



# An Error Analysis for the Sampling of Periodic Signals

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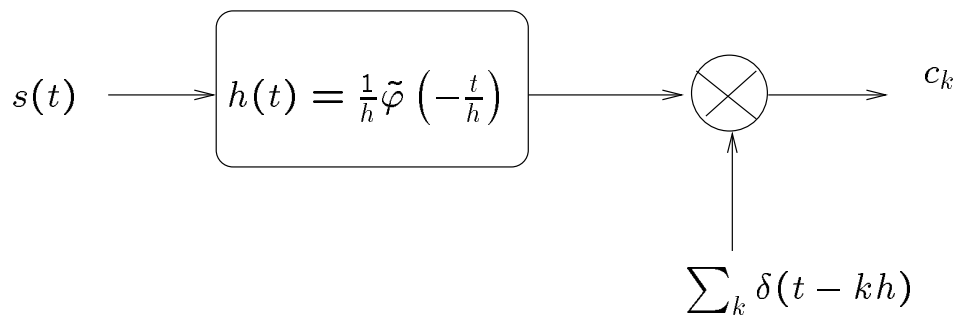
\*<http://bigwww.epfl.ch>

# Outline

1. Introduction
2. Sampling of Periodic Signals
3. Computation of Square Error
4. Asymptotic Performance
5. Experimental Validation
6. Conclusions

# Introduction

## Generalized Sampling



$$c_k = \left\langle s(t), \frac{1}{h} \tilde{\varphi}\left(\frac{t}{h} - k\right) \right\rangle$$

$$s_h(t) = \sum_{k=-\infty}^{\infty} c_k \varphi\left(\frac{t}{h} - k\right)$$

- Choice of  $\varphi$   $\iff$ 
  - Bandlimited,
  - Spline
  - Wavelet representations
- Well developed theory for  $L_2(\mathbb{R})$
- Easily extendable to periodic case by periodization of  $\varphi$  and  $\tilde{\varphi}$  [Chuang]

## Quantitative Error Analysis

- Sharp Error Estimates [Blu et. al.]  $\rightarrow$  Quantification of Approximation Error

$$MSE(h) \approx \sqrt{\frac{1}{2\pi} \int |\hat{s}(\omega)|^2 E(h\omega) d\omega}$$

$\Rightarrow$  Optimal choice of sampling step, basis functions and algorithms

- Square modulus of Fourier transform not defined  $\Rightarrow$  Extension

## Sampling of Periodic Signals

- Representation of closed curves
- Signal representation with boundary conditions(eg. periodic boundary conditions)

### Synthesis and analysis equations

$$s_N(t) = \sum_{k=0}^{N-1} c_k \varphi_p \left( \frac{t}{h} - k \right)$$

$$c_k = \int_0^T s(\xi) \tilde{\varphi}_p \left( \frac{\xi}{h} - k \right) d\frac{\xi}{h},$$

where

$$\varphi_p(t) = \sum_{l=-\infty}^{\infty} \varphi(t - lN); \quad T = Nh$$

### Approximation Operator

$$\begin{aligned} s_N(t) &= Q_N s(t) \\ &= \sum_{k=0}^{N-1} \left[ \int_0^T s(\xi) \tilde{\varphi}_p \left( \frac{\xi}{h} - k \right) d\frac{\xi}{h} \right] \varphi_p \left( \frac{t}{h} - k \right) \end{aligned}$$

Projection  $\leftrightarrow$  Consistent Sampling iff  $\varphi_p$  and  $\tilde{\varphi}_p$  are biorthogonal.

## Computation of the Square Error

- Reconstruction in  $V_\varphi$  : Not shift invariant in general

Error is dependent on

1. Time shift of the function —  $\tau$

$$s(t) \rightarrow s_\tau(t) = s(t - \tau)$$

2. Number of samples —  $N$

## Mean Square Approximation Error

$$\begin{aligned} [\gamma_s(\tau, N)]^2 &= \frac{1}{T} \int_0^T |s_\tau(t) - \mathcal{Q}_N s_\tau(t)|^2 dt \\ &= \|s_\tau - \mathcal{Q}_N s_\tau\|_{L_2(0, T]}^2 \end{aligned}$$

$T = Nh \Rightarrow \gamma_s(\tau, N)$  is  $h$  periodic in  $\tau$

Phase is unknown in most applications  $\rightarrow$  An average measure of error is desirable

