

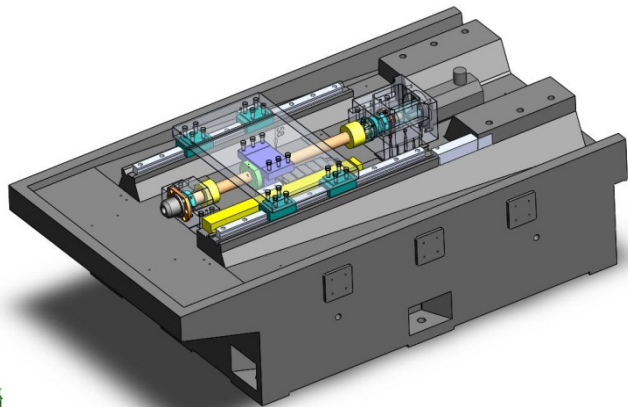
利用時頻與多尺度熵分析螺桿之螺帽預壓失效的研究

國立彰化師範大學 機電工程學系  
黃宜正

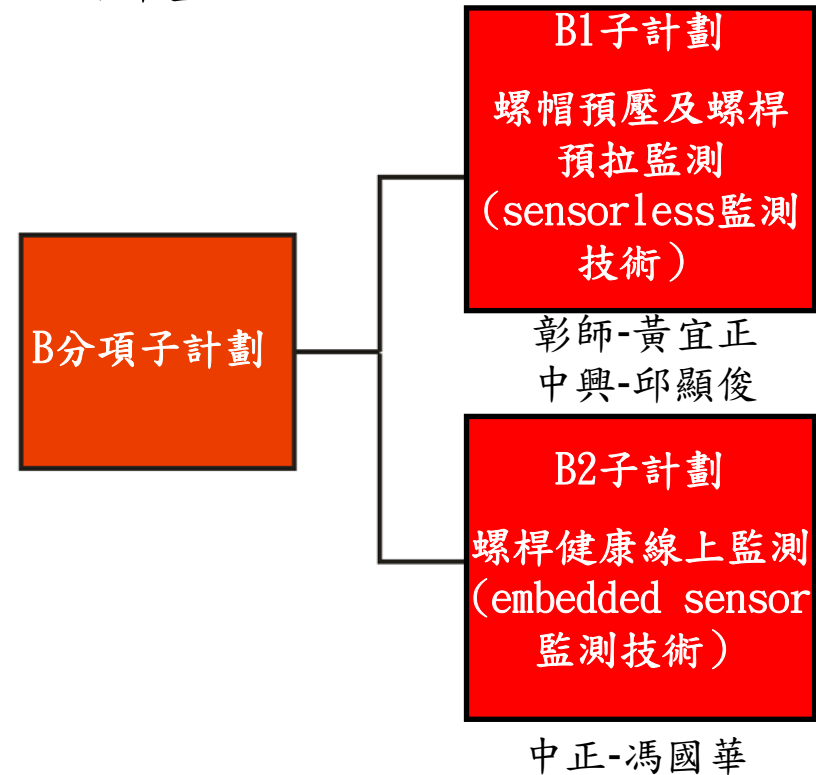
# 智能化螺桿計畫的分工架構

分項計畫目標：

- ◆ 透過精密機械發展中心、寶元數控與上銀公司共同合作，建立實驗平台
- 1. 開發不同螺帽預壓量的螺桿(上銀公司提供五組)，建立監測到螺帽預壓與螺桿預拉失效之Sensorless監測技術。
- 2. 開發嵌入式感測器(溫度、加速度)的智能化螺桿技術

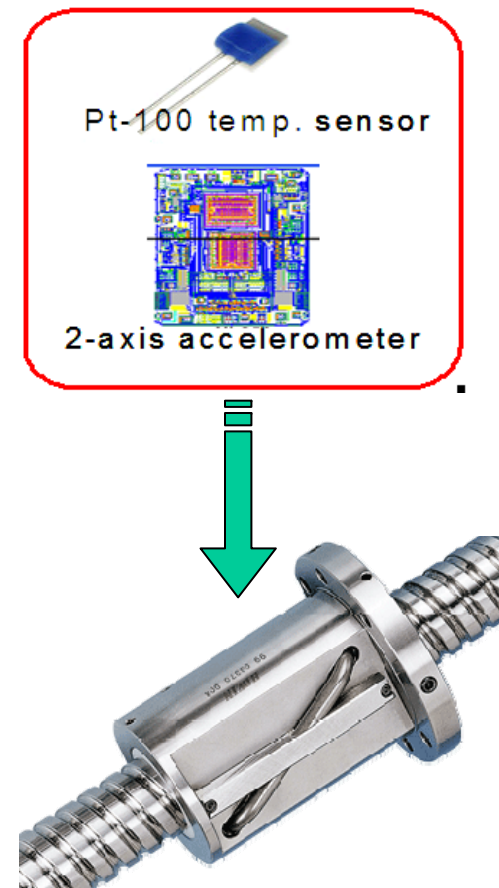


分項計畫分工：

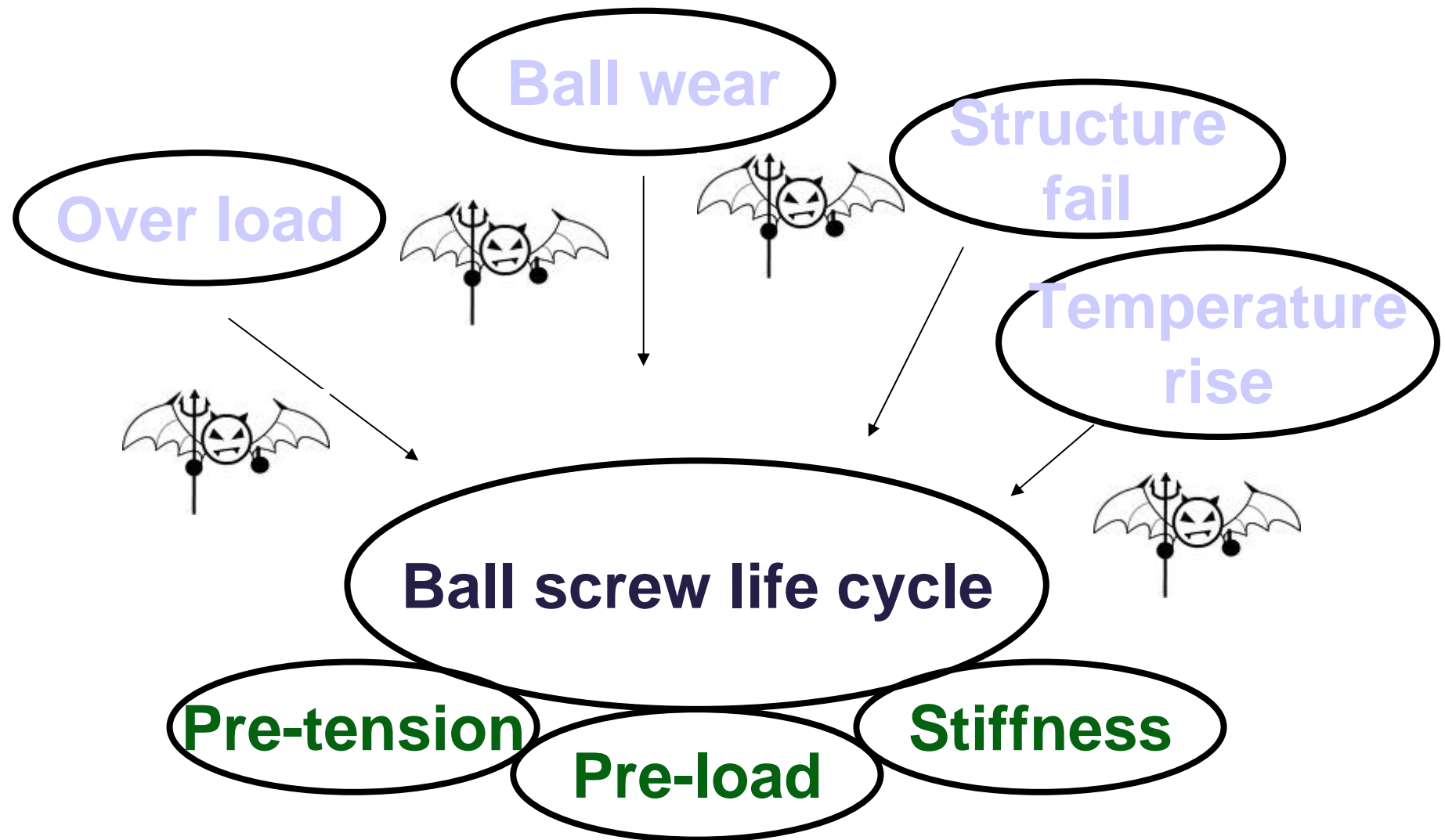


# 智能化螺桿之研究動機

- 螺桿由於長期使用，螺桿和螺帽間隙變大、滾珠過度磨耗，導致定位精確度變差，進而失效~使用者將難以選取穩定長時間的加工控制參數。
- 此外，螺桿上的切屑及油污可導致螺桿破損和滾珠的磨損增加，而縮短螺桿的壽命。
- 故可靠的螺桿健康監測乃極度重要，除可避免刀具使用者額外的成本支出；螺桿壽命的預測更可免去無預警的停工期。

