# Signal Inpainting from Fourier magnitudes: An Almost Uniqueness Result 

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#### Abstract

This document takes interest in signal inpainting from Fourier magnitudes. This task consists in reconstructing consecutive missing samples in a finite discrete 1D signal, while assuming the magnitudes of its Fourier transform are known. In this report, we theoretically show that for almost all signals of length $L$, this problem admits a unique solution if at most $(L-1) / 3$ samples are missing.


## 1 Introduction and problem setting

Signal inpainting [1] is an inverse problem that consists in restoring signals degraded by sample loss. This problem typically arises as a result of degradation during signal transmission, digitization of physically degraded media, or degradation so heavy that the information about the samples can be considered lost $[5,2,3]$. Let $\boldsymbol{x} \in \mathbb{R}^{L}$ be a signal. Let $\bar{v} \subset\{0, \ldots, L-1\}$ denote a set of consecutive indices corresponding to missing samples in $\boldsymbol{x}$ and $v$ denote its complement, i.e., the set of indices corresponding to observed samples. We denote by $\boldsymbol{x}_{\bar{v}} \in \mathbb{R}^{d}$ the sub-signal of $\boldsymbol{x}$ restricted to missing samples and $\boldsymbol{x}_{v} \in \mathbb{R}^{L-d}$ the sub-signal restricted to observed samples. We denote by $\boldsymbol{b} \in \mathbb{R}_{+}^{L}$ the magnitudes of the discrete Fourier transform (DFT) of $\boldsymbol{x}$, i.e., $\boldsymbol{b}=|\boldsymbol{\Phi} \boldsymbol{x}|$, where $\boldsymbol{\Phi} \in \mathbb{C}^{L \times L}$ is the DFT matrix. For a given observed signal $\boldsymbol{y} \in \mathbb{R}^{L-d}$ and Fourier magnitudes $\boldsymbol{b}$, the task of signal inpainting from Fourier magnitudes can then be stated as:

$$
\begin{equation*}
\text { Find } \boldsymbol{u} \in \mathbb{R}^{d} \text { such that (s.t.) }|\boldsymbol{\Phi} \boldsymbol{x}|=\boldsymbol{b} \text { with } \boldsymbol{x}_{\bar{v}}=\boldsymbol{u} \text { and } \boldsymbol{x}_{v}=\boldsymbol{y} . \tag{1}
\end{equation*}
$$

We focus on the situation where the given vector $\boldsymbol{b}$ corresponds to the true magnitudes of the Fourier transform of a completed signal $\boldsymbol{x}$. Hence, the existence of at least one solution of (1) is guaranteed. In this document, we will show that when $d<(L-1) / 3$, this solution is unique for almost all signals $\boldsymbol{x} \in \mathbb{R}^{L}$. We use a dimension counting argument, similar in spirit to the one employed by [4] in the context of sparse phase retrieval. Specifically, we show that signals $\boldsymbol{x}$ for which more than one solution exists, referred to hereinafter as counter examples, necessary lie on a manifold of $\mathbb{R}^{L}$ with strictly less than $L$ degrees of freedom. They hence form a set of measure zero, i.e., a null set.

## 2 Almost uniqueness: statement and proof

We will assume throughout this section that the missing samples are placed at the beginning of $\boldsymbol{x}$, allowing us to write $\boldsymbol{x}=[\boldsymbol{u} ; \boldsymbol{y}]$ where $[\cdot ; \cdot]$ denotes vertical concatenation. This comes without loss of generality, because for any counter-example signal $\widetilde{\boldsymbol{x}}$ with consecutive missing samples placed anywhere in the signal, one can construct a counter example with samples placed at the beginning of the signal (and reciprocally) by a simple circular shift of $\widetilde{\boldsymbol{x}}$. Indeed, a circular shift does not affect DFT magnitudes. We prove the following theorem:

Theorem 1. Let $\mathcal{E}=\left\{\boldsymbol{x}=[\boldsymbol{u} ; \boldsymbol{y}] \in \mathbb{R}^{L} \mid \exists \boldsymbol{v} \in \mathbb{R}^{d}, \boldsymbol{v} \neq \boldsymbol{u}\right.$, s.t. $\left.|\boldsymbol{\Phi}[\boldsymbol{u} ; \boldsymbol{y}]|=|\boldsymbol{\Phi}[\boldsymbol{v} ; \boldsymbol{y}]|\right\}$. For $d<$ $(L-1) / 3, \mathcal{E}$ is a manifold of $\mathbb{R}^{L}$ with strictly less than $L$ degrees of freedom.

[^0]In other words, the set of counter examples to the unicity of (1) has strictly less than $L$ degrees of freedom in $\mathbb{R}^{d}$, and is hence of measure zero.

Proof. We denote by $\mathcal{E}^{\prime}$ the set of triplets $(\boldsymbol{y}, \boldsymbol{u}, \boldsymbol{v}) \in \mathbb{R}^{L-d} \times \mathbb{R}^{d} \times \mathbb{R}^{d}$ forming a counter example, namely:

$$
\begin{equation*}
\mathcal{E}^{\prime}=\left\{(\boldsymbol{y}, \boldsymbol{u}, \boldsymbol{v}) \in \mathbb{R}^{L-d} \times \mathbb{R}^{d} \times \mathbb{R}^{d}|\boldsymbol{v} \neq \boldsymbol{u},|\boldsymbol{\Phi}[\boldsymbol{u} ; \boldsymbol{y}]|=|\boldsymbol{\Phi}[\boldsymbol{v} ; \boldsymbol{y}]|\}\right. \tag{2}
\end{equation*}
$$

We will show that this manifold of $\mathbb{R}^{L-d} \times \mathbb{R}^{d} \times \mathbb{R}^{d}$ has strictly less that $L$ degrees of freedom; This implies that its projection $\mathcal{E}$ on $\mathbb{R}^{L-d} \times \mathbb{R}^{d} \equiv \mathbb{R}^{L}$ also has strictly less that $L$ degrees of freedom. We first prove the following:

Lemma 1. There is a linear bijection between $\mathcal{E}^{\prime}$ and the following set:

$$
\begin{equation*}
\mathcal{E}^{\prime \prime}=\left\{(\boldsymbol{a}, \boldsymbol{w}) \in \mathbb{R}^{d} \times \mathbb{R}^{L} \mid \boldsymbol{a} \neq \mathbf{0}_{d}, \mathcal{R}\left(\overline{\boldsymbol{\Phi}\left[\boldsymbol{a} ; \mathbf{0}_{L-d}\right]} \odot \boldsymbol{\Phi} \boldsymbol{w}\right)=\mathbf{0}_{L}\right\} \tag{3}
\end{equation*}
$$

where, $\mathcal{R}(\cdot)$ denotes the real part of a vector and $\odot$ denotes element-wise product.
Proof. We horizontally split the DFT matrix as $\boldsymbol{\Phi}=\left[\boldsymbol{\Phi}^{(1)}, \boldsymbol{\Phi}^{(2)}\right]$ where $\boldsymbol{\Phi}^{(1)} \in \mathbb{R}^{L \times d}$ and $\boldsymbol{\Phi}^{(2)} \in$ $\mathbb{R}^{L \times L-d}$ We have the following chain of equivalences (where $\boldsymbol{v} \neq \boldsymbol{u}$ is kept implicit):