## Average Approximation Error

$$\eta_s(N) = \sqrt{\frac{1}{h} \int_0^h [\gamma_s(\tau, N)]^2 d\tau}$$

**Theorem 1 :**

$$\eta_s(N) = \sqrt{\sum_{k=-\infty}^{\infty} |S(k)|^2 E\left(\frac{2\pi k}{N}\right)}$$

$$\begin{aligned} E(\omega) &= \left| 1 - \tilde{\varphi}(\omega)^* \hat{\varphi}(\omega) \right|^2 \\ &\quad + |\tilde{\varphi}(\omega)|^2 \sum_{n \neq 0} |\hat{\varphi}(\omega + 2n\pi)|^2 \\ &= \underbrace{1 - \frac{|\hat{\varphi}(\omega)|^2}{\hat{a}_\varphi(\omega)}}_{E_{\min}(\omega)} + \underbrace{\hat{a}_\varphi(\omega) |\tilde{\varphi}(\omega) - \hat{\varphi}_d(\omega)|^2}_{E_{\text{res}}(\omega)}, \end{aligned}$$

$$S(k) = \frac{1}{M} \int_0^M s(x) e^{j\frac{2\pi kx}{M}} dx$$

Here,

$$\hat{a}_\varphi(\omega) = \sum_{k=-\infty}^{\infty} |\hat{\varphi}(\omega + 2n\pi)|^2$$

