

- In the last lecture, we studied the structure of an optimal solution  $\hat{z}$  to the problem

$$\hat{z} \in \arg \max_{z \in T^n} z^H C z \quad (\text{P})$$

In particular,  $\hat{z}$  satisfies both the first- and second-order optimality conditions. In this lecture, we show how the structural information on  $\hat{z}$  and the definition of GPM can be used to establish the convergence rate of the GPM. Specifically, our goal is to prove the following:

Theorem: Suppose that  $\|\Delta\| \leq \frac{n^{3/4}}{312}$ ,  $\|\Delta z^*\|_\infty \leq \frac{n}{24}$ ,  $\alpha \in [4, \frac{n}{\|\Delta\|}]$ , and the GPM is initialized by  $z^0 = v_c$ . Then, there exist  $\alpha_0$  and  $\lambda \in (0, 1)$  such that for any optimal solution  $\hat{z}$  to (P),

$$f(\hat{z}) - f(z^k) \leq (f(\hat{z}) - f(z^0)) \lambda^k,$$

$$d_2(z^k, \hat{z}) \leq \alpha \cdot (f(\hat{z}) - f(z^0))^{\frac{1}{2}} \cdot \lambda^{\frac{k}{2}}.$$

Here,  $f(z) = z^H C z$ .

- The proof of the theorem consists of two main parts. The first is to establish an error bound for (P). Recall that any  $z \in T^n$  that satisfies the first- and second-order optimality conditions also satisfies  $(\text{Diag}(|\tilde{C}z|) - \tilde{C})z = 0$ , where  $\tilde{C} = C + \frac{n}{\alpha} I$ . This motivates the following formulation of the error bound:

Proposition 1: Let  $\Sigma(z) = \text{Diag}(|\tilde{C}z|) - \tilde{C}$ ,  $p(z) = \|\Sigma(z)z\|_2$ . Under the assumptions of the theorem, for any  $z \in T^n$  satisfying  $d_2(z, z^*) \leq \sqrt{n}/2$  and any optimal solution  $\hat{z}$  to (P), we have

$$d_2(z, \hat{z}) \leq \frac{8}{n} p(z).$$

Before we proceed, several remarks are in order.

- (1) Recall that  $\hat{z}$  satisfies  $\|\hat{z}\|_2^2 = n$  and  $(z^*)^H C(z^*) \leq \hat{z}^H C \hat{z}$ , and hence  $d_2(\hat{z}, z^*) \leq \frac{4\|\Delta\|}{\sqrt{n}}$ . In particular, for  $\|\Delta\| \leq \frac{n}{8}$ , we have  $d_2(\hat{z}, z^*) \leq \frac{\sqrt{n}}{2}$ . Together with Proposition 1, we see that (P) has a unique optimal solution (up to a global phase).
- (2) We have previously shown that  $d_2(z^k, z^*) \leq \frac{\sqrt{n}}{2}$  whenever  $\|\Delta\| \leq \frac{n}{16}$ ; see the proof of Theorem 1 in the notes of Week 12. This shows that the error bound in Proposition 1 applies to the entire sequence of iterates  $\{z^k\}_{k \geq 0}$ .
- (3) Recall that if  $z \in T^n$  satisfies the first- and second-order optimality conditions, then  $p(z) = 0$ . Thus, all such points that also satisfy  $d_2(z, z^*) \leq \frac{\sqrt{n}}{2}$  are optimal for (P). In other words, the first- and second-order optimality conditions are sufficient for optimality.

The second part of the proof of the theorem is to establish the following properties of the GPM:

Proposition 2: Under the assumptions of the theorem, there exist  $\alpha_0, \alpha_1, \alpha_2 > 0$  such that for any optimal solution  $\hat{z}$  to (P), the following hold:

- (a) (Sufficient ascent)  $f(z^{k+1}) - f(z^k) \geq \alpha_0 \cdot \|z^{k+1} - z^k\|_2^2$
- (b) (cost-to-go estimate)  $f(\hat{z}) - f(z^k) \leq \alpha_1 \cdot d_2(z^k, \hat{z})^2$
- (c) (safeguard)  $f(z^k) \leq \alpha_2 \cdot \|z^{k+1} - z^k\|_2$ .