$$
\begin{align*}
& (\boldsymbol{y}, \boldsymbol{u}, \boldsymbol{v}) \in \mathcal{E}^{\prime}  \tag{4}\\
\Leftrightarrow & |\boldsymbol{\Phi}[\boldsymbol{u} ; \boldsymbol{y}]|^{2}=|\boldsymbol{\Phi}[\boldsymbol{v} ; \boldsymbol{y}]|^{2}  \tag{5}\\
\Leftrightarrow & \left|\boldsymbol{\Phi}^{(1)} \boldsymbol{u}+\boldsymbol{\Phi}^{(2)} \boldsymbol{y}\right|^{2}=\left|\boldsymbol{\Phi}^{(1)} \boldsymbol{v}+\boldsymbol{\Phi}^{(2)} \boldsymbol{y}\right|^{2}  \tag{6}\\
\Leftrightarrow & \left|\boldsymbol{\Phi}^{(1)} \boldsymbol{u}\right|^{2}+2 \mathcal{R}\left(\overline{\boldsymbol{\Phi}^{(1)} \boldsymbol{u}} \odot \boldsymbol{\Phi}^{(2)} \boldsymbol{y}\right)+\left|\boldsymbol{\Phi}^{(2)} \boldsymbol{y}\right|^{2}=\left|\boldsymbol{\Phi}^{(1)} \boldsymbol{v}\right|^{2}+2 \mathcal{R}\left(\overline{\boldsymbol{\Phi}^{(1)} \boldsymbol{v}} \odot \boldsymbol{\Phi}^{(2)} \boldsymbol{y}\right)+\left|\boldsymbol{\Phi}^{(2)} \boldsymbol{y}\right|^{2}  \tag{7}\\
\Leftrightarrow & \left|\boldsymbol{\Phi}^{(1)} \boldsymbol{u}\right|^{2}-\left|\boldsymbol{\Phi}^{(1)} \boldsymbol{v}\right|^{2}+2 \mathcal{R}\left(\overline{\boldsymbol{\Phi}^{(1)}(\boldsymbol{u}-\boldsymbol{v})} \odot \boldsymbol{\Phi}^{(2)} \boldsymbol{y}\right)=\mathbf{0}_{L}  \tag{8}\\
\Leftrightarrow & \mathcal{R}\left(\overline{\left.\left(\overline{\boldsymbol{\Phi}^{(1)} \boldsymbol{u}}-\overline{\boldsymbol{\Phi}^{(1)} \boldsymbol{v}}\right) \odot\left(\boldsymbol{\Phi}^{(1)} \boldsymbol{u}+\boldsymbol{\Phi}^{(1)} \boldsymbol{v}\right)\right)+2 \mathcal{R}\left(\overline{\boldsymbol{\Phi}^{(1)}(\boldsymbol{u}-\boldsymbol{v})} \odot \boldsymbol{\Phi}^{(2)} \boldsymbol{y}\right)=\mathbf{0}_{L}}\right.  \tag{9}\\
\Leftrightarrow & \mathcal{R}\left(\overline{\boldsymbol{\Phi}^{(1)}(\boldsymbol{u}-\boldsymbol{v})} \odot\left(\boldsymbol{\Phi}^{(1)} \boldsymbol{u}+\boldsymbol{\Phi}^{(1)} \boldsymbol{v}+2 \boldsymbol{\Phi}^{(2)} \boldsymbol{y}\right)=\mathbf{0}_{L}\right.  \tag{10}\\
\Leftrightarrow & \mathcal{R}\left(\overline{\boldsymbol{\Phi}\left[\boldsymbol{u}-\boldsymbol{v} ; \mathbf{0}_{L-d}\right]} \odot(\boldsymbol{\Phi}[\boldsymbol{u}+\boldsymbol{v} ; 2 \boldsymbol{y}])=\mathbf{0}_{L}\right.  \tag{11}\\
\Leftrightarrow & \mathcal{R}\left(\overline{\boldsymbol{\Phi}\left[\boldsymbol{a} ; \mathbf{0}_{L-d}\right]} \odot \boldsymbol{\Phi} \boldsymbol{w}\right)=\mathbf{0}_{L} \text { where } \boldsymbol{a}=\boldsymbol{u}-\boldsymbol{v} \neq \mathbf{0}_{d} \text { and } \boldsymbol{w}=[\boldsymbol{u}+\boldsymbol{v} ; 2 \boldsymbol{y}] \in \mathbb{R}^{L} . \tag{12}
\end{align*}
$$