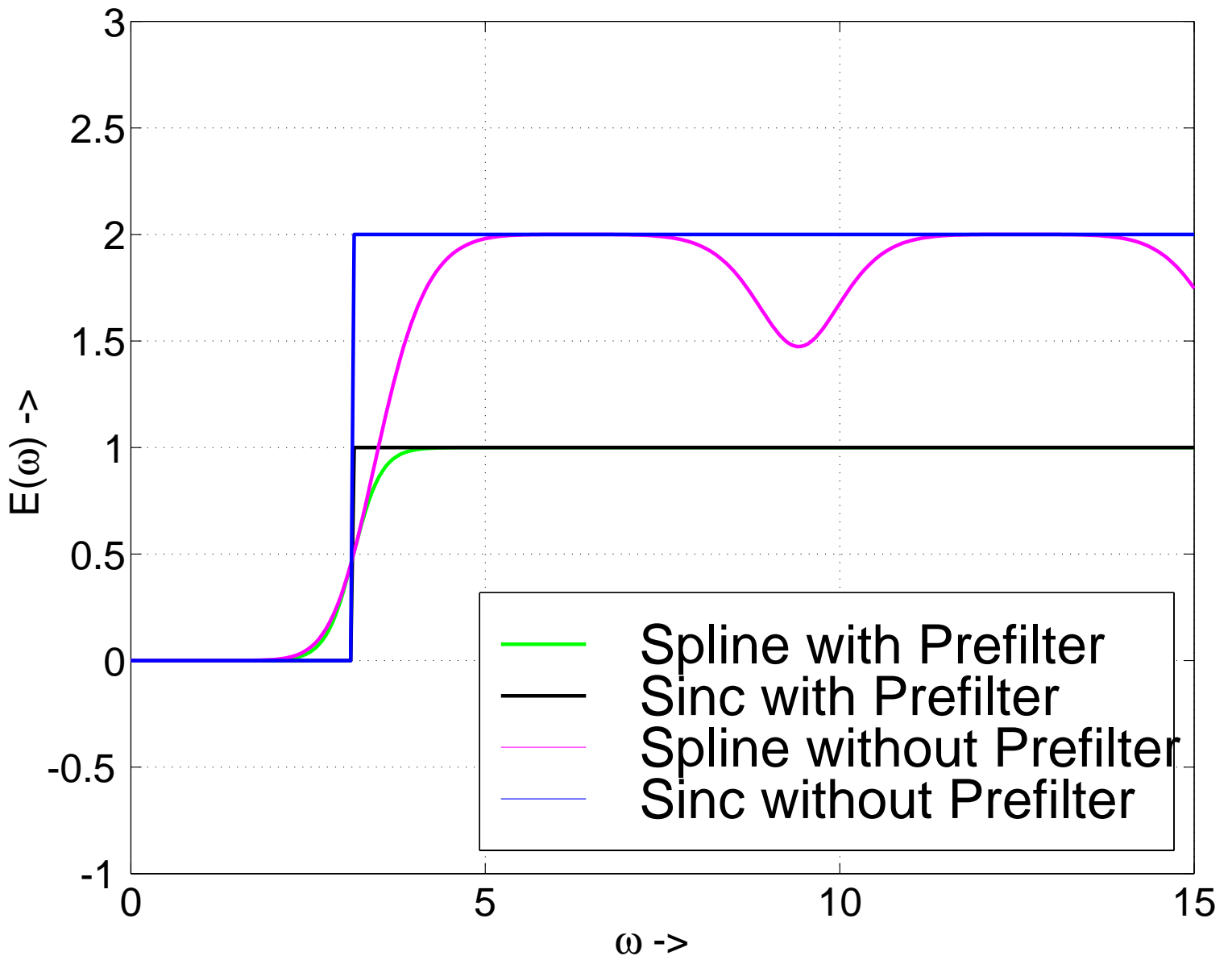
and

$$\hat{\varphi}_d(\omega) = \frac{\hat{\varphi}(\omega)}{\hat{a}_\varphi(\omega)}$$

- $\eta_s(N)$  independent of  $\tau \Rightarrow$  No alias terms
- Kernel identical to  $L_2(\mathbb{R})$  case
- Error formula is a discrete sum as compared to an integral in  $L_2(\mathbb{R})$
- $\tilde{\varphi} = \varphi_d \Rightarrow E(\omega) = E_{\min}(\omega)$  (orthogonal projection)

$$\eta_s(N) = \eta_{s,\min}(N); \quad \forall N \quad \text{iff} \quad \tilde{\varphi} = \varphi_d$$





Error kernels for cubic B-Spline and Sinc interpolation

## Asymptotic Performance

Analysis of  $\eta_s(N)$  as  $N \rightarrow \infty$

Is linked to the behavior of the kernel as  $\omega \rightarrow 0$

Rate of Decay of Error  $\iff$  Differentiability of the kernel at the origin

Strang Fix Conditions of Order  $L$  ( $E(\omega) = \mathcal{O}(\omega^{2L})$ )

$$\hat{\varphi}(0) \neq 0$$

$$\hat{\varphi}^{(n)}(2k\pi) = 0, \quad \forall k \in \mathbb{Z} \setminus \{0\}$$

for  $n = 0, 1 \dots L - 1$

$\rightarrow \varphi$  is called an  $L^{\text{th}}$  order scaling function

**Theorem 2** : Let  $\varphi$  and  $\tilde{\varphi}$  be bi-orthonormal

$\varphi$  is an  $L^{\text{th}}$  order generating function



$$\begin{aligned} \eta_s(N) &\underset{N \rightarrow \infty}{=} C_{\varphi, \tilde{\varphi}} \| (2\pi k)^L S(k) \|_{l_2} \left(\frac{1}{N}\right)^L \\ &= C_{\varphi, \tilde{\varphi}} T^L \| s^{(L)} \|_{L_2[0, T)} \left(\frac{1}{N}\right)^L \end{aligned}$$

$$C_{\varphi, \tilde{\varphi}} = \frac{1}{L!} \sqrt{|m_{\varphi_d}^L - m_{\tilde{\varphi}}^L|^2 + \sum_{k \neq 0} |\hat{\varphi}^L(2\pi k)|^2}$$

- Similar expression as [\[Unser\]](#) for functions in  $L_2(\mathbb{R})$

- $m_{\varphi_d}^L = m_{\tilde{\varphi}}^L \Rightarrow C_{\varphi, \tilde{\varphi}} = C_{\varphi}^-$