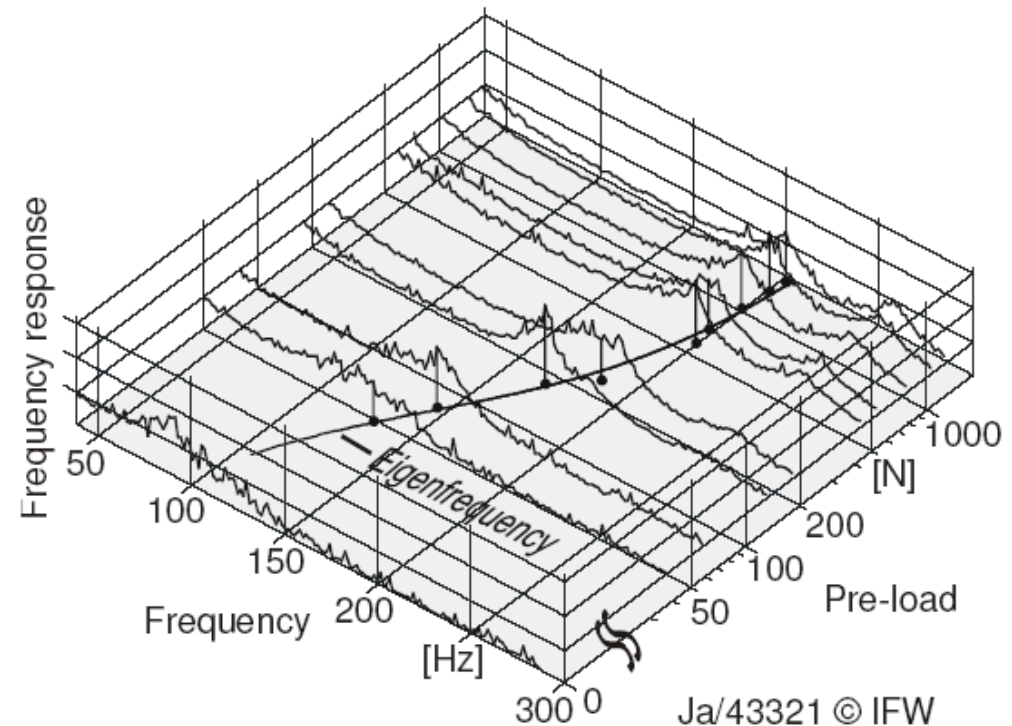


# Research Motivation



# 滾珠螺桿壽命的預估 (I)

- 影響螺桿的自然頻率的因素，計有滑塊的質量、螺桿本身及螺帽的剛性等。
- 預壓的損失將導致螺桿系統特徵頻率 (Eigen-frequency) 降低。

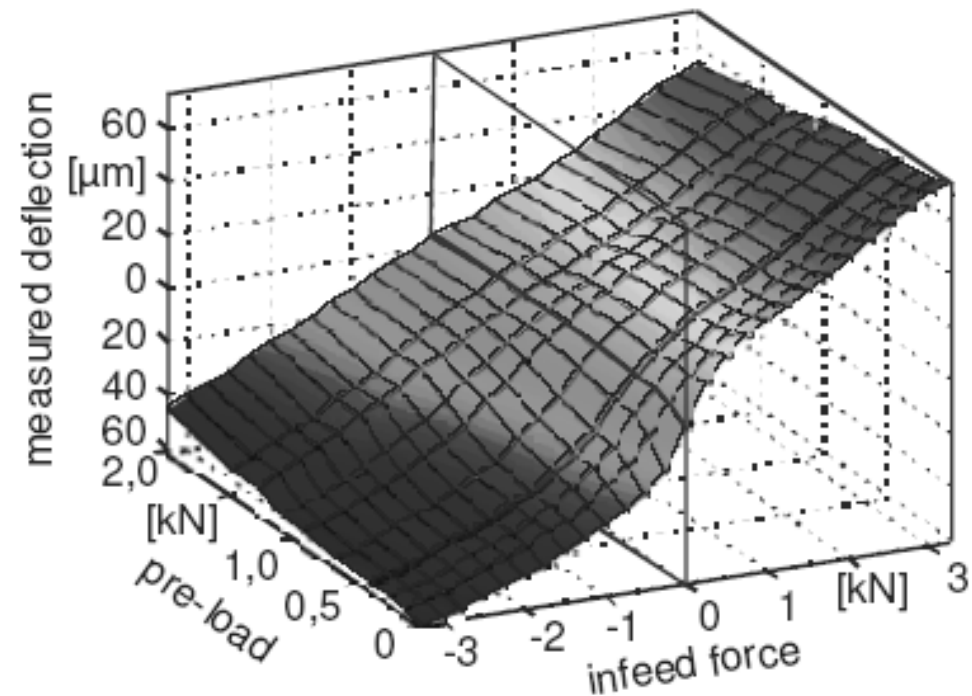


Eigen-frequency 與預壓的關係



## 滾珠螺桿壽命的預估(II)

- 未磨耗的滾珠螺桿具近線性的彈簧常數特徵。
- 當滾珠磨損後，螺桿間隙變大，使得低進給力時螺桿系統有較低的剛性。因而線性的彈簧常數呈“S”形變化（如右圖）。

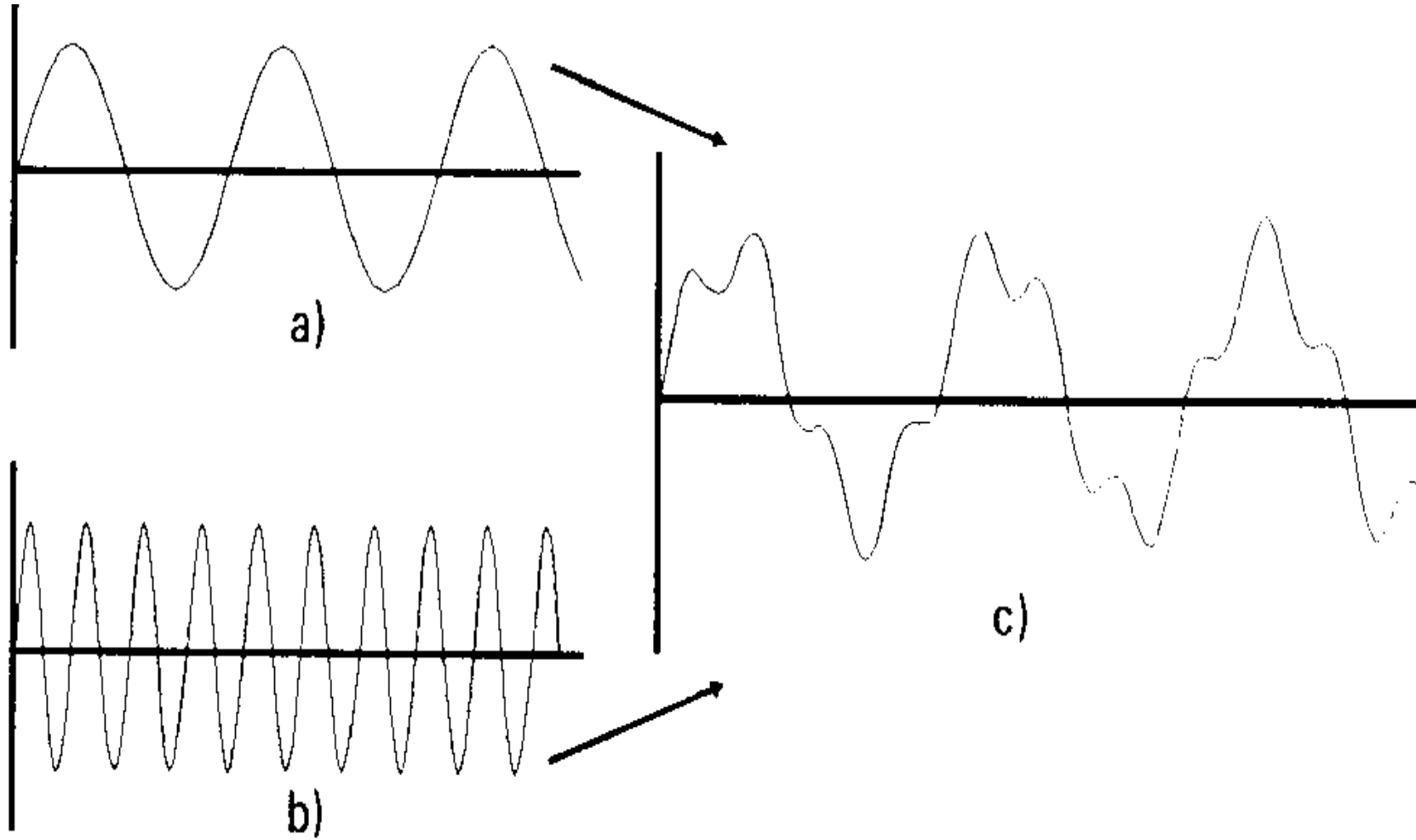


螺桿量測的進給力與變形量之關係

# Outline

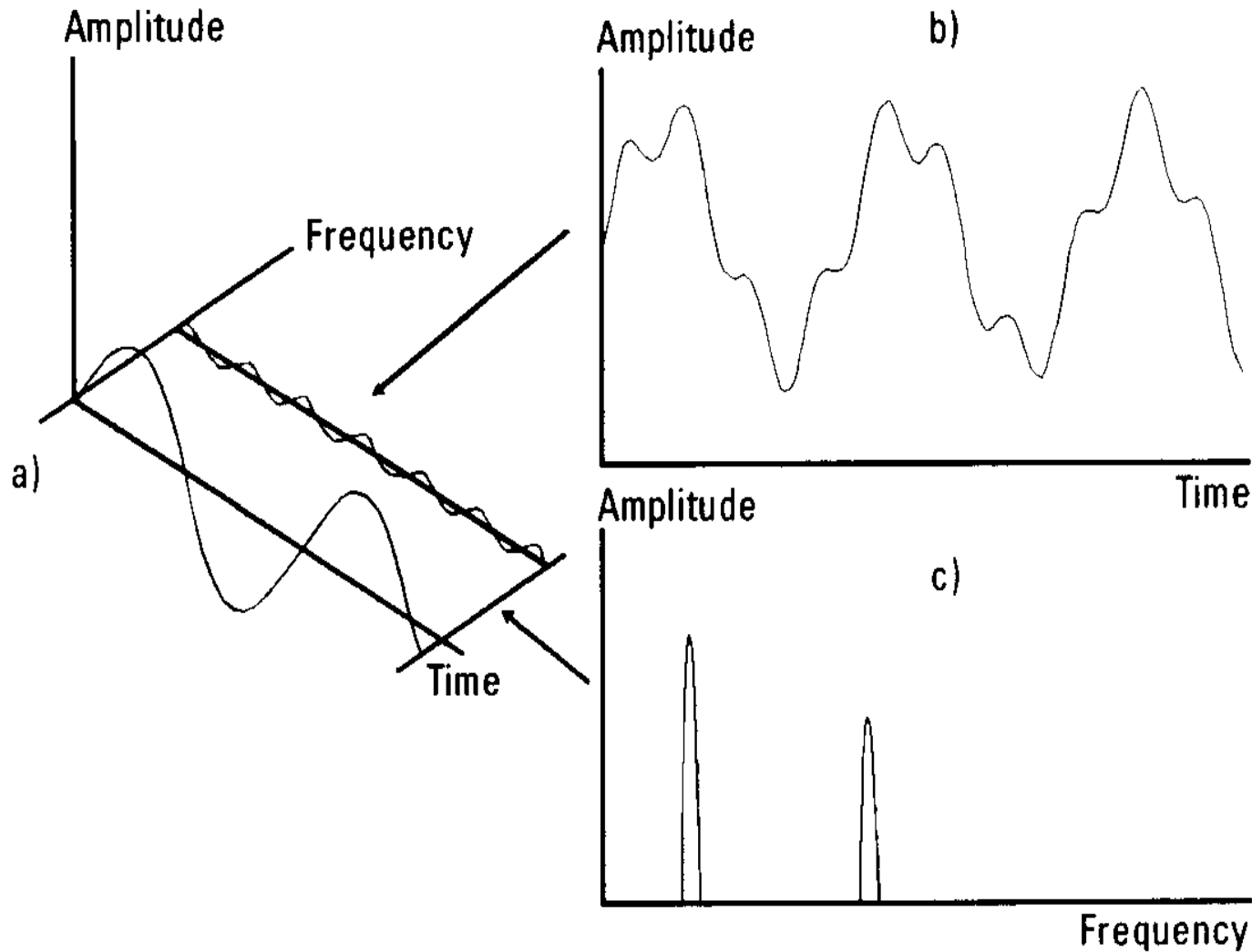
- 1 EMD/HHT (Hilbert Huang Transform)
- 2 MSE (Multiscale Entropy)
- 3 Torque Data Acquisition Analysis
- 4 Conclusion Remarks

Hybrid Signals



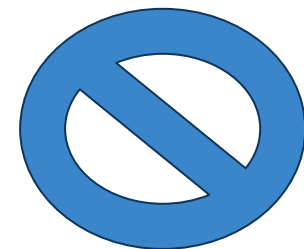
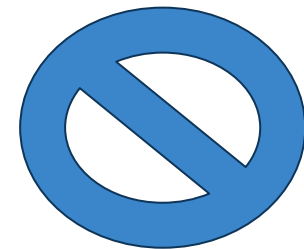
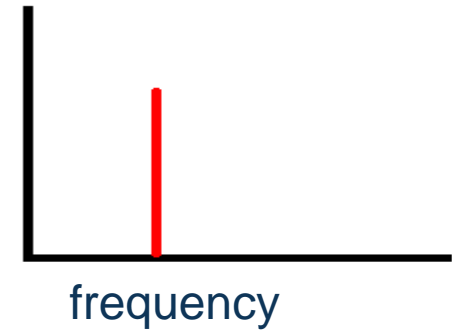
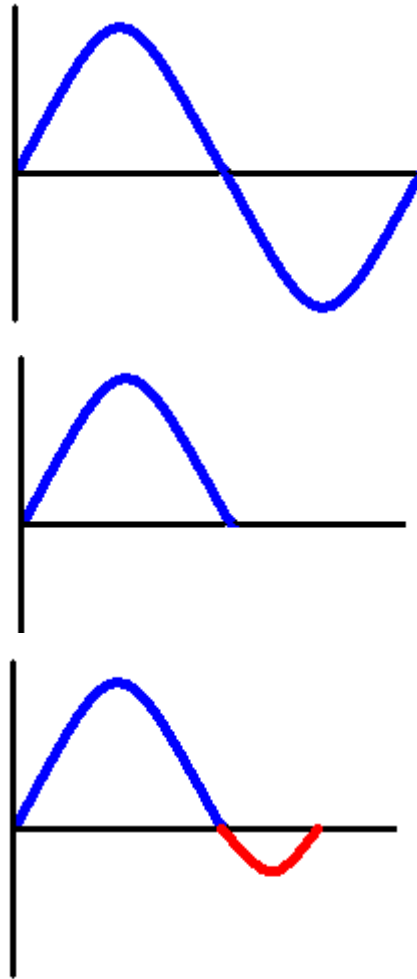
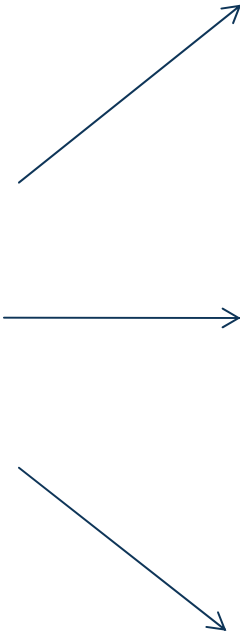


時域信號/頻率譜分析



# Fourier Condition

**Fourier Transform**



## WHY 時頻分析

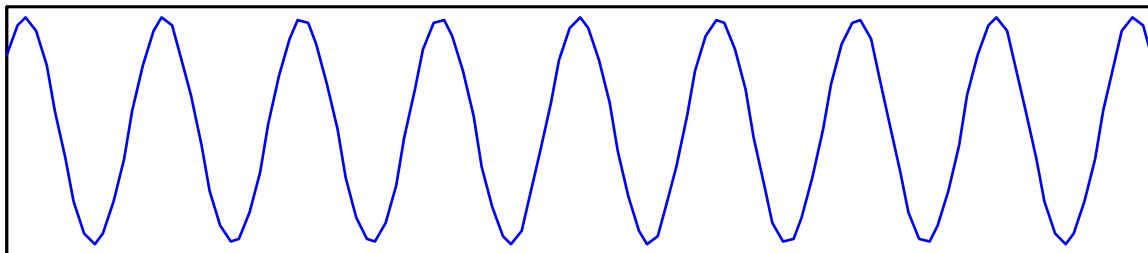
- ❖ 頻率為時間訊號的重要特徵，傳統用頻譜分析來了解一段時間內頻譜的分佈（傅利葉分析），但有時，我們更有興趣的是**頻率隨時間變化的情形**，如聲音語調的變化，機器運轉特異震動的頻率出現的時間其持續的長短。
- ❖ 了解頻率隨時間變化的情形，**分析各種不同頻率隨時間變化的情形**稱為**時頻分析**。時頻分析相較於頻譜分析多了頻率對時間的解析。利用時頻分析可以看出頻率隨著時間變化的情形。



