# A Generalization of Lagrange Interpolation Theorem* 

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## 1. Introduction

Let $f$ be an analytic function on the unit closed disk $\bar{D}=\{z \in \mathbb{C}:|z| \leqslant 1\}$. In notation $f \in A(\bar{D})$. Let $z_{1}, z_{2}, \ldots$, be a sequence of points in $\bar{D}$ satisfying

$$
\begin{equation*}
z_{n} \text { are all distinct and } \lim z_{n}=0 . \tag{1.1}
\end{equation*}
$$

Define

$$
\begin{align*}
& P_{n-1}\left(z ; f ; z_{1}, z_{2}, \ldots, z_{n}\right) \\
& \quad=\sum_{k=1}^{n} f\left(z_{k}\right)\left(\prod_{\substack{i=1 \\
i \neq k}}^{n}\left(z-z_{i}\right)\right) /\left(\prod_{\substack{i=1 \\
i \neq k}}^{n}\left(z_{k}-z_{i}\right)\right) \tag{1.2}
\end{align*}
$$

to be the Lagrange interpolation polynomial of $f$ at $z_{1}, z_{2}, \ldots, z_{n}$. Then it is well known [3] that

$$
\begin{equation*}
\lim _{n}\left\|P_{n-1}-f\right\|_{\infty}=0 \tag{1.3}
\end{equation*}
$$

Here and in the following $\|\cdot\|_{\infty}$ will always be the sup norm on $\bar{D}$.
From (1.3) it follows that, given $\left\{f\left(z_{k}\right)\right\}$, we can recover $f(z)$ as a limit. It is natural to study a similar setting where, however, errors are allowed. More precisely, can we reconstruct $f$ if we are given sequentially the data

$$
\begin{equation*}
f\left(z_{1}\right)+e_{1}, f\left(z_{2}\right)+e_{2}, f\left(z_{3}\right)+e_{3}, \ldots, \tag{1.4}
\end{equation*}
$$

where the errors $e_{k}$ are assumed to be independent and identically distributed random vectors on a probability space with mean $(0,0)$ and finite variance.

[^0]By (1.3) it is natural to take

$$
\begin{equation*}
\sum_{k=1}^{n}\left(f\left(z_{k}\right)+e_{k}\right)\left(\prod_{\substack{i=1 \\ i \neq k}}^{n}\left(z-z_{i}\right)\right) /\left(\prod_{\substack{i=1 \\ i \neq k}}^{n}\left(z_{k}-z_{i}\right)\right) \tag{1.5}
\end{equation*}
$$

as our estimator. Unfortunately, this approach does not work [1]. This leads to the consideration of the following problem

$$
\begin{equation*}
\min _{g \in S_{m}} \sum_{k=1}^{n}\left|g\left(z_{k}\right)-\left(f\left(z_{k}\right)+e_{k}\right)\right|^{2} \tag{1.6}
\end{equation*}
$$

where $m \leqslant n-1$ and $S_{m}$ is the set of all polynomials with order $\leqslant m$. Note that (1.5) is the solution to problem (1.6) if $m=n-1$. The reason why (1.5) is not suitable as an estimator to $f$ is because the corresponding minimization space $S_{n-1}$ is too big. According to Grenander's sieve method [5,6], this difficulty can often be overcome if the minimum in (1.6) is taken over a sequence of smaller subspaces $S_{m}, m \leqslant n-1$, which grows slowly to $A(\bar{D})$ as $n \rightarrow \infty$.

Problem (1.6) can be regarded as a minimization problem over $2(m+1)$ real variables. By using elementary calculus it is not hard to show that (1.6) has a unique solution

$$
\begin{equation*}
\left(F_{n}+E_{n}\right) \cdot\left(A_{m+1, n}\right)^{*} \cdot C_{m+1}^{-1} \cdot W_{m+1} \tag{1.7}
\end{equation*}
$$

where $\quad F_{n}=\left(f\left(z_{1}\right), f\left(z_{2}\right), \ldots, f\left(z_{n}\right)\right), \quad E_{n}=\left(e_{1}, e_{2}, \ldots, e_{n}\right), \quad W_{m+1} \quad$ is the $(m+1) \times 1$ column vector $\left(1, z, \ldots, z^{m}\right)$,

$$
A_{m+1, n}=\left(z_{j}^{i-1}\right)_{(m+1) \times n}
$$

is the $(m+1) \times n$ matrix with $\left(z_{j}\right)^{i-1}$ as its $(i, j)$ th element, and

$$
C_{m+1}=\left(A_{m+1, n}\right) \cdot\left(A_{m+1, n}\right)^{*}
$$

Here * means complex transposition. Note that $C_{m+1}^{-1}$ exists, because $C_{m+1}$ is positive definite by (1.1) and the assumption that $m \leqslant n-1$. By letting all $e_{n}=0$, it is clear that the deterministic part