- Armed with Propositions 1 and 2, we are ready to prove the theorem:

Proof of the Theorem:

$$\begin{aligned}
 f(\hat{z}) - f(z^{k+1}) &= (f(\hat{z}) - f(z^k)) - (f(z^{k+1}) - f(z^k)) \\
 &\leq \alpha_1 \cdot d_2(z^k, \hat{z})^2 - (f(z^{k+1}) - f(z^k)) \quad (\text{cost-to-go}) \\
 &\leq \frac{64\alpha_1}{n^2} P(z^k)^2 - (f(z^{k+1}) - f(z^k)) \quad (\text{error bound}) \\
 &\leq \frac{64\alpha_1\alpha_2^2}{n^2} \|z^{k+1} - z^k\|_2^2 - (f(z^{k+1}) - f(z^k)) \quad (\text{safeguard}) \\
 &\leq \left( \frac{64\alpha_1\alpha_2^2}{\alpha_0 n^2} - 1 \right) (f(z^{k+1}) - f(\hat{z}) + f(\hat{z}) - f(z^k)). \quad (\text{sufficient ascent})
 \end{aligned}$$

Since  $f(\hat{z}) \geq f(z^k)$  for all  $k$ , we may assume without loss

that  $\alpha' = \frac{64\alpha_1\alpha_2^2}{\alpha_0 n^2} > 1$ . Hence,

$$f(\hat{z}) - f(z^{k+1}) \leq \frac{\alpha'-1}{\alpha'} (f(\hat{z}) - f(z^k)),$$

which yields  $f(\hat{z}) - f(z^{k+1}) \leq (f(\hat{z}) - f(z^0)) \lambda^k$ . Furthermore,

$$\begin{aligned}
 d_2(z^k, \hat{z})^2 &\leq \frac{64}{n^2} P(z^k)^2 \quad (\text{error bound}) \\
 &\leq \frac{64\alpha_2^2}{n^2} \|z^{k+1} - z^k\|_2^2 \quad (\text{safeguard}) \\
 &\leq \frac{64\alpha_2^2}{\alpha_0 n^2} (f(z^{k+1}) - f(z^k)) \quad (\text{sufficient ascent}) \\
 &\leq \frac{64\alpha_2^2}{\alpha_0 n^2} (f(\hat{z}) - f(z^k)) \\
 &\leq \frac{64\alpha_2^2}{\alpha_0 n^2} (f(\hat{z}) - f(z^0)) \lambda^k.
 \end{aligned}$$

This completes the proof.

- The above result once again demonstrates the power of the error bound-based convergence rate analysis framework.
- Let us now proceed to the proof of Proposition 1.

4

### Proof of Proposition 1

Observe that

$$P(z) = \|\Sigma(z)z\|_2 \geq \|\Sigma(\hat{z})z\|_2 - \|(\Sigma(z) - \Sigma(\hat{z}))z\|_2.$$

$$\text{Let } \hat{\theta} = \arg \min_{\theta \in [0, 2\pi)} \|z - e^{i\theta}\hat{z}\|_2, \quad \hat{\theta}^* = \arg \min_{\theta \in [0, 2\pi)} \|\hat{z} - e^{i\theta}z^*\|_2$$

and note that  $\tilde{C} = C + \frac{n}{\alpha}I = z^*(z^*)^H + \Delta + \frac{n}{\alpha}I$ . We first bound

$$\begin{aligned} & \|(\Sigma(z) - \Sigma(\hat{z}))z\|_2 = \|(\text{Diag}(|\tilde{C}z|) - \text{Diag}(|\tilde{C}\hat{z}|))z\|_2 \\ &= \left( \sum_{j=1}^n |(|\tilde{C}z|)_j - (|\tilde{C}\hat{z}|)_j| z_j|^2 \right)^{1/2} \\ &= \|\|\tilde{C}e^{-i\hat{\theta}}z\| - \|\tilde{C}\hat{z}\|\|_2 \quad (\text{since } |z_j|=1) \\ &\leq \|\tilde{C}(e^{-i\hat{\theta}}z - \hat{z})\|_2 \\ &\leq \|z^*(z^*)^H(e^{-i\hat{\theta}}z - \hat{z})\|_2 + \|\Delta(e^{-i\hat{\theta}}z - \hat{z})\|_2 + \frac{n}{\alpha} \|e^{-i\hat{\theta}}z - \hat{z}\|_2 \\ &\leq \sqrt{n} \cdot \|z^*(z^*)^H(e^{-i\hat{\theta}}z - \hat{z})\|_2 + (\|\Delta\| + \frac{n}{\alpha}) d_2(z, \hat{z}). \end{aligned} \tag{+}$$

By definition of  $\hat{\theta}$ , we have  $\hat{z}^H(e^{-i\hat{\theta}}z) = |\hat{z}^H z|$ . Thus,

$$d_2(z, \hat{z})^2 = \|e^{-i\hat{\theta}}z - \hat{z}\|_2^2 = 2(n - |\hat{z}^H z|). \tag{*}$$