Minimum attainable constant

Independent of the analysis function

# Experimental Validation

## Reference shape

Map of Switzerland represented as a polygon with 807 edges

Represented using two periodic functions  $x(t)$  and  $y(t)$

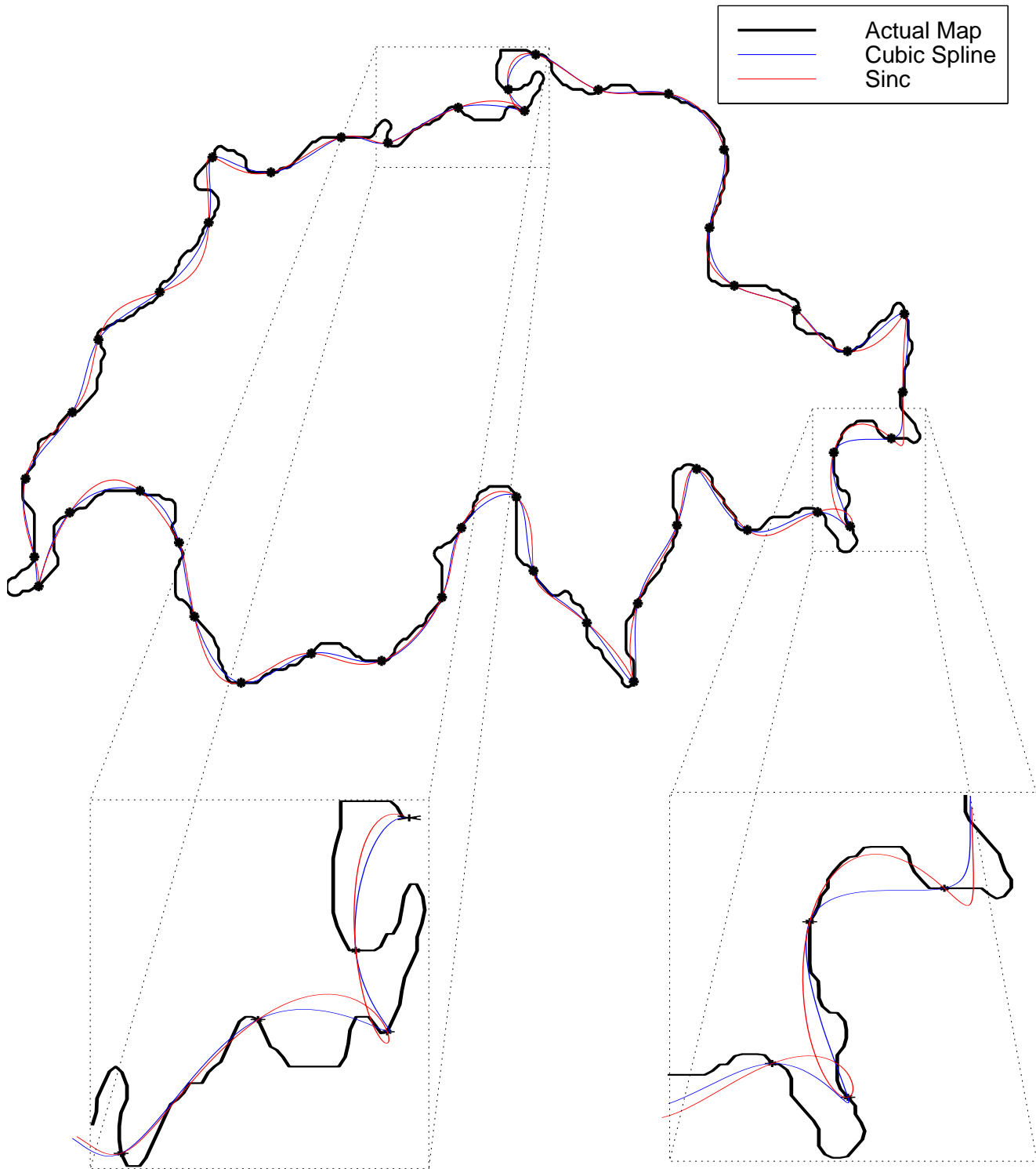
## Experiment

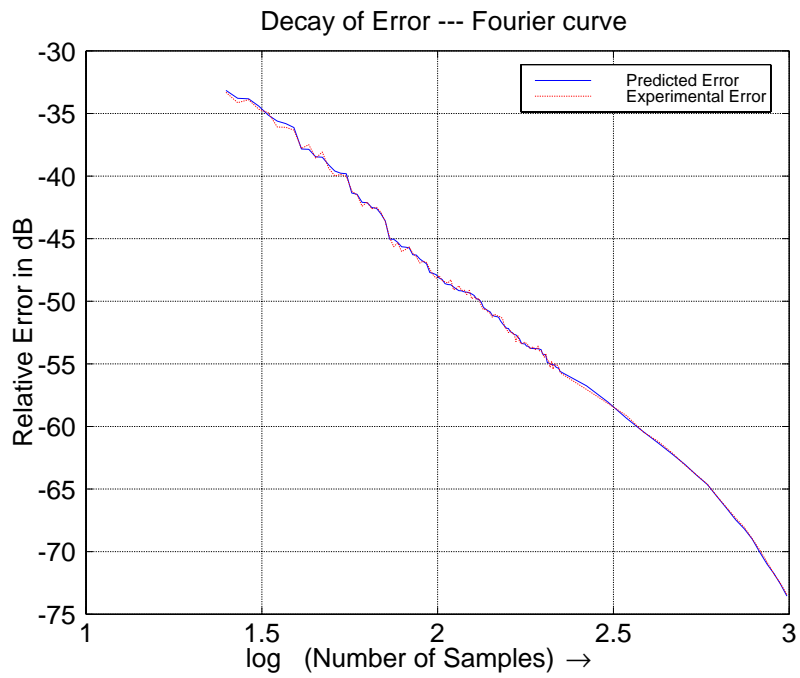
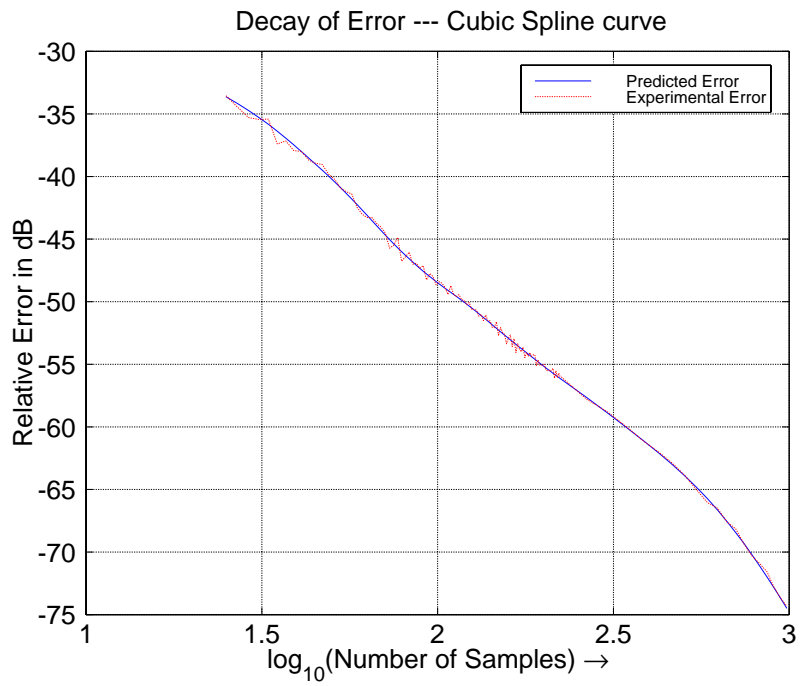
a) Initial model resampled

