithm follows from that of ANAD algorithm.