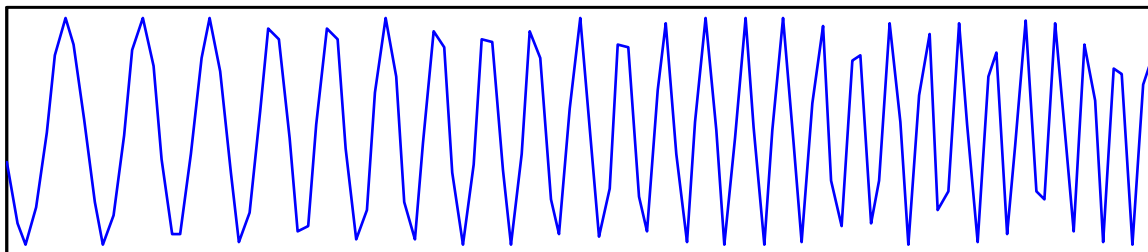
# ❖ Empirical Mode Decomposition



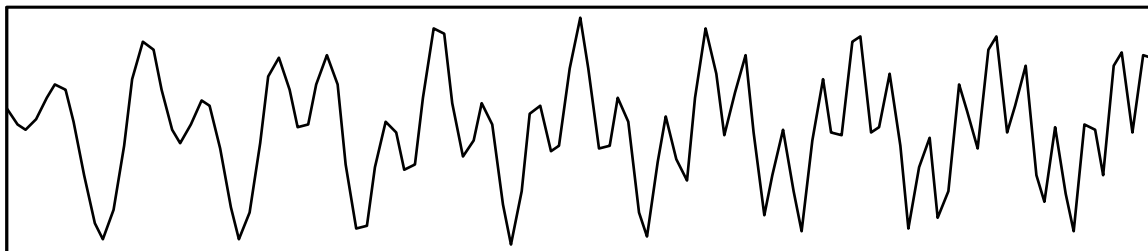
tone



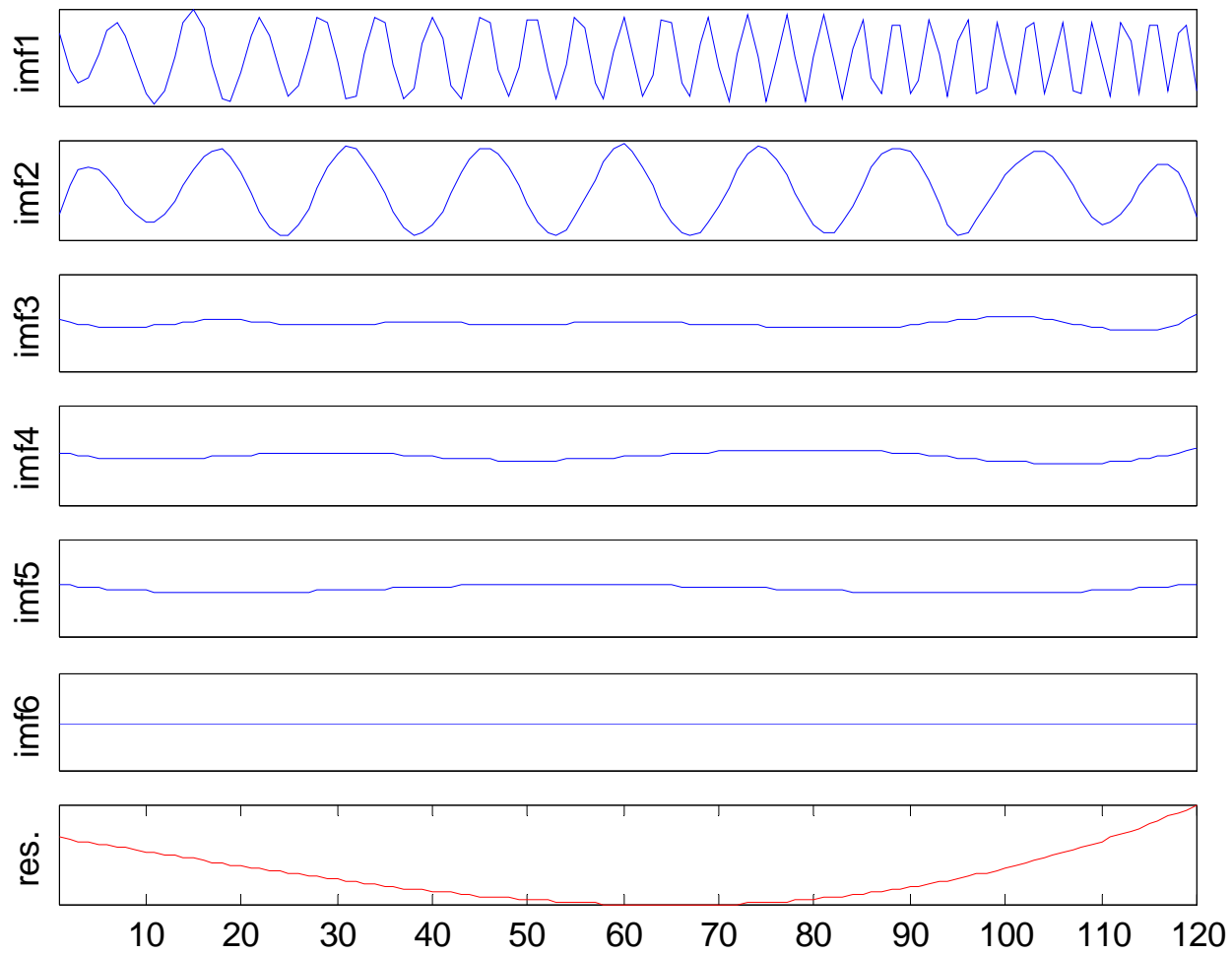
chirp



tone + chirp



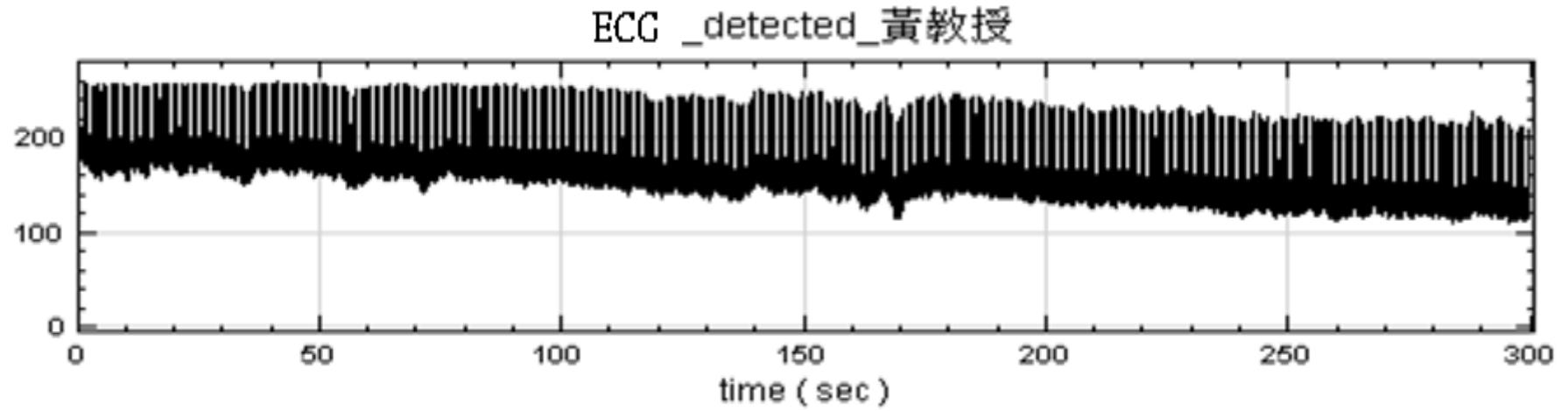
# Empirical Mode Decomposition



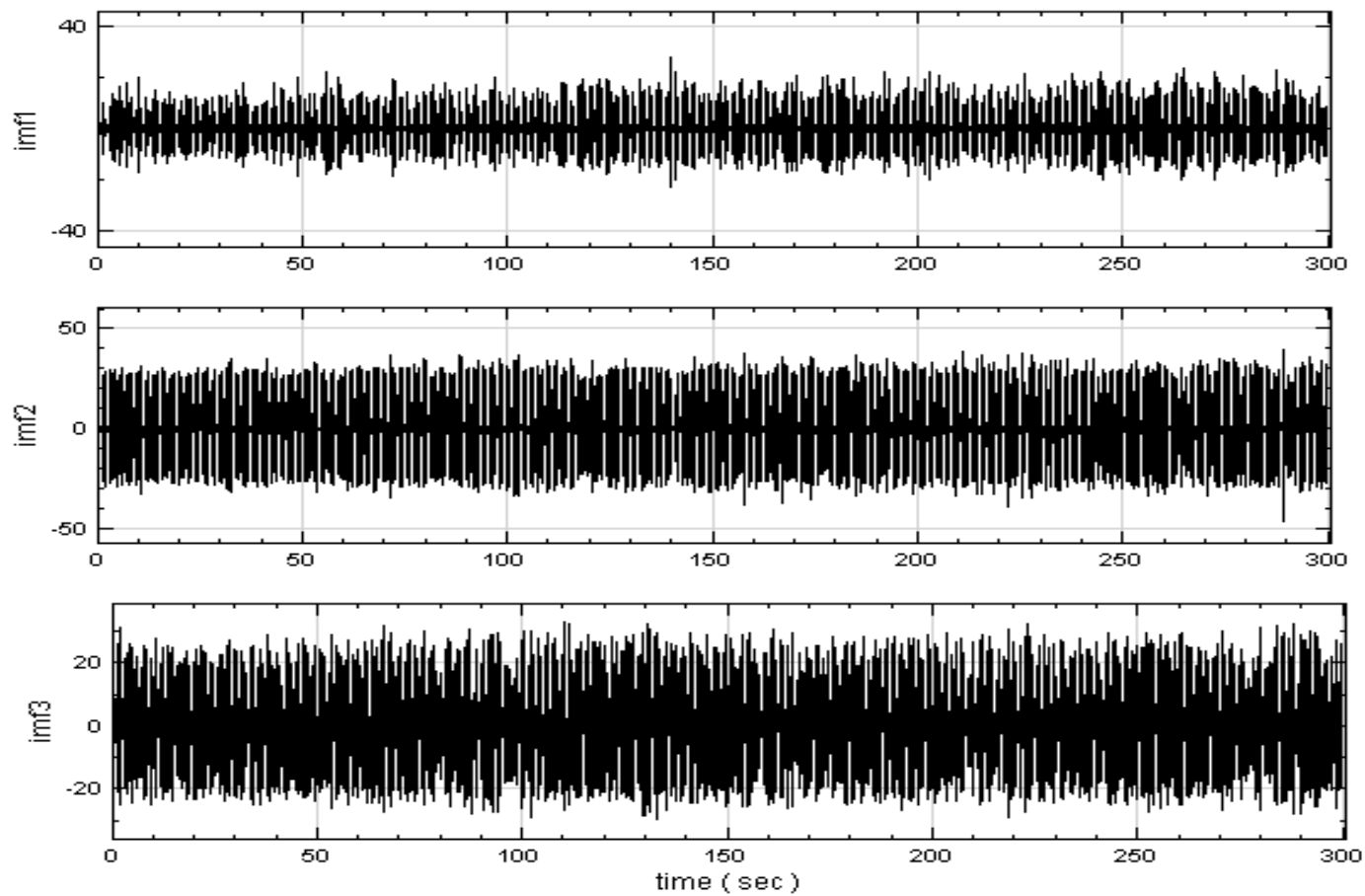


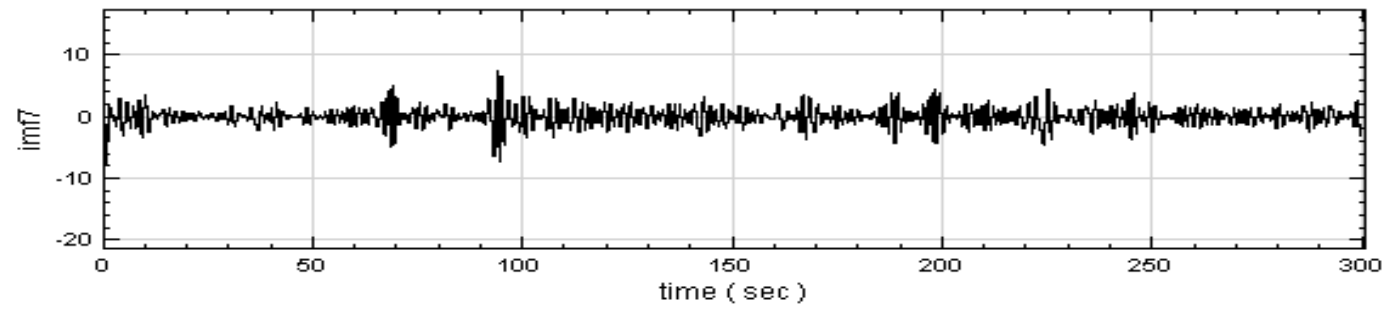
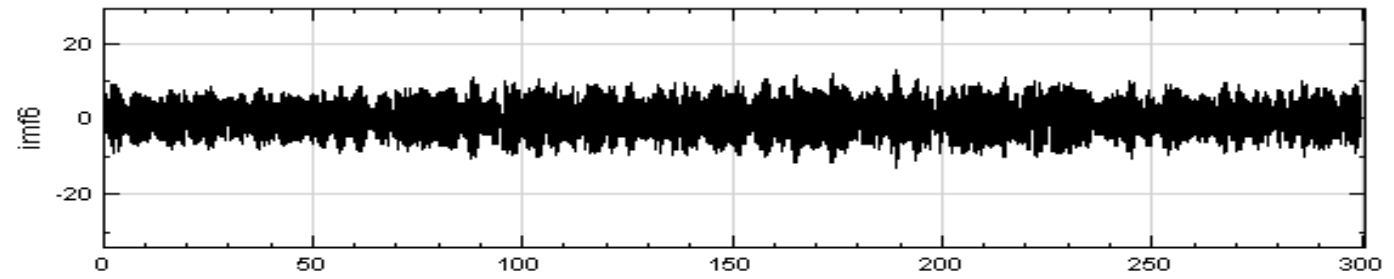
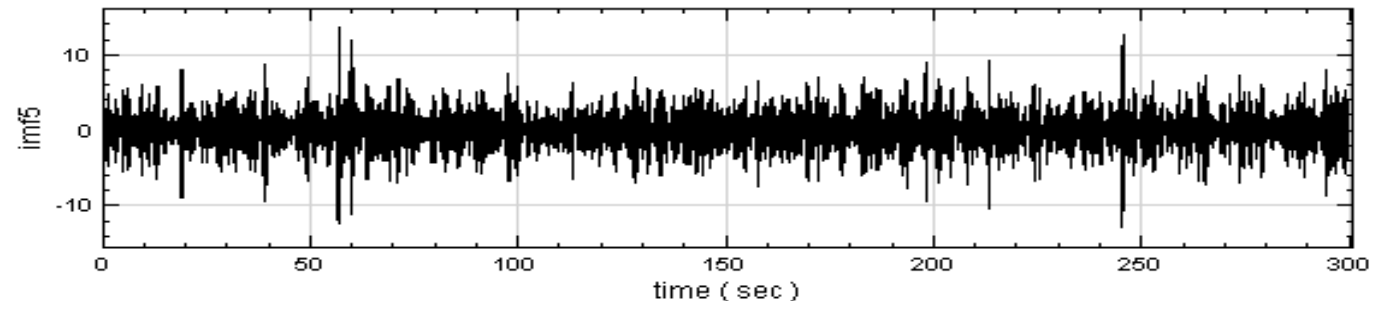
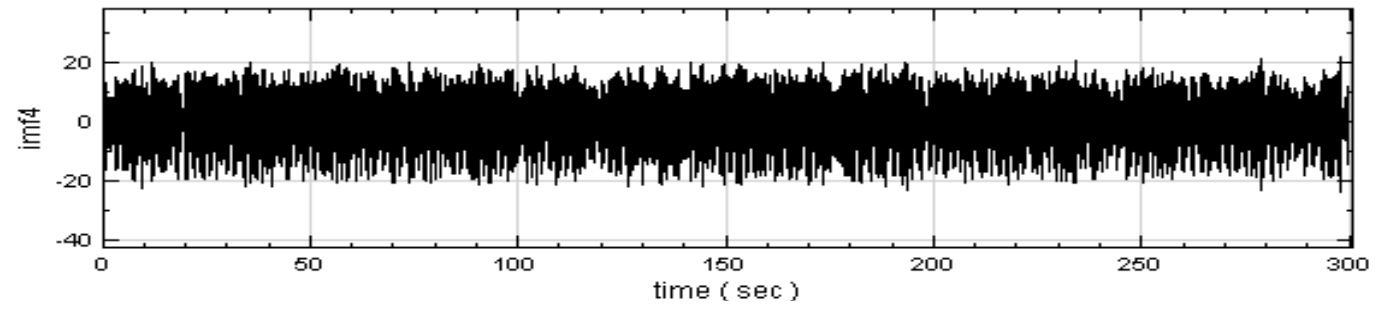
# An Example

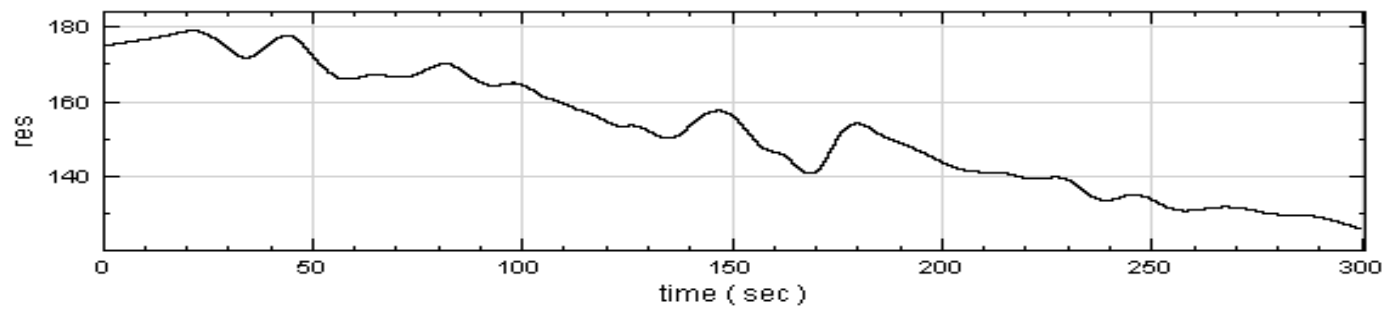
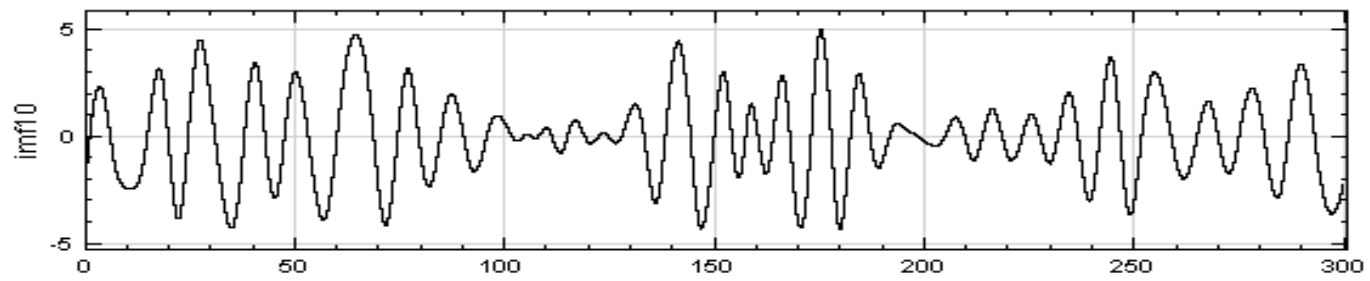
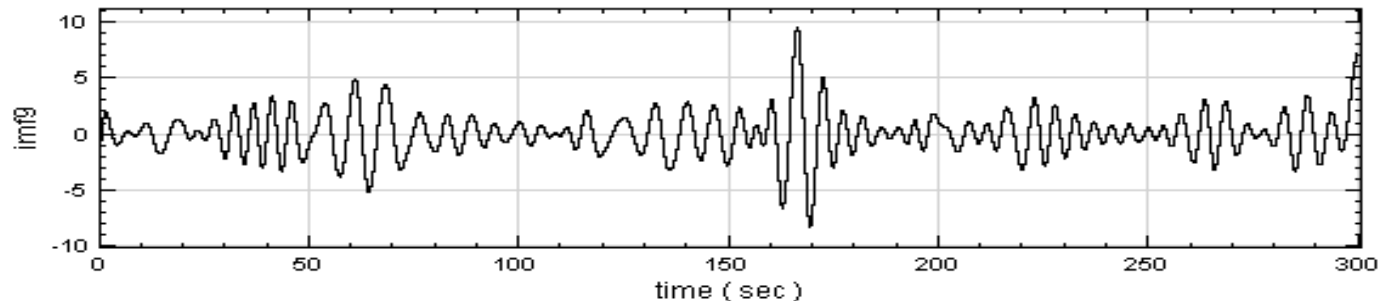
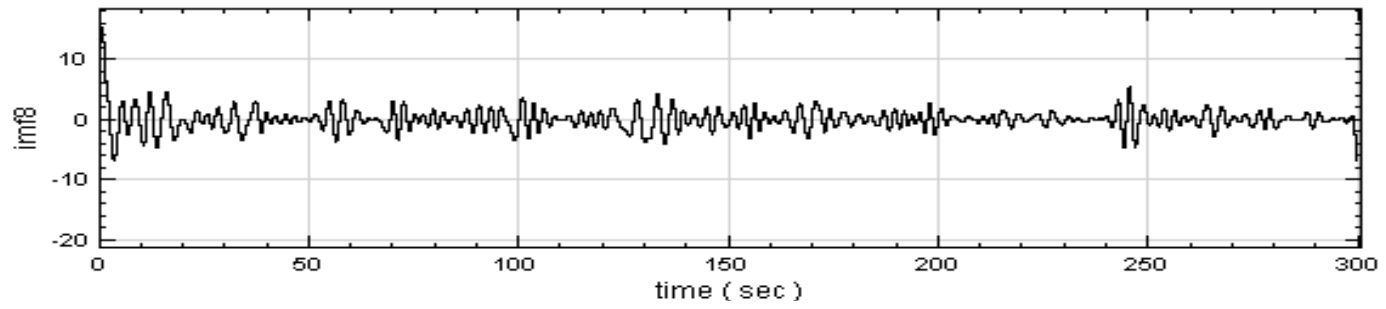
# 原始ECG訊號