Recalling that  $d_2(\hat{z}, z^*) \leq \frac{4\|\Delta\|}{\sqrt{n}}$ , we obtain

$$\begin{aligned} |(z^*)^H(e^{-i\hat{\theta}}z - \hat{z})| &\leq |(z^* - e^{-i\hat{\theta}}\hat{z})^H(e^{-i\hat{\theta}}z - \hat{z})| + |(e^{-i\hat{\theta}}\hat{z})^H(e^{-i\hat{\theta}}z - \hat{z})| \\ &\leq \|z^* - e^{-i\hat{\theta}}\hat{z}\|_2 \cdot \|e^{-i\hat{\theta}}z - \hat{z}\|_2 + |\hat{z}^H z| - n \\ &\leq \frac{4\|\Delta\|}{\sqrt{n}} \cdot d_2(z, \hat{z}) + \frac{1}{2} d_2(z, \hat{z})^2 \end{aligned} \tag{++}$$

It follows from (+) and (++) that

$$\|(\Sigma(z) - \Sigma(\hat{z}))z\|_2 \leq (5\|\Delta\| + \frac{n}{\alpha}) d_2(z, \hat{z}) + \frac{\sqrt{n}}{2} d_2(z, \hat{z})^2.$$

Next, we show that  $\Sigma(\hat{z}) \succ 0$  and  $\Sigma(\hat{z})$  is pd on the orthogonal complement of  $\text{Span}(\hat{z})$ . Indeed, we have  $\Sigma(\hat{z})\hat{z} = 0$  because  $\hat{z}$  satisfies both the first- and second-order optimality conditions.

Next, let  $\hat{u} = (I - \frac{1}{n}\hat{z}\hat{z}^H)(e^{-i\hat{\theta}}\hat{z} - \hat{z})$  be the projection of  $e^{-i\hat{\theta}}\hat{z} - \hat{z}$  onto the orthogonal complement of  $\text{span}(\hat{z})$ . Then, we have  $\hat{u}^H\hat{z} = 0$  and

$$\begin{aligned}\|\hat{u}\|_2 &\geq \|e^{i\hat{\theta}}\hat{z} - \hat{z}\|_2 - \frac{1}{n}\|\hat{z}\hat{z}^H(e^{-i\hat{\theta}}\hat{z} - \hat{z})\|_2 \\ &= d_2(z, \hat{z}) - \frac{1}{2\sqrt{n}}d_2(z, \hat{z})^2 \quad (\text{by } (*) \text{ and } \hat{z}^H(e^{-i\hat{\theta}}\hat{z}) = |\hat{z}^H\hat{z}|). \quad -(*)\end{aligned}$$

Also, by definition of  $\hat{z}^*$ , we have

$$(z^*)^H(e^{-i\hat{\theta}^*}\hat{z}) = |(z^*)^H\hat{z}| \quad \text{and} \quad |(z^*)^H\hat{z}| = n - \frac{1}{2}\|\hat{z} - e^{i\hat{\theta}^*}z^*\|_2^2. \quad -(**)$$

It follows that

$$\begin{aligned}\hat{u}^H \Sigma(\hat{z}) \hat{u} &= \hat{u}^H (\text{Diag}(|\hat{C}\hat{z}|) - \hat{C}) \hat{u} = \hat{u}^H (\text{Diag}(|C\hat{z}|) - C) \hat{u} \\ &\quad (\text{since } \hat{z} \text{ satisfies } |\langle \hat{C}\hat{z}, \cdot \rangle| = \langle \hat{C}\hat{z}, \frac{\hat{z}}{\|\hat{z}\|} \rangle = \langle C\hat{z}, \frac{\hat{z}}{\|\hat{z}\|} \rangle + \frac{n}{\alpha} = |\langle C\hat{z}, \cdot \rangle| + \frac{n}{\alpha};)\end{aligned}$$

see pp. 10-11 of the notes of Week 10)