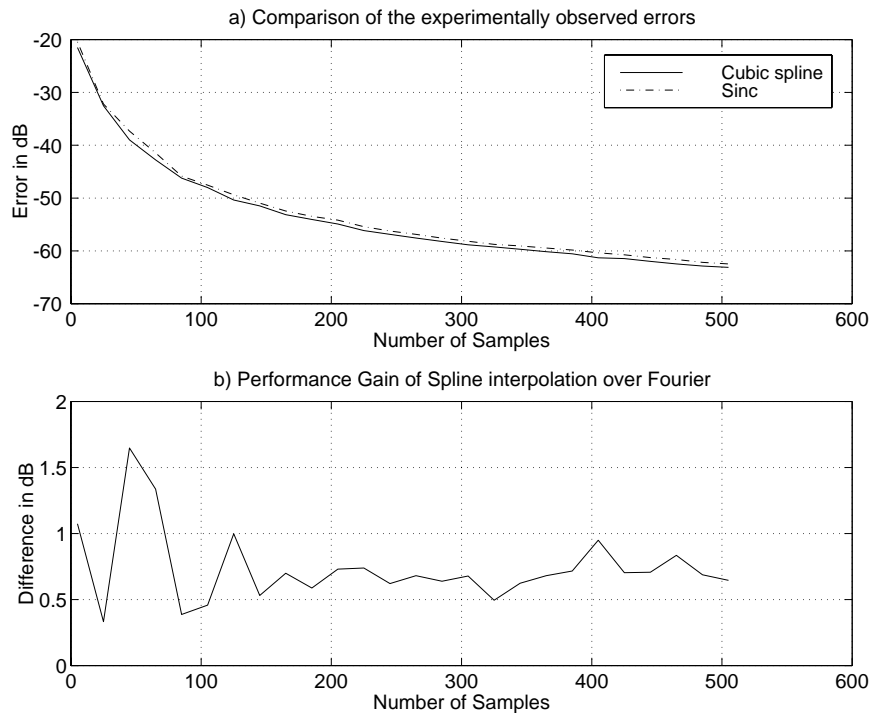
1) Cubic spline

b) Interpolated with 2) Sinc  $\Leftrightarrow$  Fourier representation

c) Error compared with theoretical prediction







## Conclusions

- Exact expression of approximation error for periodic signals
- Allows the optimization of  $\varphi$  and  $\tilde{\varphi}$
- Experimental verification of the error formula
- Asymptotic performance