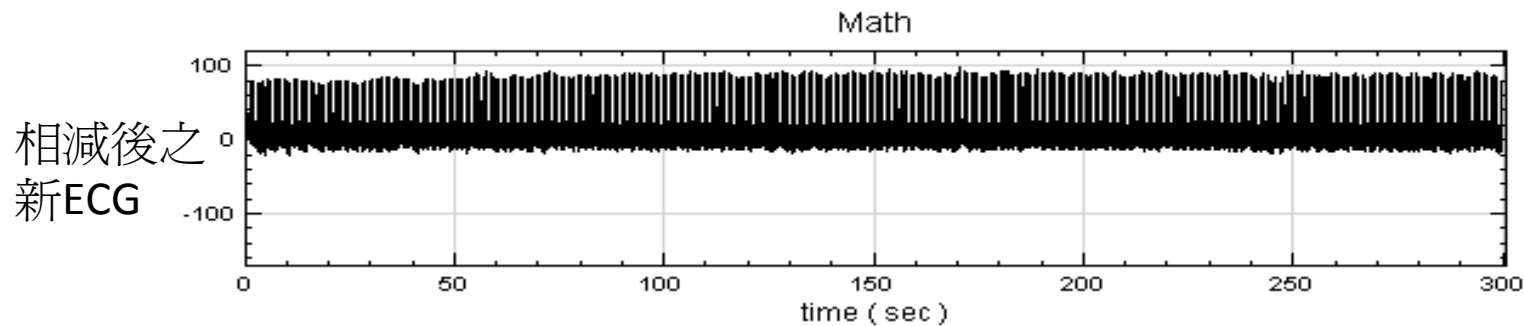
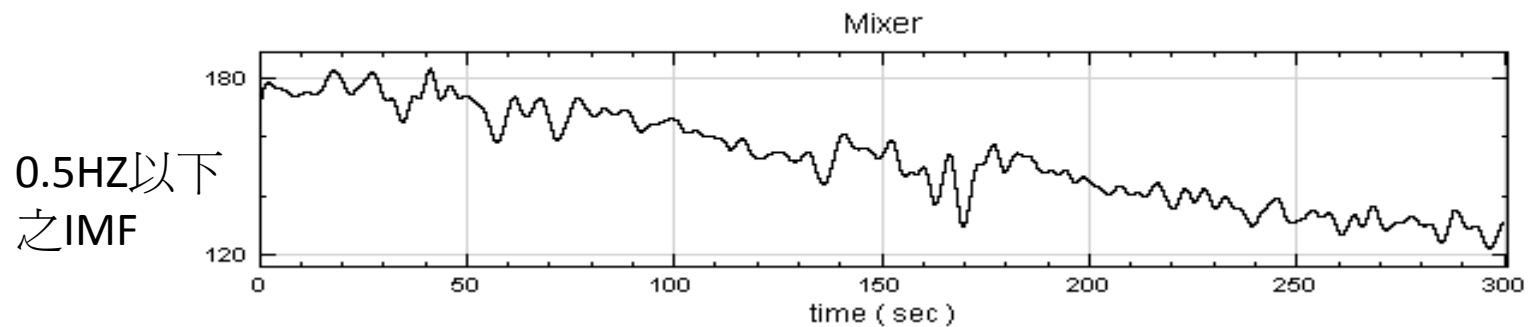
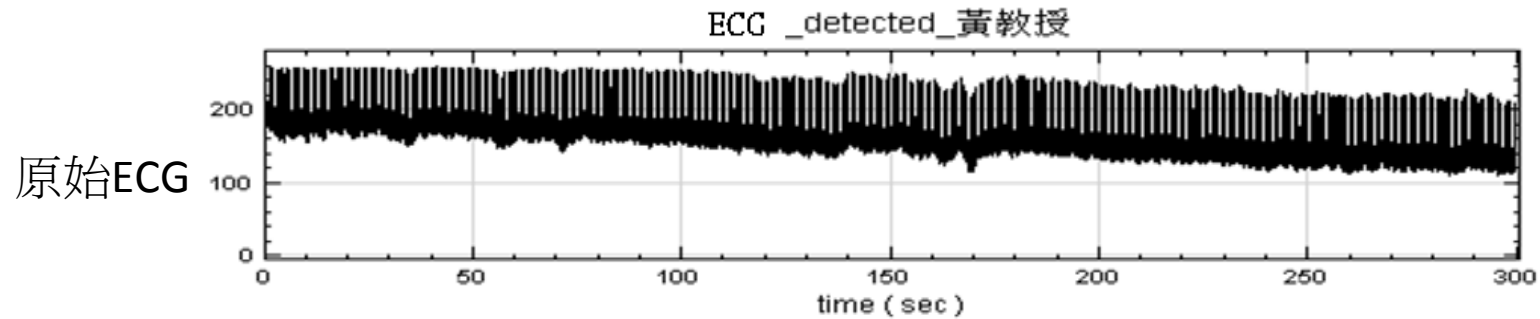
# EMD分解





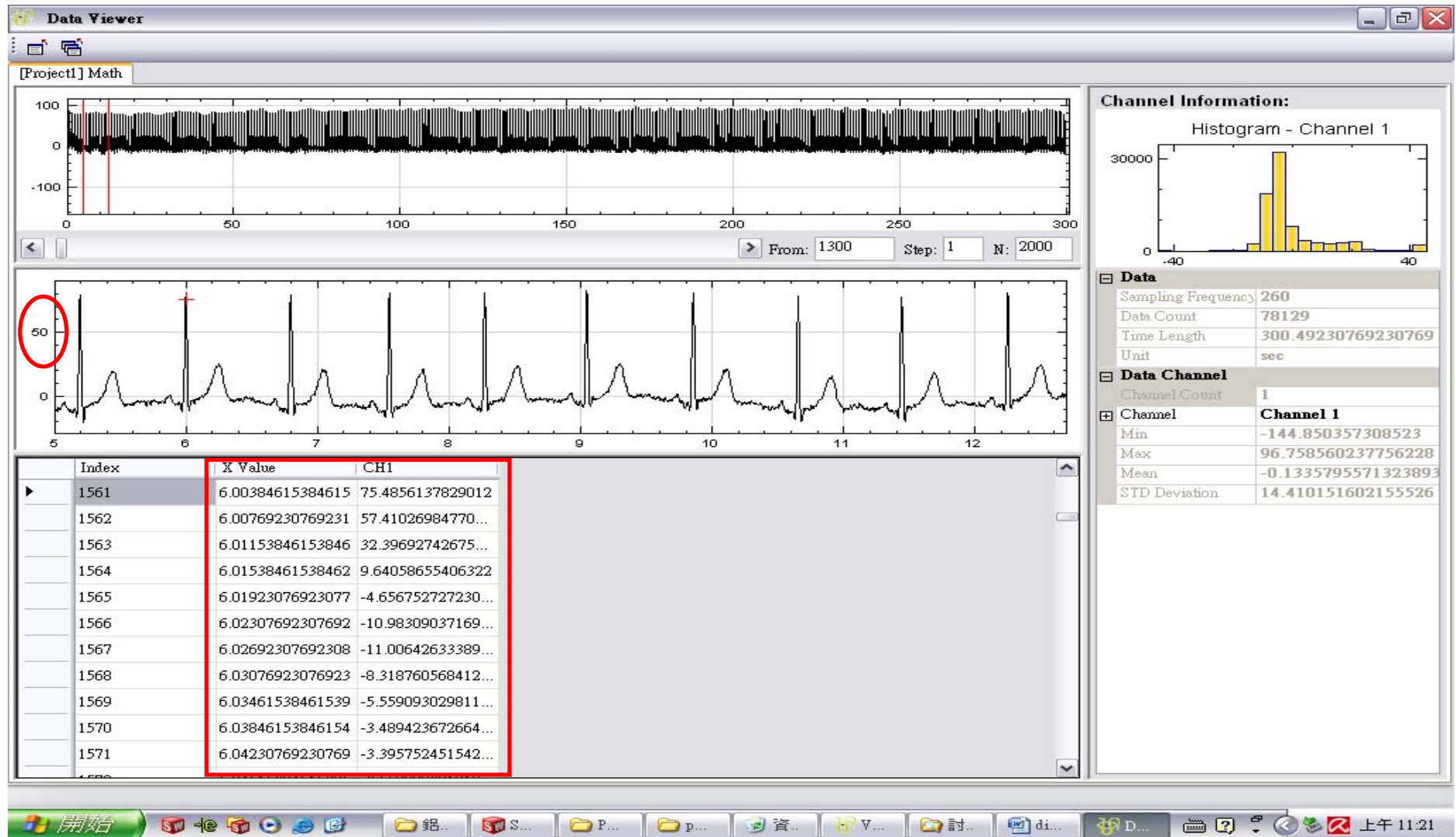


# 去除0.5Hz以下之IMF 8-IMF10



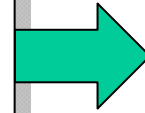


# Heart Beat



# HHT/Instantaneous Frequency

(Empirical Mode  
Decomposition)



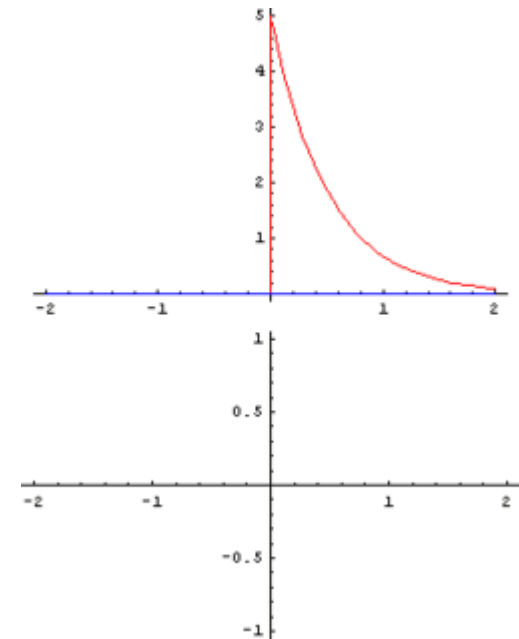
Hilbert Transform

**Hilbert Transform:**

$$\tilde{x}(t) = \mathbf{H} [x(t)] = P \int_{-\infty}^{\infty} \frac{x(u)}{\pi(t-u)} du$$

$$\tilde{x}(t) = x(t) * \left(\frac{1}{\pi t}\right)$$

$$z(t) = x(t) + j\tilde{x}(t)$$



# Instantaneous frequency

$$Y(t) = \frac{1}{\pi} PV \int_{-\infty}^{\infty} \frac{X(\tau)}{t - \tau} d\tau$$

we can have an analytic signal,  $Z(t)$ , as

$$Z(t) = X(t) + iY(t) = a(t)e^{i\theta t}$$

the instantaneous frequency as

$$\omega = \frac{d\theta(t)}{dt}$$

$Y(t)$ : Hilbert transform.

PV: Cauchy principal value.

$X(t)$ : arbitrary time series.

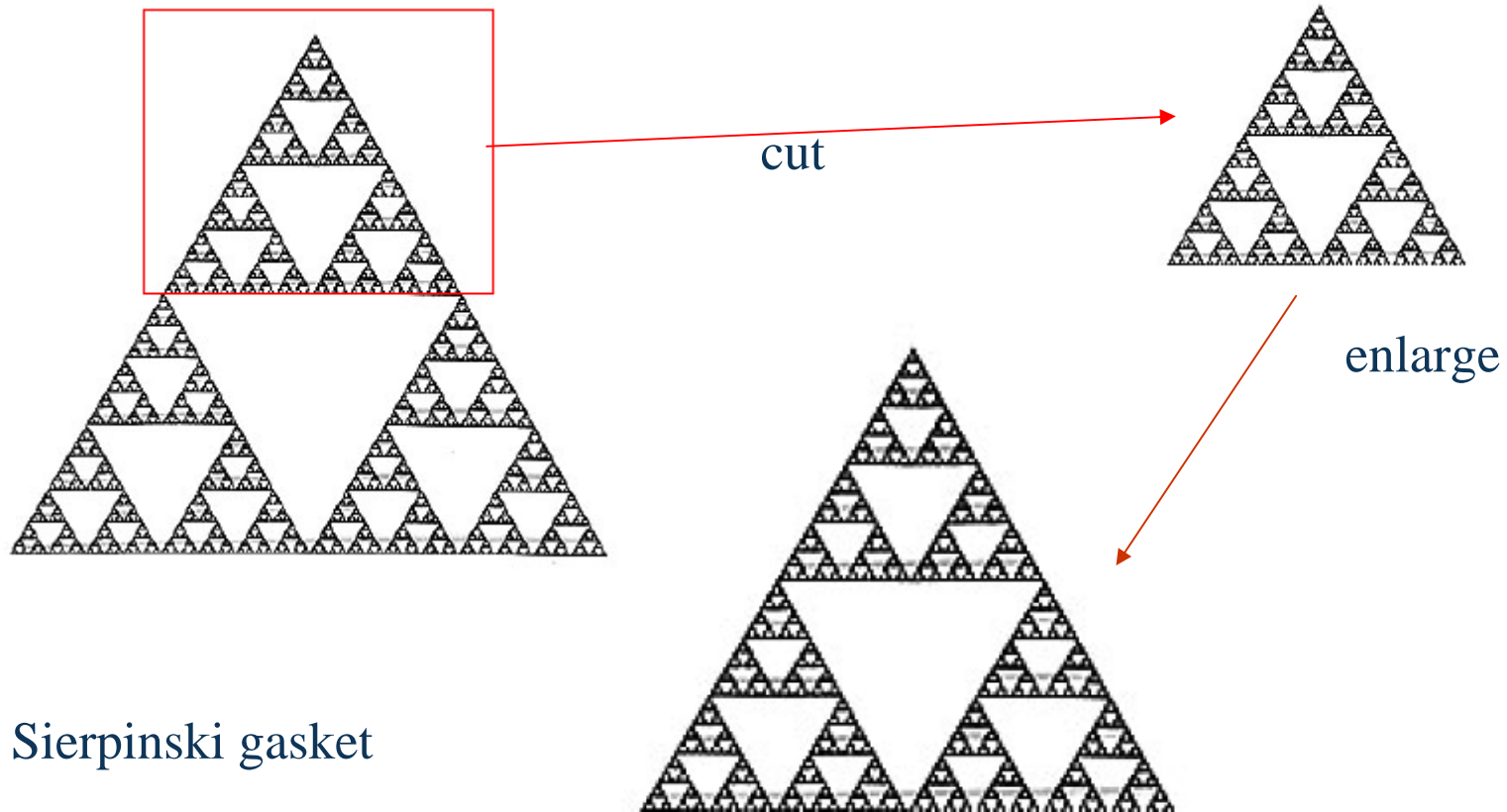
$$a(t) = \sqrt{X^2(t) + Y^2(t)}$$

$$\theta(t) = \tan^{-1}\left(\frac{Y(t)}{X(t)}\right)$$

# Outline

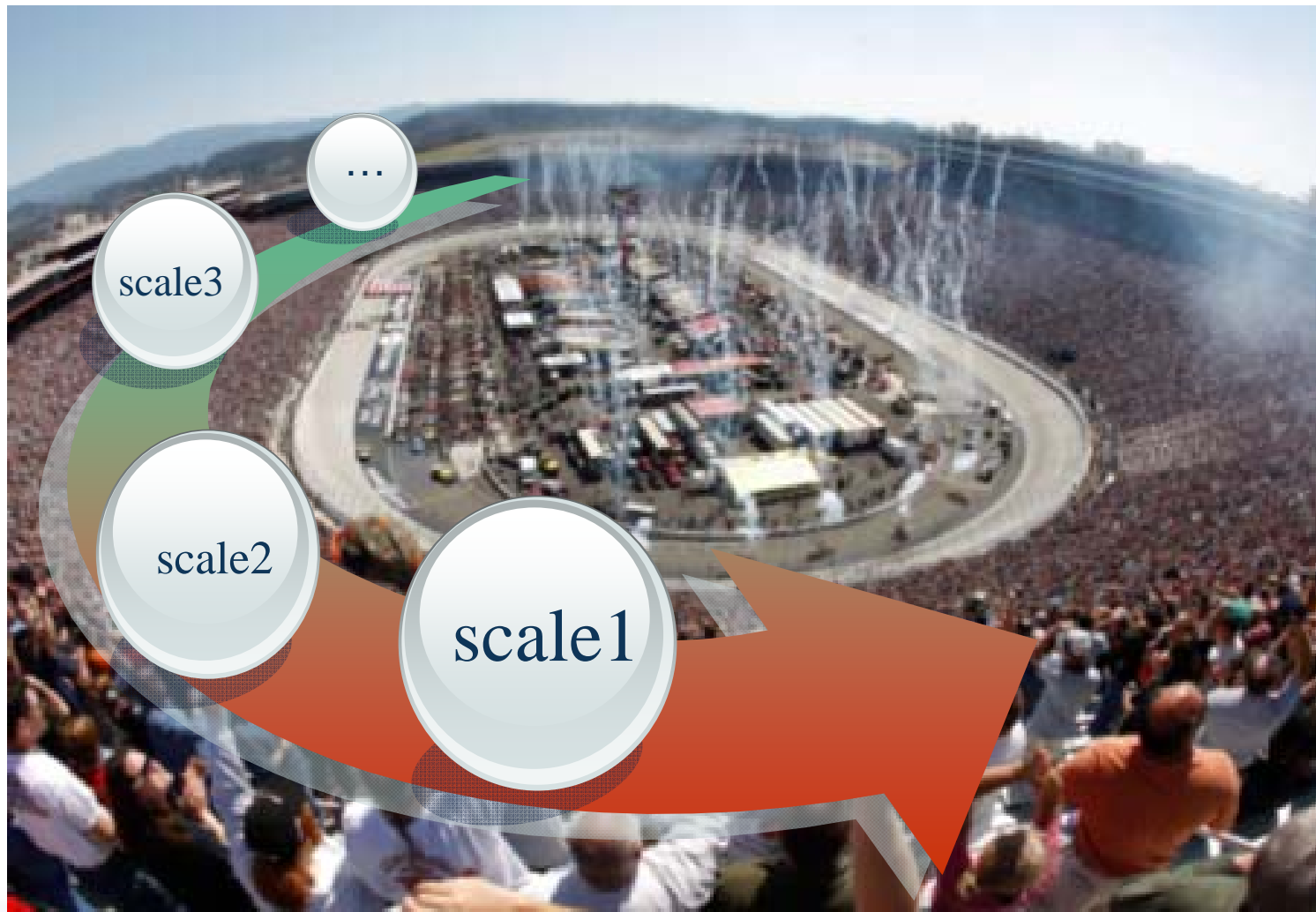
- 1 EMD/HHT
- 2 MSE (Multiscale Entropy) Fundamental
- 3 Torque Current Data Acquisition Analysis
- 4 Concluding Remarks