$$\begin{aligned}&\geq \sum_{j=1}^n |\langle C\hat{z}, j \rangle| \cdot |\hat{u}_j|^2 - |(z^*)^H \hat{u}|^2 - \hat{u}^H \Delta \hat{u} \quad (\text{since } C = z^*(z^*)^H + \Delta) \\ &\geq (|(z^*)^H \hat{z}| - \|\Delta \hat{z}\|_\infty) \cdot \|\hat{u}\|_2^2 - |\hat{u}^H (z^* - e^{-i\hat{\theta}^*} \hat{z})|^2 - \|\Delta\| \cdot \|\hat{u}\|_2^2 \\ &\geq (n - \|\Delta \hat{z}\|_\infty - \|\Delta\| - \frac{3}{2}\|z^* - e^{-i\hat{\theta}^*} \hat{z}\|_2^2) \cdot \|\hat{u}\|_2^2 \quad (\text{by } (**)) \\ &\geq (n - \|\Delta \hat{z}\|_\infty - \|\Delta\| - \frac{24\|\Delta\|^2}{n}) \cdot \|\hat{u}\|_2^2 \quad (\text{since } d_2(z, \hat{z}^*) \leq \frac{4\|\Delta\|}{\sqrt{n}}).\end{aligned}$$

In particular, since  $\Sigma(\hat{z})\hat{z} = 0$ ,

$$\begin{aligned}\|\Sigma(\hat{z})z\|_2 &= \|\Sigma(\hat{z})(e^{-i\hat{\theta}}\hat{z} - \hat{z})\|_2 = \|\Sigma(\hat{z})\hat{u}\|_2 \\ &\geq (n - \|\Delta \hat{z}\|_\infty - \|\Delta\| - \frac{24\|\Delta\|^2}{n}) (d_2(z, \hat{z}) - \frac{1}{2\sqrt{n}}d_2(z, \hat{z})^2) \quad (\text{by } (*))\end{aligned}$$

Since  $d_2(z, \hat{z}^*) \leq \frac{\sqrt{n}}{2}$  by assumption, we have

$$d_2(z, \hat{z}) \leq d_2(z, \hat{z}^*) + d_2(\hat{z}, \hat{z}^*) \leq \frac{\sqrt{n}}{2} + \frac{4\|\Delta\|}{\sqrt{n}},$$

which implies

$$d_2(z, \hat{z})^2 \leq \left( \frac{\sqrt{n}}{2} + \frac{4\|\Delta\|}{\sqrt{n}} \right) d_2(z, \hat{z}).$$

Moreover,

$$\begin{aligned}\|\Delta \hat{z}\|_\infty &\leq \|\Delta z^*\|_\infty + \|\Delta(e^{i\hat{\theta}^*} \hat{z} - z^*)\|_\infty \\ &\leq \|\Delta z^*\|_\infty + \|\Delta\| \cdot d_2(\hat{z}, z^*) \\ &\leq \|\Delta z^*\|_\infty + \frac{4\|\Delta\|^2}{\sqrt{n}}.\end{aligned}$$

Putting everything together and making the assumptions, we have

$$p(z) \geq \|\Sigma(z)z\|_2 - \|\Sigma(z) - \Sigma(\hat{z})\|_2 z \geq \frac{n}{8} d_2(z, \hat{z}),$$

as desired.

- Next, let us turn to the proof of Proposition 2.

### Proof of Proposition 2

(a) Recall that  $\tilde{C} = C + \frac{n}{\alpha} I$ . We compute

$$\begin{aligned}f(z^{k+1}) - f(z^k) &= (z^{k+1} - z^k)^H \tilde{C} (z^{k+1} - z^k) - 2(z^k)^H \tilde{C} z^k \\ &\quad + 2 \operatorname{Re}[(z^{k+1})^H \tilde{C} z^k].\end{aligned}$$

By definition of  $z^{k+1}$ , we have  $(z^{k+1})^H \tilde{C} z^k = (\frac{\tilde{C} z^k}{\|\tilde{C} z^k\|})^H \tilde{C} z^k$ ,

which, as one would observe, is a real number. Moreover,

$$z^{k+1} \in \underset{z \in \mathbb{T}^n}{\operatorname{argmax}} \operatorname{Re}(z^H \tilde{C} z^k),$$

which implies that  $\operatorname{Re}[(z^{k+1})^H \tilde{C} z^k] \geq (z^k)^H \tilde{C} z^k$ . Hence,

$$f(z^{k+1}) - f(z^k) \geq (z^{k+1} - z^k)^H \tilde{C} (z^{k+1} - z^k) \geq \alpha_0 \cdot \|z^{k+1} - z^k\|_2^2$$

with  $\alpha_0 = \lambda_{\min}(C + \frac{n}{\alpha} I) > 0$  by assumption on  $C$ .