$$
\begin{equation*}
F_{n, m}(z)=F_{n} \cdot\left(A_{m+1, n}\right)^{*} \cdot C_{m+1}^{-1} \cdot W_{m+1} \tag{1.8}
\end{equation*}
$$

of (1.7) is the solution to the problem

$$
\begin{equation*}
\min _{g \in S_{m}} \sum_{k=1}^{n}\left|g\left(z_{k}\right)-f\left(z_{k}\right)\right|^{2}, \quad m \leqslant n-1 \tag{1.9}
\end{equation*}
$$

and $F_{n, m}(z)$ should converge to $f(z)$ in some way if the sieve method is applicable to the estimation problem (1.4).

The main purpose of this paper is to verify this claim. In fact, $F_{n, m}(z)$ has the following representation.

Theorem 1. Let $z_{1}, z_{2}, \ldots, z_{n}$ be $n$ distinct points and $f$ any function. Then the unique solution $F_{n, m}(z)$ to problem (1.9) satisfies

$$
\begin{align*}
& F_{n, m}(z) \\
& \qquad= \frac{\sum_{1 \leqslant i_{1}<i_{2}<\cdots<i_{m+1} \leqslant n}\left(\prod_{1 \leqslant j<k \leqslant m+1}\left|z_{i_{j}}-z_{i_{k}}\right|^{2}\right) P_{m}\left(z ; f ; z_{i_{1}}, z_{i_{2}}, \ldots, z_{i_{m+1}}\right)}{\sum_{1 \leqslant i_{1}<i_{2}<\cdots<i_{m+1} \leqslant n}\left(\prod_{1 \leqslant j<k \leqslant m+1}\left|z_{i_{j}}-z_{i_{k}}\right|^{2}\right)}, \tag{1.10}
\end{align*}
$$

where $P_{m}\left(z ; f ; z_{i_{1}}, z_{i_{2}}, \ldots, z_{i_{m+1}}\right)$, as in (1.2), is the $m$ th-order Lagrange interpolation polynomial of $f$ at $z_{i_{1}}, z_{i_{2}}, \ldots, z_{i_{m+1}}$.

In view of (1.3) it is not hard to believe the following result.
Theorem 2. Let $f \in A(\bar{D})$ and $\left\{z_{n}\right\}$ satisfy (1.1). Then

$$
\lim \left\|F_{n, m}(z)-f(z)\right\|_{\infty}=0
$$

as both $n, m$ tend to $\infty$ with $m \leqslant n-1$.
Finally, let us say a few words on the estimation problem (1.4). In order that there is a restoration algorithm working for every $f \in A(\bar{D})$, it is necessary and sufficient that $\left\{z_{n}\right\}$ has infinite convergence exponent, i.e.,

$$
\sum_{n=1}^{\infty}\left|z_{n}\right|^{p}=\infty \quad \text { for every } \quad p>0
$$

This result will appear in [2]. There, how $m=m(n)$ tends to $\infty$ will depend on $\left\{z_{1}, z_{2}, \ldots, z_{n}\right\}$ and is not explicitly known.

## 2. Proof of Theorem 1

Define the symmetric sums $T_{k}\left(z_{1}, z_{2}, \ldots, z_{u}\right)$ as follows

$$
\begin{equation*}
\prod_{i=1}^{u}\left(z-z_{i}\right)=\sum_{k=0}^{u}(-1)^{k} T_{k}\left(z_{1}, z_{2}, \ldots, z_{k}\right) z^{u-k} \tag{2.1}
\end{equation*}
$$

We shall first prove a lemma on $C_{m+1}$.
Lemma. Let $C_{m+1}^{-1}=(d(i, j))_{(m+1) \times(m+1)}$. Then

$$
\begin{equation*}
\operatorname{det} C_{m+1}=\sum_{1 \leqslant i_{1}<i_{2}<\cdots<i_{m+1} \leqslant n}\left(\prod_{1 \leqslant j<k \leqslant m+1}\left|z_{i_{j}}-z_{i_{k}}\right|^{2}\right) \tag{2.2}
\end{equation*}
$$

and for $0 \leqslant r, s \leqslant m$,

$$
\begin{align*}
d(r+1, s+1)= & (-1)^{r+s} \sum_{1 \leqslant i_{1}<i_{2}<\cdots<i_{m} \leqslant n} T_{m-s}\left(z_{i_{1}}, z_{i_{2}}, \ldots, z_{i_{m}}\right) \\
& \times T_{m-r}\left(\bar{z}_{i_{1}}, \bar{z}_{i_{2}}, \ldots, \bar{z}_{i_{m}}\right) \\
& \times\left(\prod_{1 \leqslant j<k \leqslant m}\left|z_{i_{j}}-z_{i_{k}}\right|^{2}\right) /\left(\operatorname{det} C_{m+1}\right) . \tag{2.3}
\end{align*}
$$