Self-similarity



Sierpinski gasket

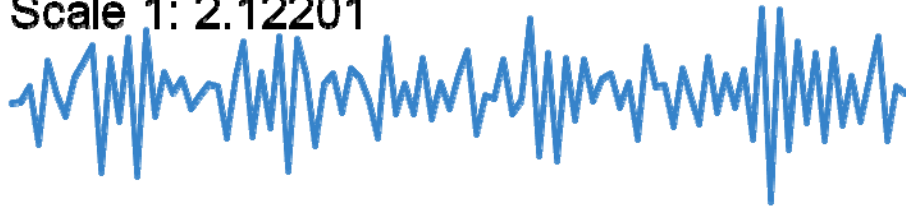
Crowd (chaos)



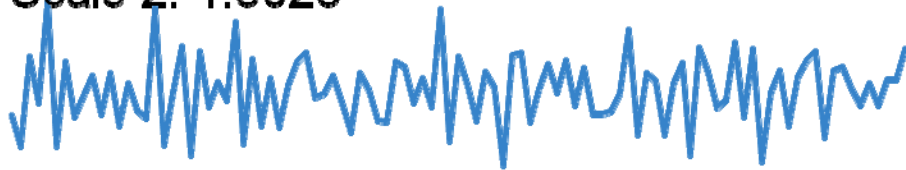


# Violet noise

Scale 1: 2.12201



Scale 2: 1.5028



Scale 3: 1.0744

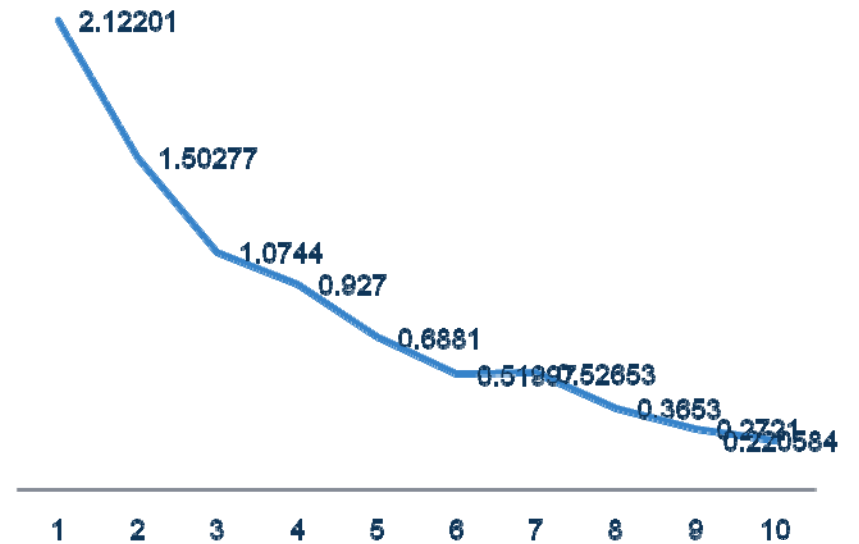


⋮

Scale 10: 0.22058



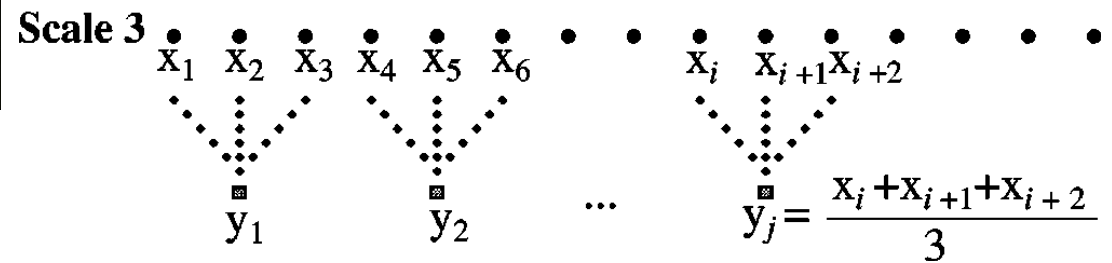
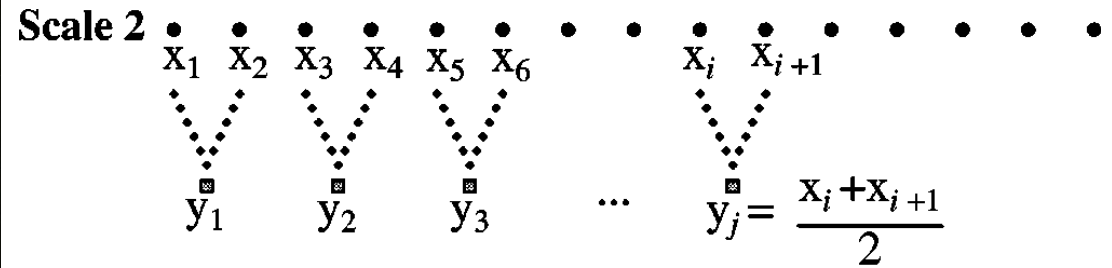
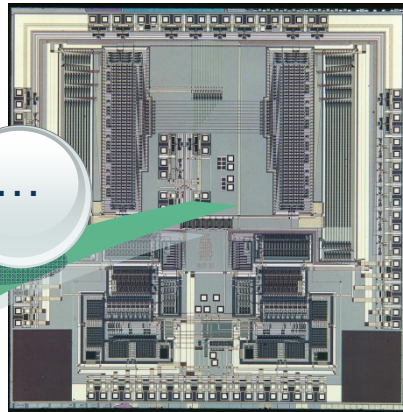
Sample entropy



Scale factor



# Circuit (complex)



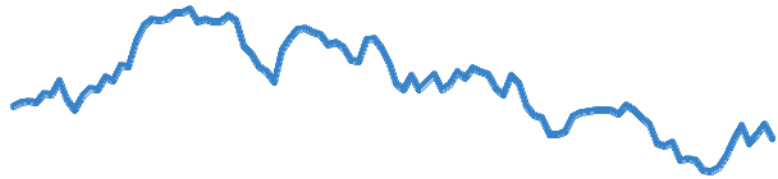
scale3

scale2

scale1

# Brown noise

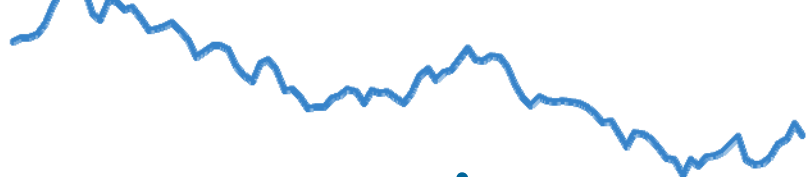
Scale 1: 0.45062



Scale 2: 0.61853

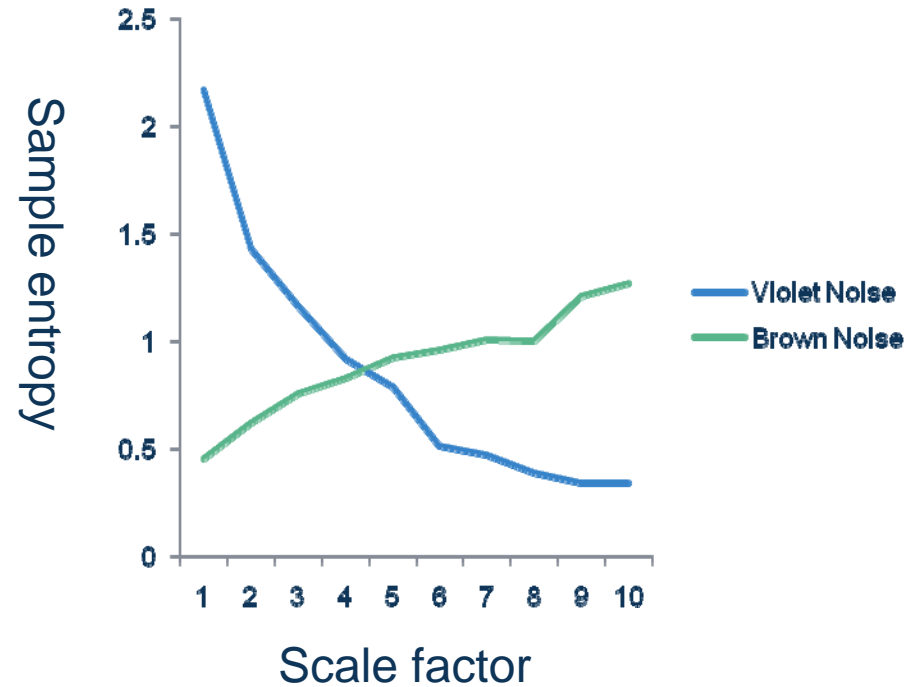
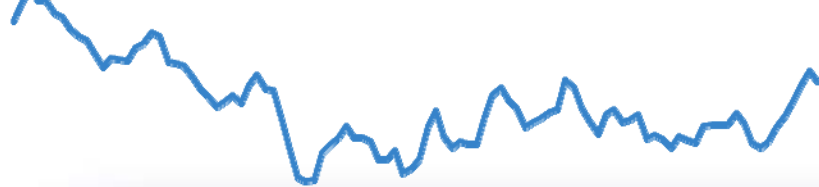


Scale 3: 0.75445



⋮

Scale 10: 1.2657



# Outline

- 1 HHT (Hilbert Huang Transform) Fundamental
- 2 MSE (Multiscale Entropy) Fundamental
- 3 Torque Current Data Acquisition and Analysis
- 4 Concluding Remarks

## 實驗設備介紹



### ASDA-A2 臺達電驅動器

- 內建電子凸輪功能，可以滿足多重軸動控制
- 由調機軟體可即時監視馬達運轉時多項變量



### LNC-310i 寶元控制器

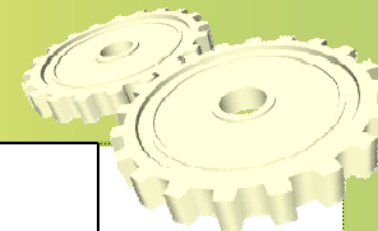
- 3軸同動，Pulse Command位置迴路控制
- 具有分段插補之功能



### LS-177 海德漢光學尺

- Pulse command
- 提供AB相四倍解波，解析度0.25um

# 實驗架構-單螺帽導螺桿



滾珠直徑 Ball Diameter	6.35mm
螺桿根徑 Root Diameter	30.91mm
最大動負荷 Dynamic Load	3750kgf
最大靜負荷 Static Load	9542kgf
挫曲負荷 Buckling Load	70288kgf

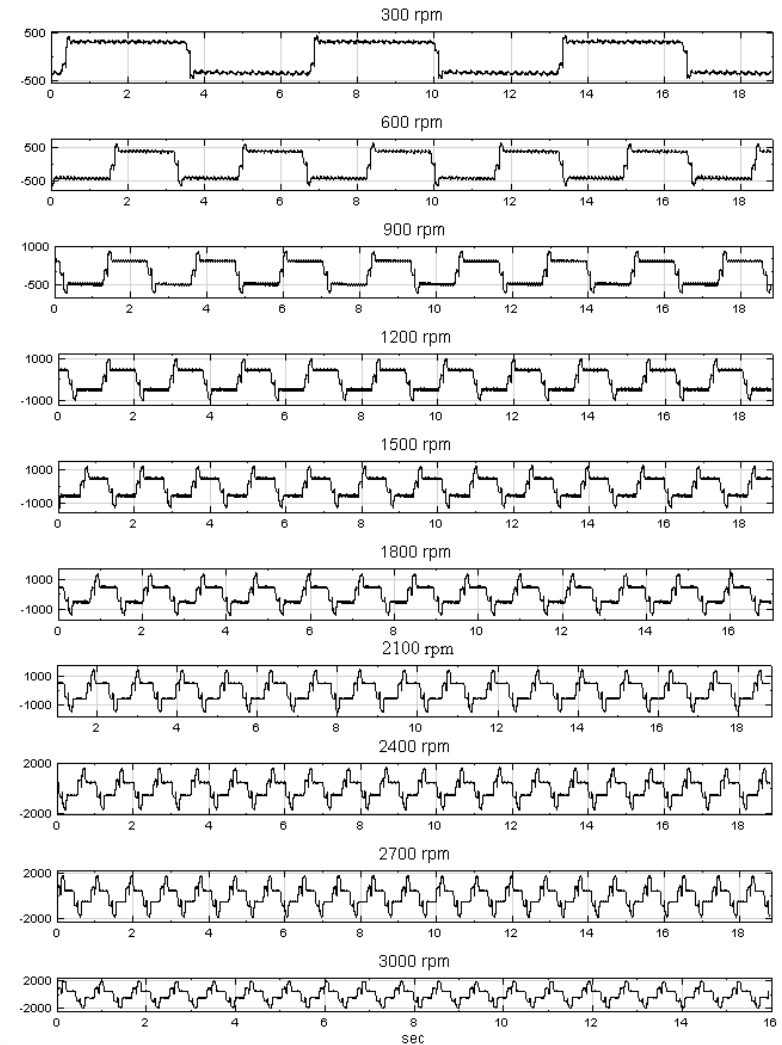
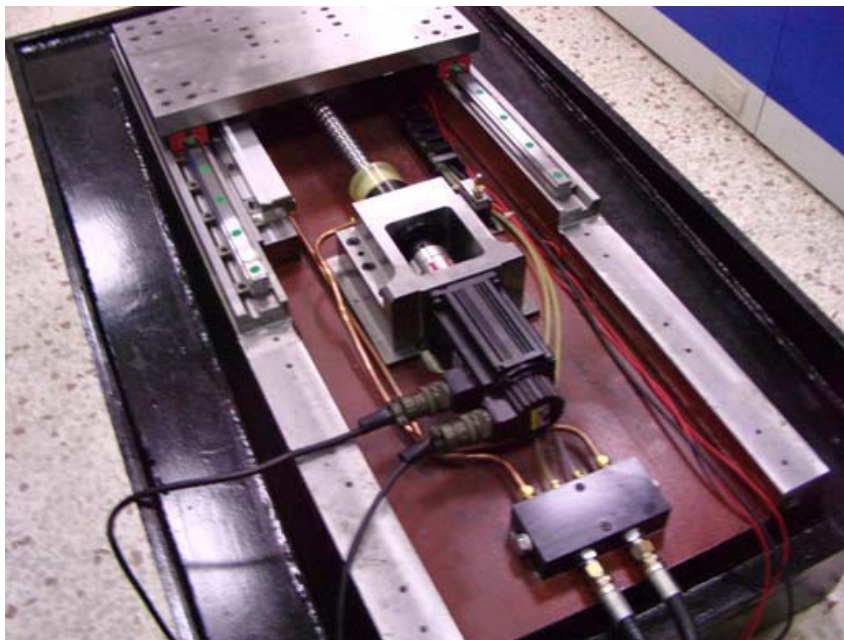
與最大動負荷之比例	實際預壓之量值
2%	75kgf
3%	113kgf
<b>4%</b>	<b>150kgf</b>
5%	188kgf
6%	250kgf