Since the transformation from $(\boldsymbol{y}, \boldsymbol{u}, \boldsymbol{v})$ to $(\boldsymbol{a}, \boldsymbol{w})$ is linear and bijective, this concludes the proof.
Based on Lemma 1, it is sufficient to show that $\mathcal{E}^{\prime \prime}$ has strictly less than $L$ degrees of freedom. Since the non-zero signal $\boldsymbol{a} \in \mathbb{R}^{d}$ in (3) can be chosen arbitrarily ( $d$ degrees of freedom), it remains to show that for a fixed $\boldsymbol{a} \neq \mathbf{0}_{d}$, the set of $\boldsymbol{w} \in \mathbb{R}^{L}$ such that $(\boldsymbol{a}, \boldsymbol{w}) \in \mathcal{E}^{\prime \prime}$ has strictly less than $L-d$ degrees of freedom. For conciseness, we will only treat here the case where $L$ is even, as the odd case only requires minor adjustments.

Let $\hat{\boldsymbol{a}}=\boldsymbol{\Phi}\left[\boldsymbol{a} ; \mathbf{0}_{L-d}\right]$ and $\hat{\boldsymbol{w}}=\boldsymbol{\Phi} \boldsymbol{w}$ be the DFTs of $\left[\boldsymbol{a} ; \mathbf{0}_{L-d}\right]$ and $\boldsymbol{w}$, indexed by the $L$ discrete frequency numbers $f \in\{-L / 2+1, \ldots, L / 2\}$. Since the signals $\boldsymbol{a}$ and $\boldsymbol{w}$ are real-valued, their DFTs are fully determined by their values at non-negative frequencies, two of which are real (at $f=0$ and $f=L / 2$ ), the rest being complex. For every frequency number $f \in\{1, \cdots, L / 2-1\}$ such that $\hat{a}(f) \neq 0$, the constraint $\mathcal{R}(\overline{\hat{\boldsymbol{a}}} \odot \hat{\boldsymbol{w}})=\mathbf{0}_{L}$ fixes the phase of $\hat{w}(f)$ up to $\pm \pi / 2$, reducing the degrees of freedom of $\boldsymbol{w}$ by 1 (from a total of $L$ ). Let us now count for how many distinct $f \in\{1, \cdots, L / 2-1\}$ we can have $\hat{a}(f) \neq 0$. The z-transform of $\left[\boldsymbol{a} ; \mathbf{0}_{L-d}\right]$ is a polynomial of degree at most $d-1$. Hence, this polynomial admits at most $d-1$ roots, and since $\boldsymbol{a}$ is real-valued, these roots are either real or come in conjugate pairs. This implies that $\hat{a}(f)$ can be 0 for at most $\lfloor(d-1) / 2\rfloor$ distinct $f$ in $\{1, \cdots, L / 2-1\}$. Hence, the constraint $\mathcal{R}(\overline{\hat{\boldsymbol{a}}} \odot \hat{\boldsymbol{w}})=\mathbf{0}_{L}$ enforces at least $L / 2-1-\lfloor(d-1) / 2\rfloor$ phase constraints on $\hat{\boldsymbol{w}}$. Subtracting these constraints from $L$, we get that $\boldsymbol{w}$ has at most $P=L / 2+1+\lfloor(d-1) / 2\rfloor$ degrees of freedom. By hypothesis, $d<L / 3-1$, which implies $P<L-d$ and concludes the proof.

## 3 Conclusion

We have conducted a theoretical study on the solutions to the problem of signal inpainting from Fourier magnitudes. We have shown that if the number of missing samples $d$ is strictly less than $(L-1) / 3$, where $L$ is the total signal length, then almost all signals containing a subset of $d$ consecutive missing values are uniquely determined from the magnitudes of their Fourier transform.

## References

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