Proof. Since $C_{m+1}=\left(A_{m+1, n}\right) \cdot\left(A_{m+1, n}\right)^{*}$, the Binet-Cauchy formula [4] tells us that det $C_{m+1}$ and $d(r+1, s+1)$ can be expressed as sums of products of the corresponding minors of $A_{m+1, n}$ and $\left(A_{m+1, n}\right)^{*}$. Equalities (2.2) and (2.3) will then be proved if one notes that

$$
\operatorname{det} A_{n, n}=\prod_{1 \leqslant j<i \leqslant n}\left(z_{i}-z_{j}\right)
$$

and the $n \times n$ matrix obtained from $A_{n+1, n}$ by deleting its $(k+1)$ th row $\left(z_{1}^{k}, z_{2}^{k}, \ldots, z_{n}^{k}\right)$ has determinant

$$
\prod_{1 \leqslant j<i \leqslant n}\left(z_{i}-z_{j}\right) \cdot T_{n-k}\left(z_{1}, z_{2}, \ldots, z_{n}\right) .
$$

Now we may prove Theorem 1. By symmetry it is enough to show that for each $0 \leqslant s \leqslant m, f\left(z_{1}\right) z^{s}$ has the same coefficient on both sides of (1.10). By (1.2), (2.1), and (2.2), this is equivalent to show that

$$
\begin{align*}
& \sum_{r=0}^{m} \bar{z}_{1}^{r} d(r+1, s+1) \\
& =\sum_{2 \leqslant i_{2}<i_{j}<\cdots<i_{m+1} \leqslant n}\left(\prod_{k=2}^{m+1}\left(\bar{z}_{1}-\bar{i}_{i_{k}}\right)\right)\left(\prod_{2 \leqslant j<k \leqslant m+1}\left|z_{i_{j}}-z_{i_{k}}\right|^{2}\right) \\
& \times(-1)^{m-s} T_{m-s}\left(z_{i}, \ldots, z_{i_{m+1}}\right) /\left(\operatorname{det} C_{m+1}\right) . \tag{2.4}
\end{align*}
$$

Here we have used the fact that $\left|z_{1}-z_{i}\right|^{2}=\left(z_{1}-z_{i}\right)\left(\bar{z}_{1}-\bar{z}_{i}\right)$. By (2.1) and (2.3), the left-hand side of (2.4) equals

$$
\begin{aligned}
& (-1)^{m+s} \sum_{1 \leqslant i_{1}<i_{2}<\cdots<i_{m} \leqslant n} T_{m-s}\left(z_{i_{1}}, z_{i_{2}}, \ldots, z_{i_{m}}\right) \cdot\left(\prod_{k=1}^{m}\left(\bar{z}_{1}-\bar{z}_{i_{k}}\right)\right) \\
& \quad \times\left(\prod_{1 \leqslant j<k \leqslant m}\left|z_{i_{j}}-z_{i_{k}}\right|^{2}\right) /\left(\operatorname{det} C_{m+1}\right) .
\end{aligned}
$$

In order that $\prod_{k=1}^{m}\left(\bar{z}_{1}-\bar{z}_{i_{k}}\right) \neq 0$ it is necessary that $i_{1} \geqslant 2$. Now change the index from $i_{k}$ to $i_{k+1}$, we get the right-hand side of (2.4). This completes the proof.

## 3. Proof of Theorem 2

Since $f \in A(\bar{D})$, there is a number $r>1$ such that $f$ is analytic on $\{z:|z| \leqslant r\}$. Under the assumption (1.1), it is well known [3, Theorem 3.6.1] that $f(z)-P_{m}\left(z ; f ; z_{i_{1}}, z_{i_{2}}, \ldots, z_{i_{m+1}}\right)$ equals the following contour integral

$$
(2 \pi i)^{-1} \int_{C}\left[f(t) \prod_{k=1}^{m+1}\left(z-z_{i_{k}}\right)\right] /\left[(t-z) \prod_{k=1}^{m+1}\left(t-z_{i k}\right)\right] d t
$$

where $C=\{z:|z|=r\}$. Because $\lim z_{n}=0$, it can be shown easily [3, Theorem 4.4.3] that there are positive constants $M, \delta$ with $r-\delta>1$, such that

$$
\left\|f(z)-P_{m}\left(z ; f ; z_{i_{1}}, z_{i_{2}}, \ldots, z_{i_{m+1}}\right)\right\|_{\infty} \leqslant M(r-\delta)^{-m}
$$

holds for all $\left\{z_{i_{1}}, z_{i_{2}}, \ldots, z_{i_{m+1}}\right\}$. Then the theorem follows from Theorem 1 .

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