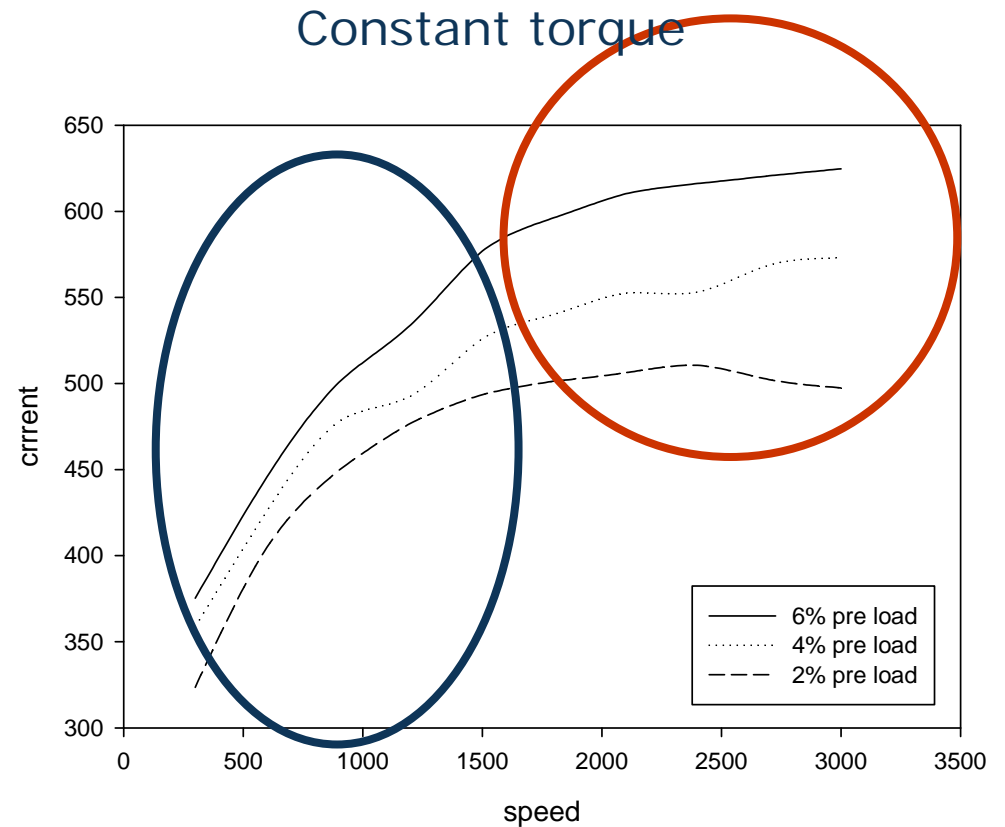
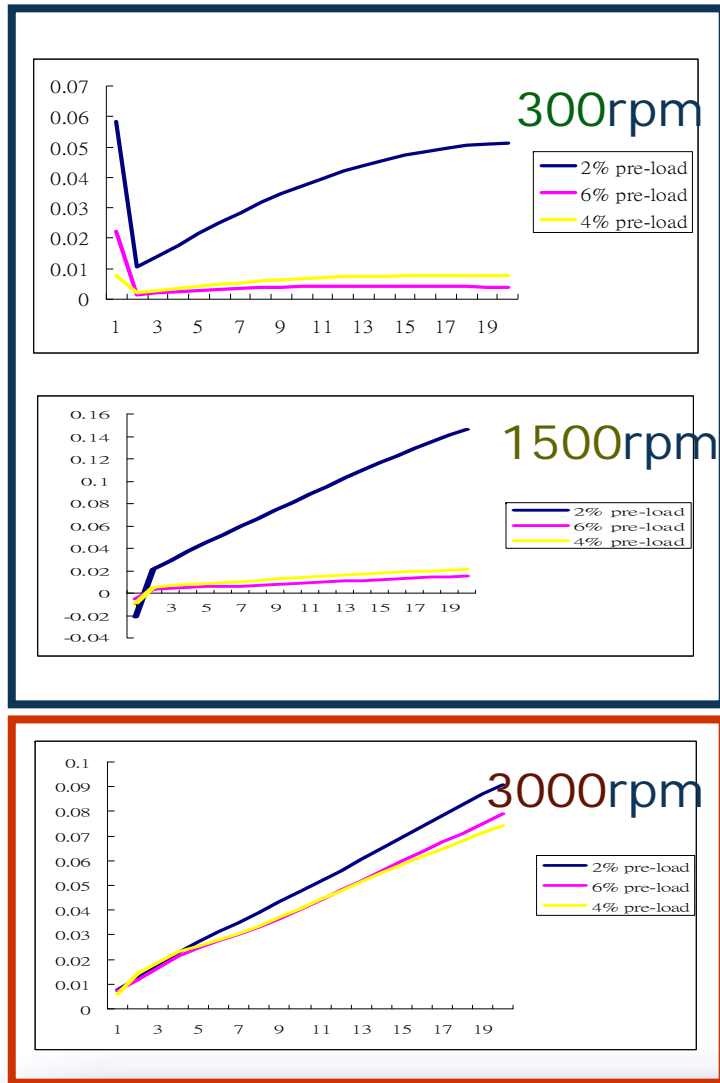


## Experimental tests

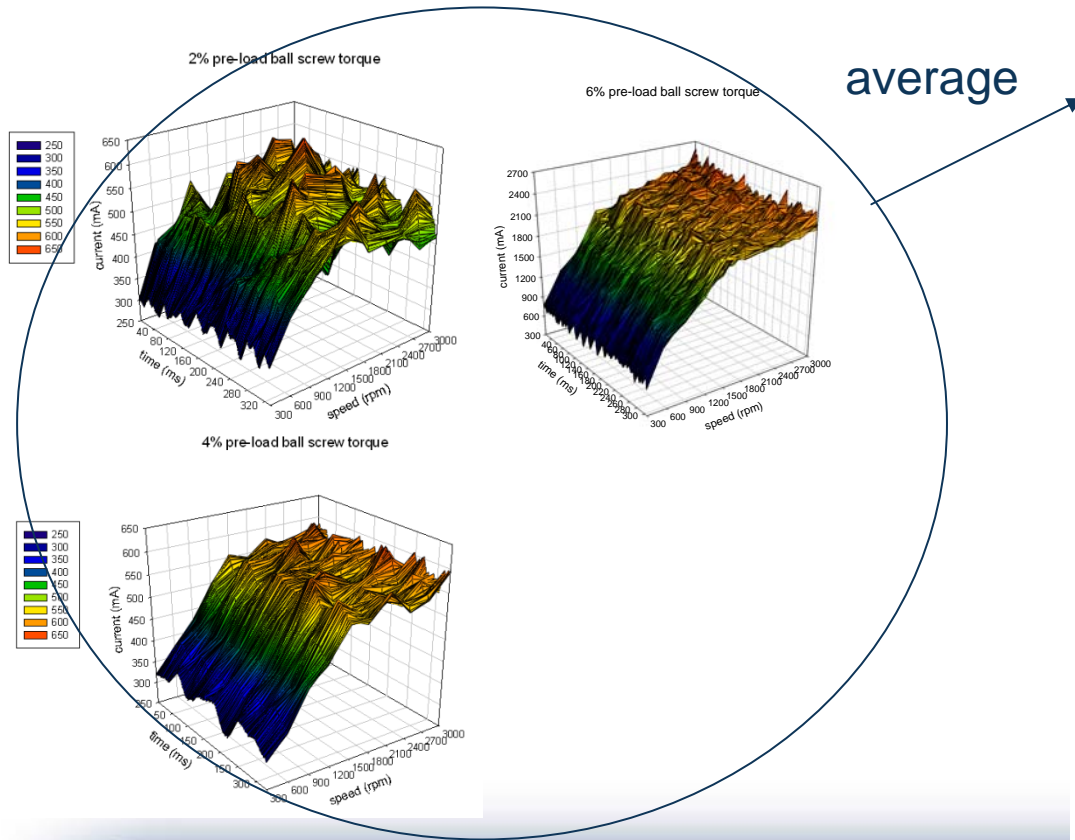
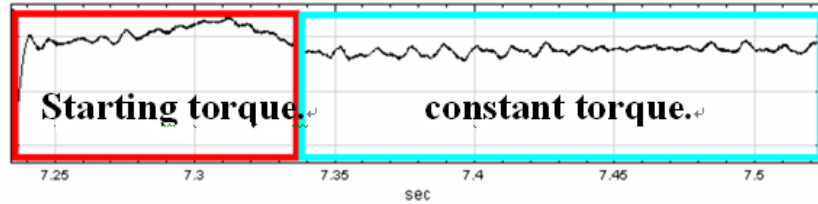
The table is commanded for the traveling distance of 270mm with reciprocating movement. Driven by a servo motor followed by a coupler and single nut ball screw table with different preloads.



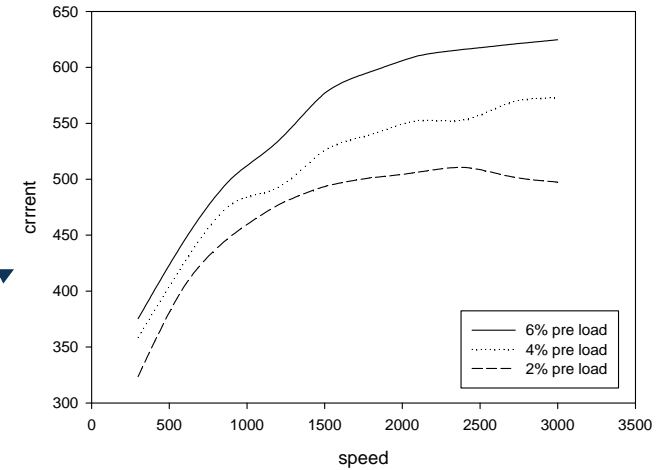
# Steady state



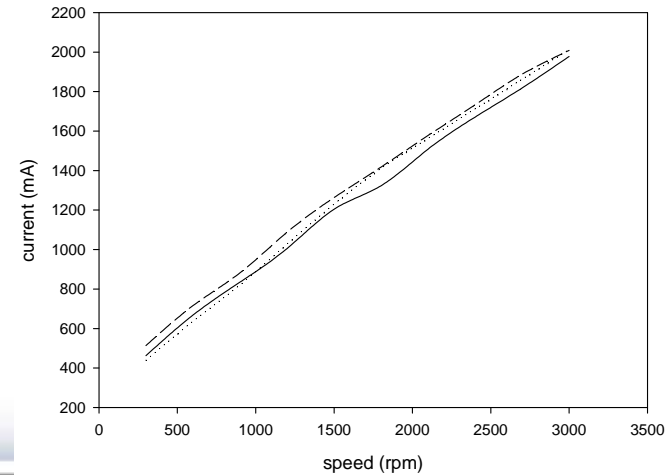
# Data observe



Constant torque

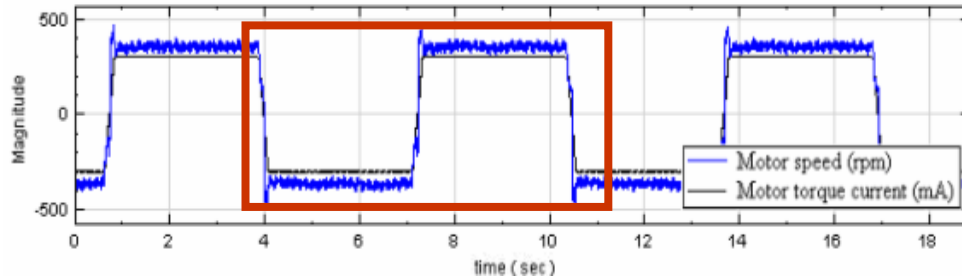


Starting torque

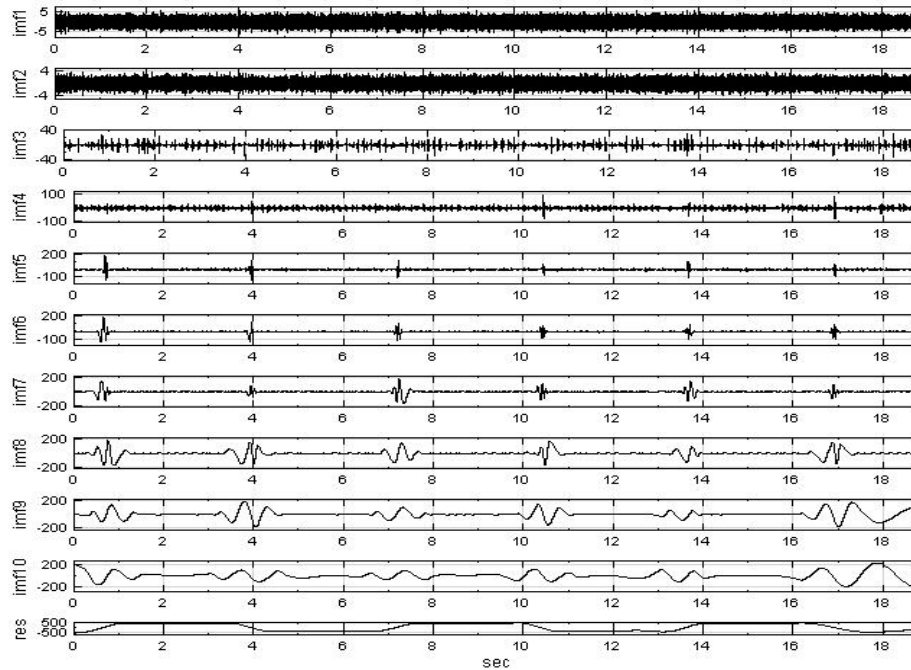
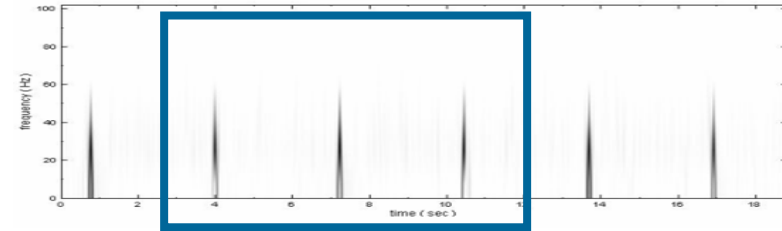