(b) Let  $\hat{\theta}_k = \underset{\theta \in [0, 2\pi)}{\operatorname{argmin}} \|z^k - e^{i\theta} \hat{z}\|_2$ . Then,

$$f(\hat{z}) - f(z^k) = \hat{z}^H \tilde{C} \hat{z} - (z^k)^H \tilde{C} z^k$$

$$= \|\tilde{C} \hat{z}\|_1 - (z^k)^H \tilde{C} z^k$$

(by Proposition 5 of Week 10's notes)

$$= (z^k)^H [\operatorname{Diag}(\|\tilde{C} \hat{z}\|_1) - \tilde{C}] z^k$$

(since  $(z^k)^H \operatorname{Diag}(\|\tilde{C} \hat{z}\|_1) z^k = \sum_i \|\tilde{C} \hat{z}\|_i = \|\tilde{C} \hat{z}\|_1$ )

$$\begin{aligned}
&= (e^{-i\hat{\theta}_k} z^k - \hat{z})^H (\text{Diag}(|\hat{C}\hat{z}|) - \tilde{C})(e^{-i\hat{\theta}_k} z^k - \hat{z}) \\
&\quad (\text{since } [\text{Diag}(|\hat{C}\hat{z}|) - \tilde{C}] \hat{z} = 0) \\
&\leq (\|\tilde{C}\| + \|\tilde{C}\hat{z}\|_\infty) \cdot d_2(z^k, \hat{z})^2
\end{aligned}$$

We bound

$$\begin{aligned}
\|\tilde{C}\| + \|\tilde{C}\hat{z}\|_\infty &\leq \|C\| + \|C\hat{z}\|_\infty + \frac{2n}{\alpha} \\
&\leq \|z^*(z^*)^H\| + \|\Delta\| + \|z^*(z^*)^H \hat{z}\|_\infty + \|\Delta\hat{z}\|_\infty + \frac{2n}{\alpha}.
\end{aligned}$$

Observe that  $\|z^*(z^*)^H\| = n$ ,  $\|z^*(z^*)^H \hat{z}\|_\infty = |(z^*)^H \hat{z}| \cdot \|z^*\|_\infty \leq n$  and

$$\begin{aligned}
\|\Delta\hat{z}\|_\infty &\leq \|\Delta z^*\|_\infty + \|\Delta(e^{-i\hat{\theta}^*} \hat{z} - z^*)\|_\infty \\
&\leq \|\Delta z^*\|_\infty + \|\Delta\| \cdot d_2(\hat{z}, z^*) \leq \|\Delta z^*\|_\infty + \frac{4\|\Delta\|^2}{\sqrt{n}}.
\end{aligned}$$

Thus,

$$\|\tilde{C}\| + \|\tilde{C}\hat{z}\|_\infty \leq 2n + \|\Delta\| + \|\Delta z^*\|_\infty + \frac{4\|\Delta\|^2}{\sqrt{n}} + \frac{2n}{\alpha} < 3n,$$

where the last inequality is due to our assumptions. Hence, we have

$$f(\hat{z}) - f(z^k) \leq \alpha_1 \cdot d_2(z^k, \hat{z})^2$$

for some  $\alpha_1 \in (0, 3n)$ .

(c) By definition of  $z^{k+1}$ , we have

$$\text{Diag}(|\hat{C}z^k|)(z^{k+1} - z^k) = [\tilde{C} - \text{Diag}(|\hat{C}z^k|)]z^k.$$

Hence,

$$P(z^k) = \|\text{Diag}(|\hat{C}z^k|)(z^{k+1} - z^k)\|_2 \leq \|\tilde{C}z^k\|_\infty \cdot \|z^{k+1} - z^k\|_2.$$

Recall that  $d_2(z^k, z^*) \leq \frac{\sqrt{n}}{2}$  (See Remark 2 on p.2). Let

$$\theta_k^* = \underset{\theta \in [0, 2\pi)}{\operatorname{argmin}} \|z^k - e^{i\theta} z^*\|_2. \text{ Then, by our assumptions,}$$

$$\begin{aligned}
\|\tilde{C}z^k\|_\infty &\leq \|z^*(z^*)^H z^k\|_\infty + \|\Delta z^k\|_\infty + \frac{n}{\alpha} \\
&\leq |(z^*)^H z^k| + \|\Delta(e^{-i\theta_k^*} z^k - z^*)\|_\infty + \|\Delta z^k\|_\infty + \frac{n}{\alpha} \\
&\leq n + \|\Delta\| \cdot d_2(z^k, z^*) + \|\Delta z^k\|_\infty + \frac{n}{\alpha} \\
&< \frac{3}{2} n^{5/4}.
\end{aligned}$$

Hence,  $P(z^k) \leq \alpha_2 \cdot \|z^{k+1} - z^k\|_2$  for some  $\alpha_2 \in (0, \frac{3}{2} n^{5/4})$ .