# HHT process

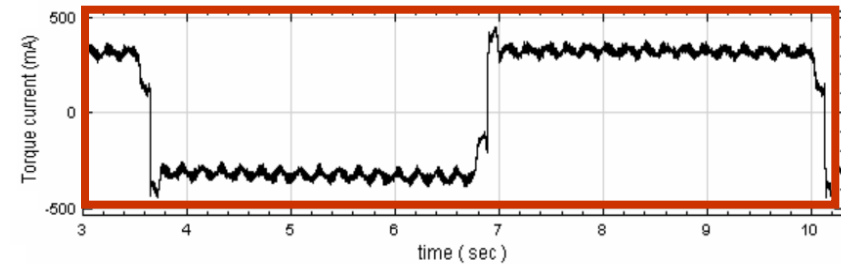
(a) Motor speed and torque current



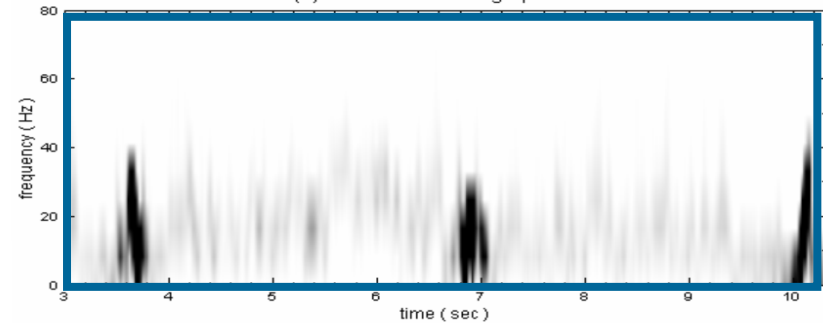
IMF7-Hilbert Huang Spectrum



(a) Motor speed in 300rpm

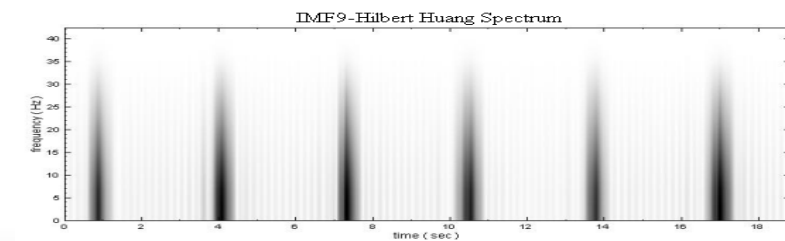
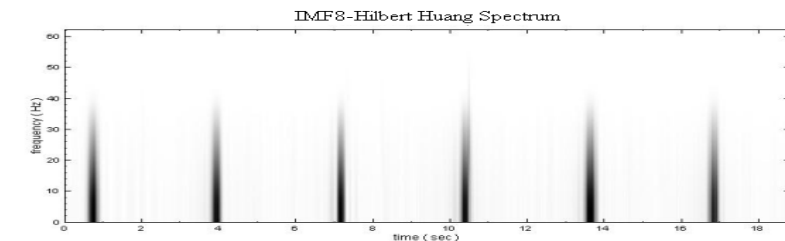
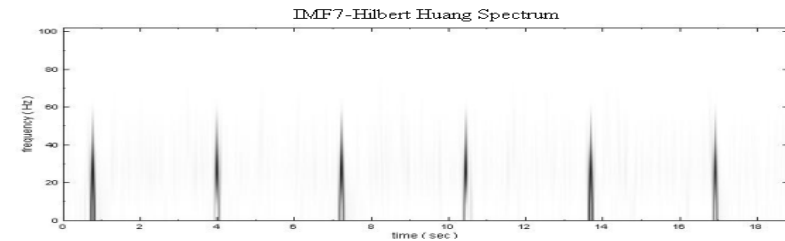
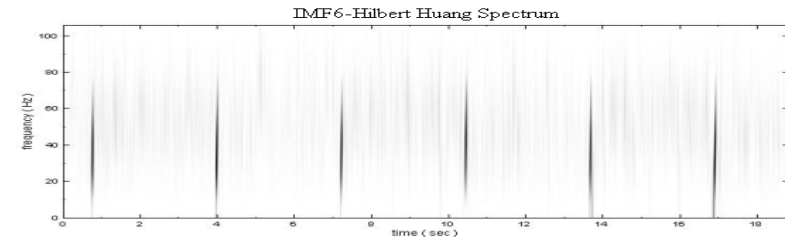
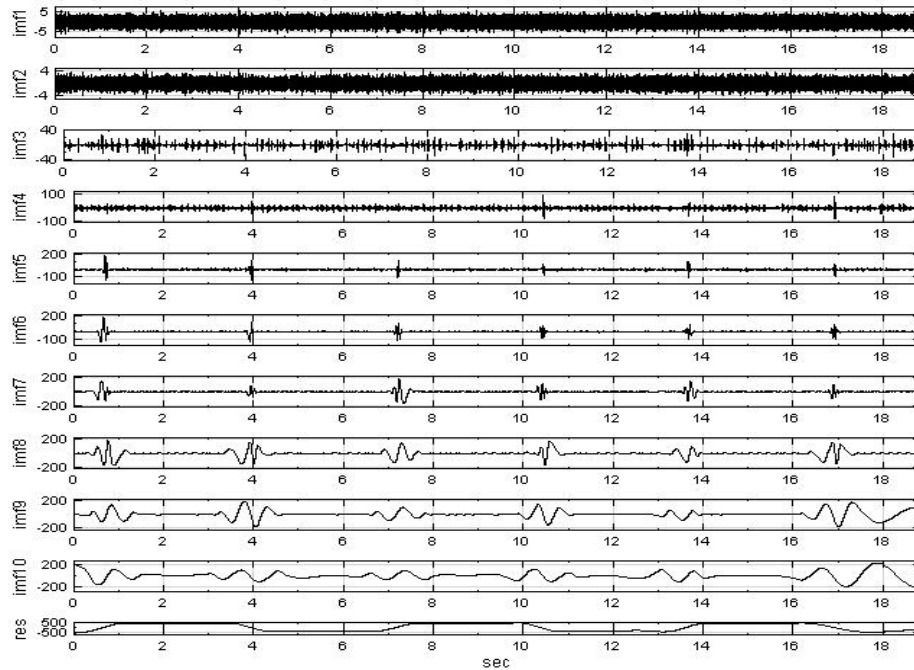
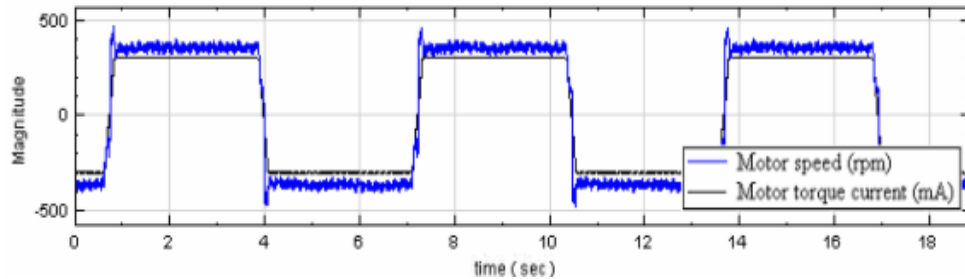


(b) IMF7-Hilbert Huang Spectrum



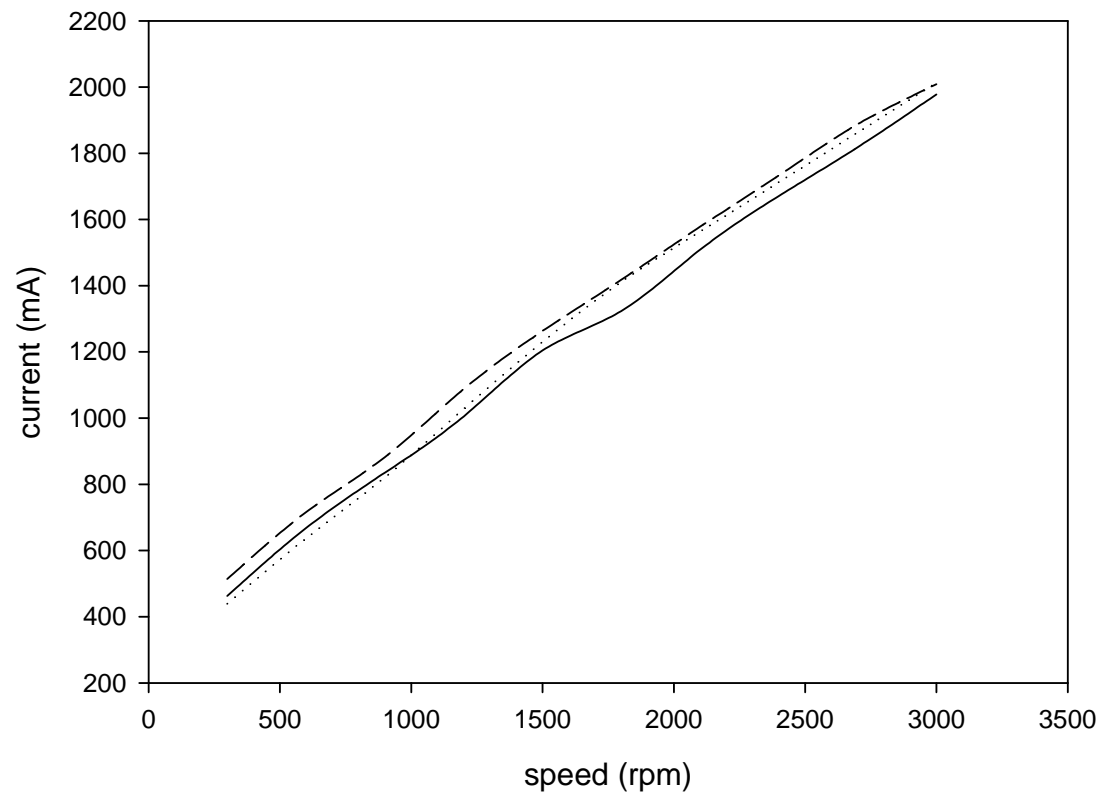
# HHT process

(a) Motor speed and torque current

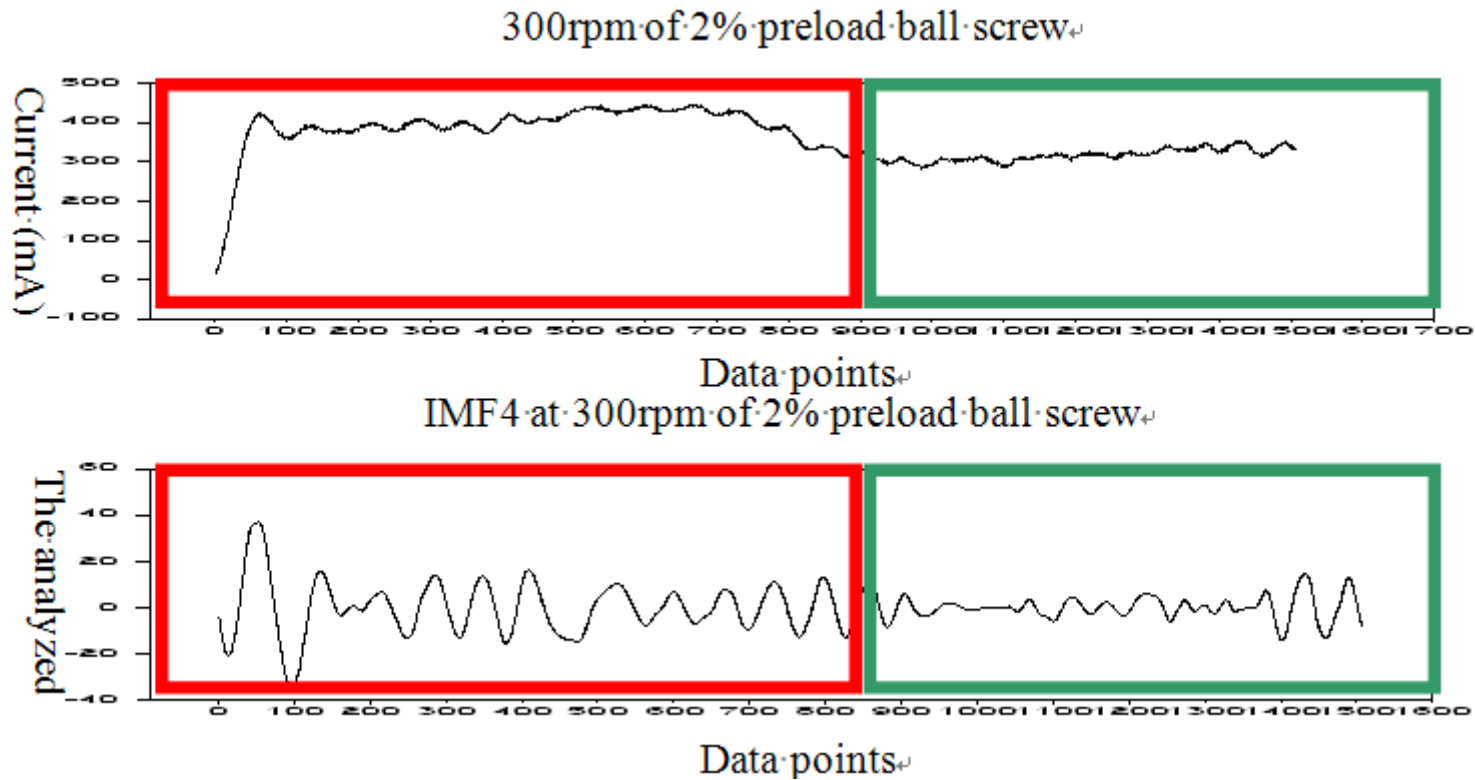




- ❖ In Figure we enlarged the torque currents marked in red and green from the original torque current history and followed the IMF4 of EMD process. The red regime implies the time span of starting torque current whereas the green regime indicates the time span of constant rpm for quasi-steady-state torque current. **The spiky amplitude for starting torque of 2 %, 4 %, and 6 % preloads in red is approximately 32mA, 64mA, and 96mA**, respectively. This correlates to the specific time duration for instantaneous required torque of different preloads in Figure 7(b).



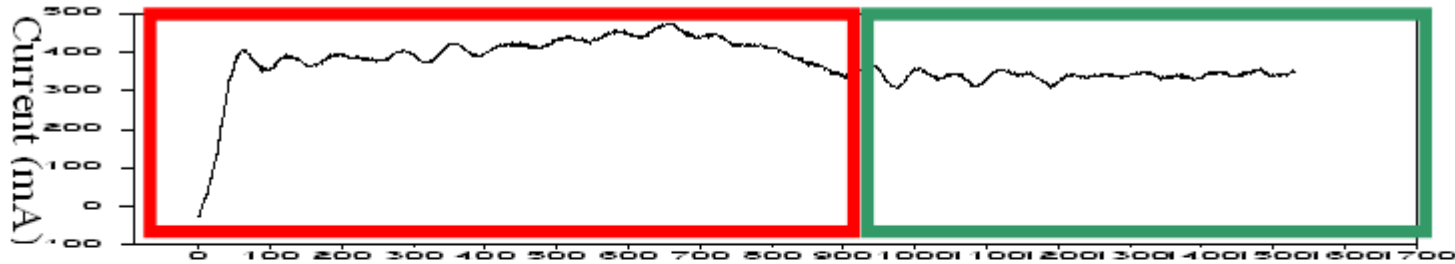
# IMFs Identify



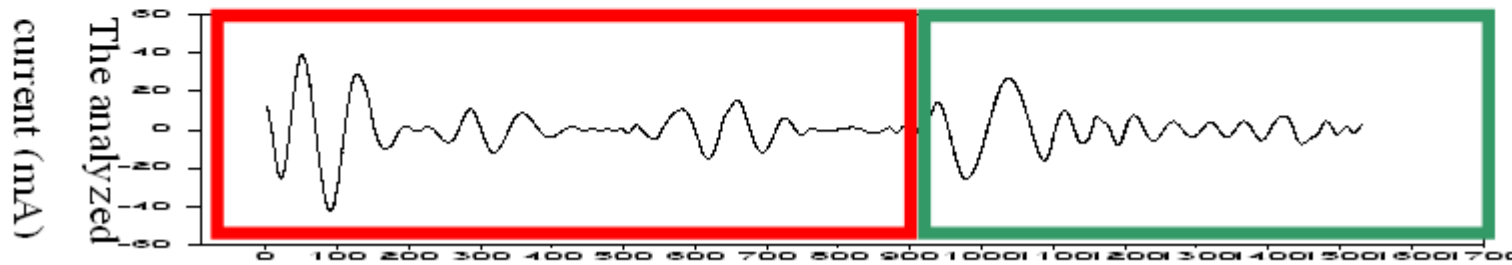


# IMFs Identify

300rpm of 4% preload ball screw

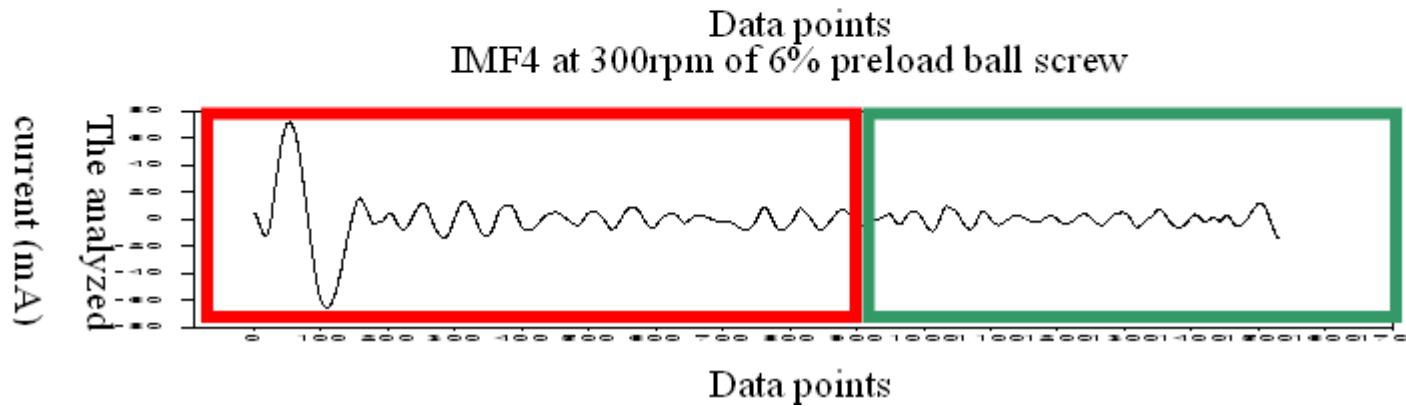
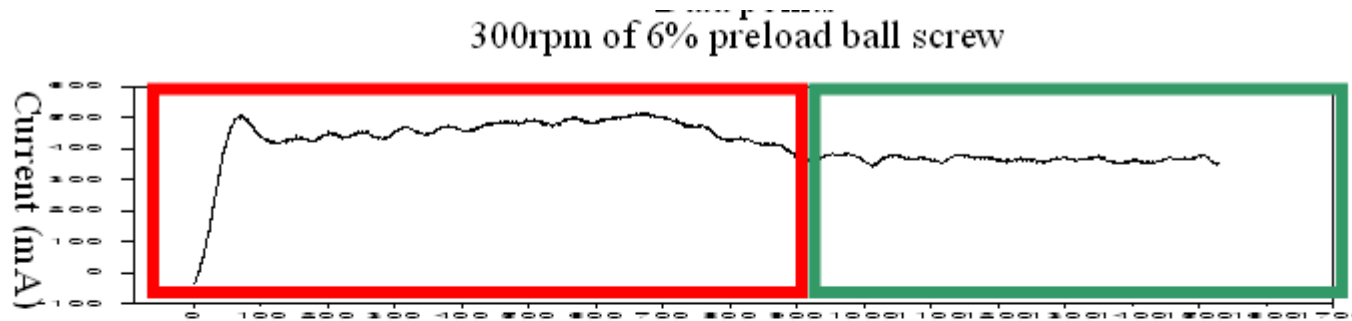


Data points  
IMF4 at 300rpm of 4% preload ball screw

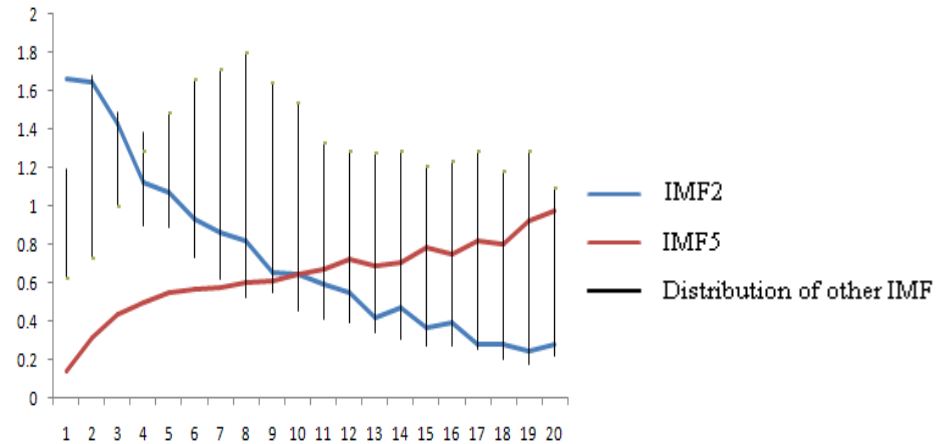
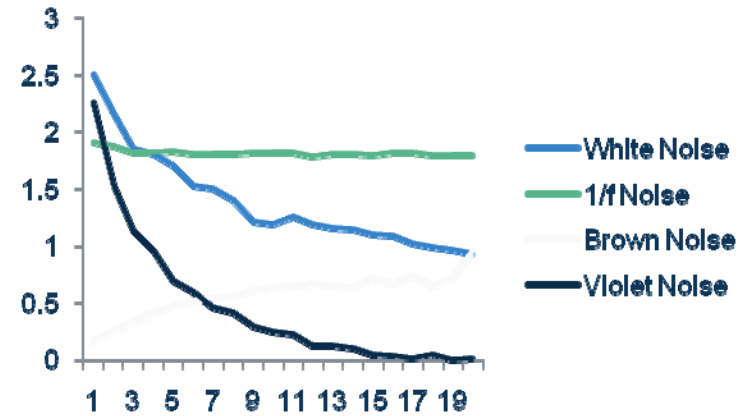
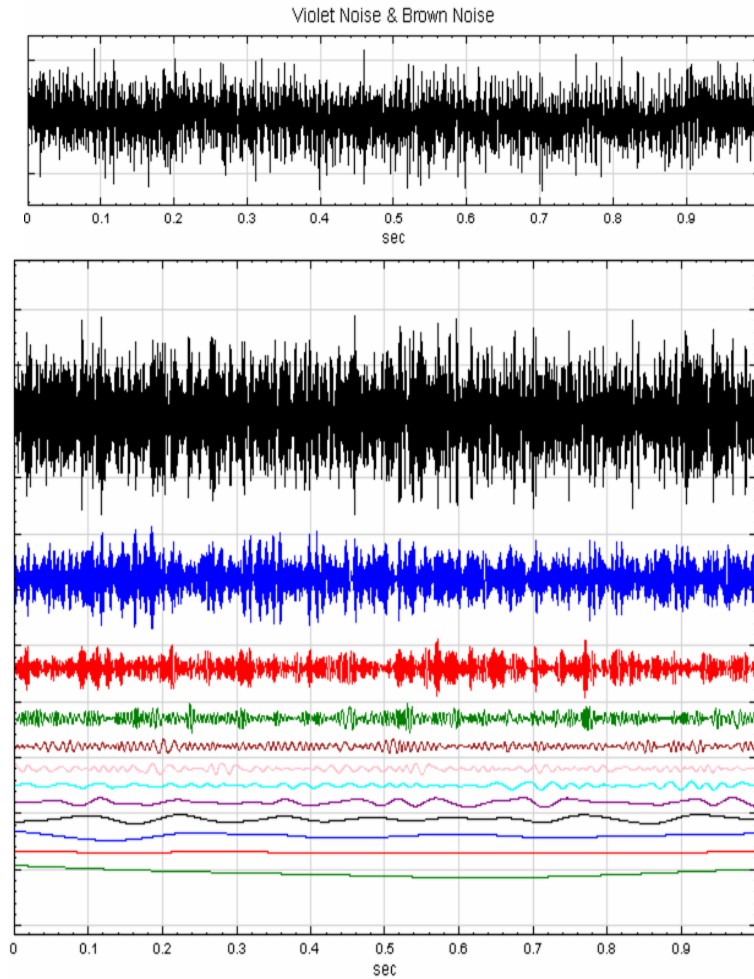


Data points

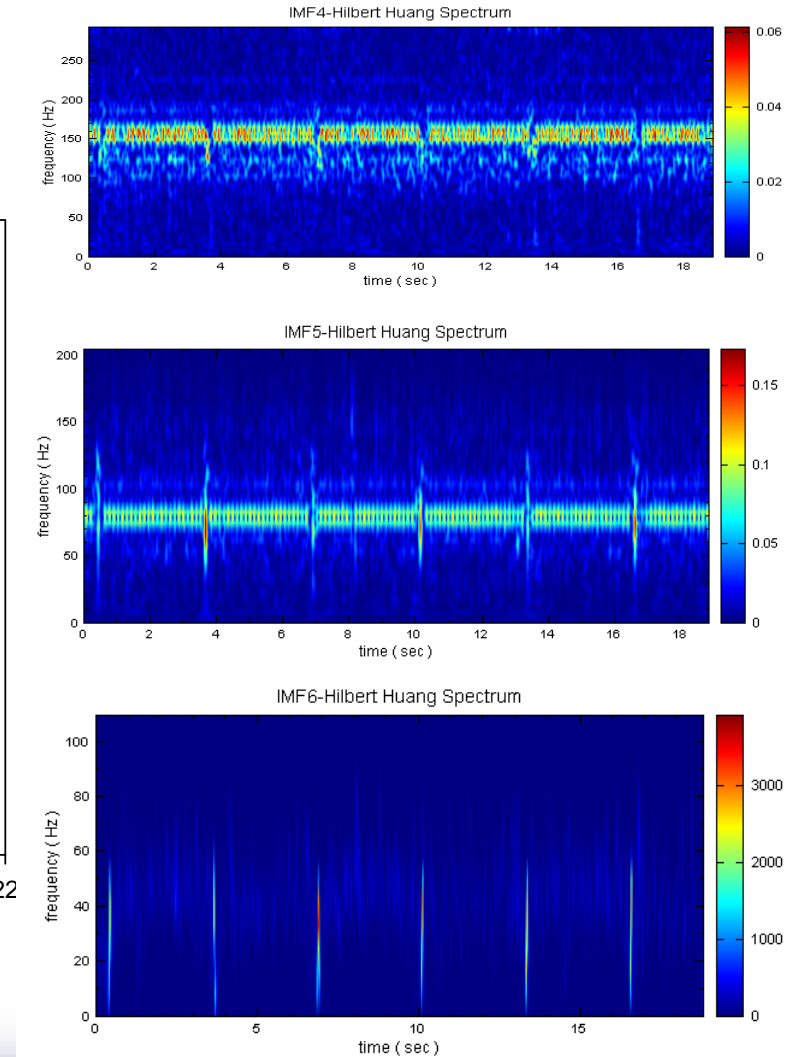
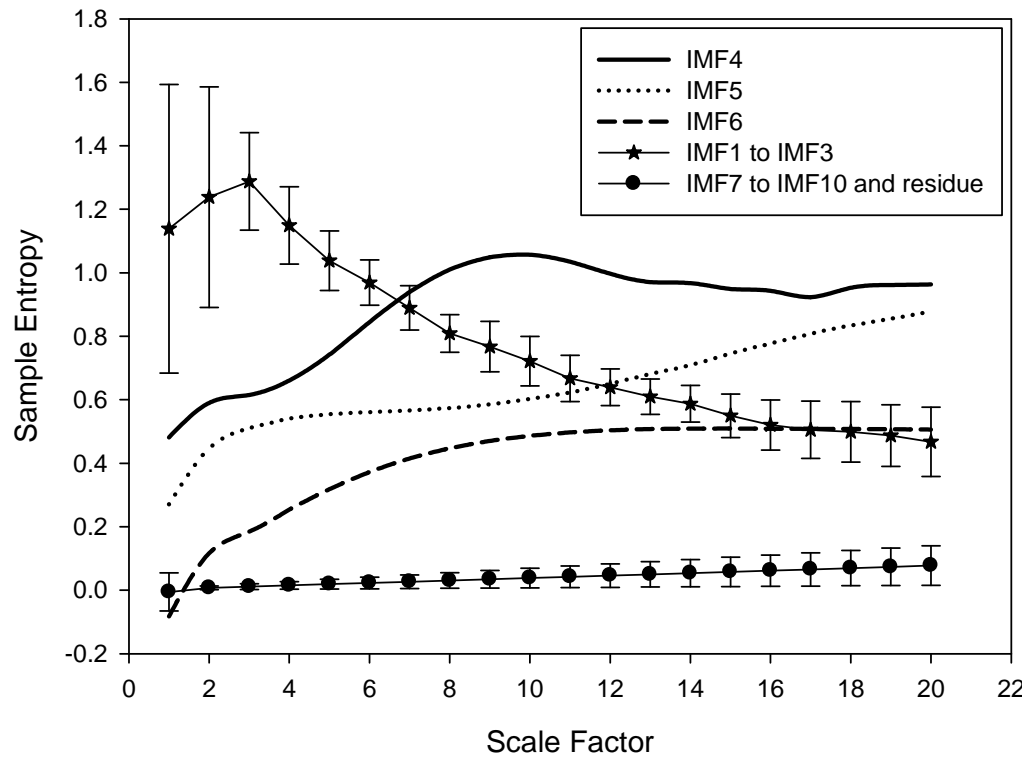
# IMFs Identify



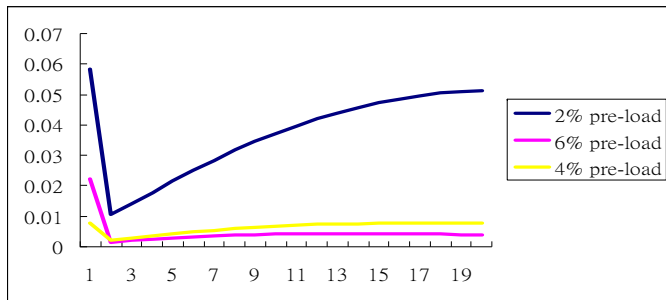
# HHT&MSE



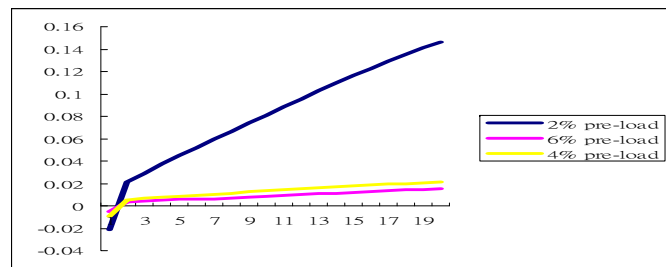
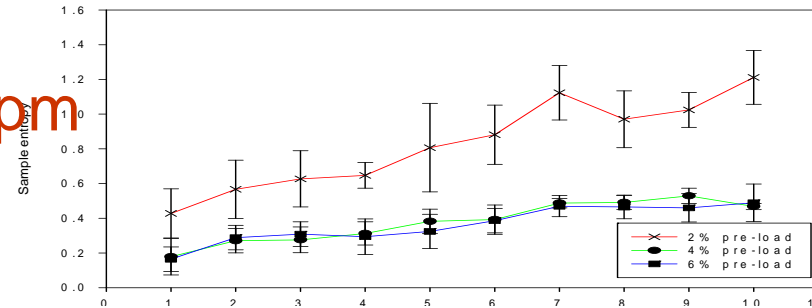
# IMFs Identify



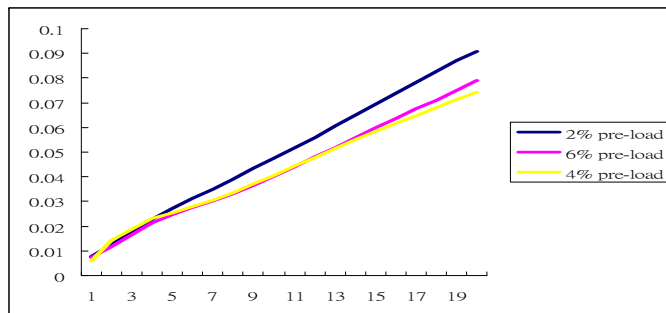
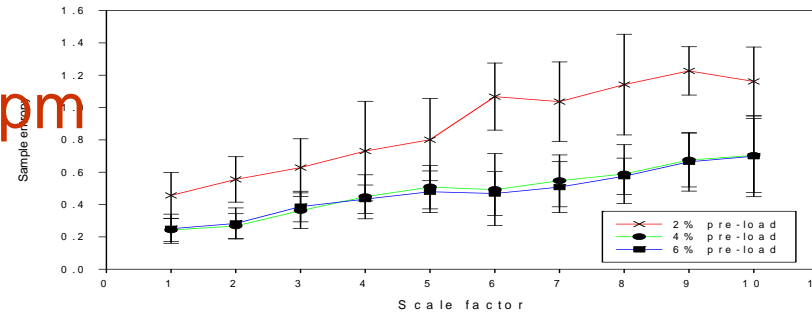
# MSE



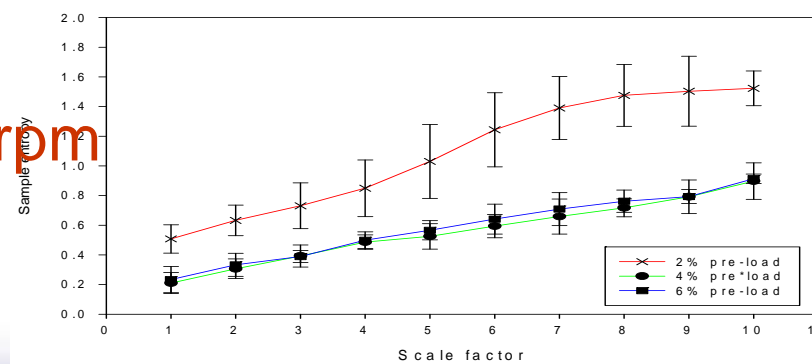
300rpm

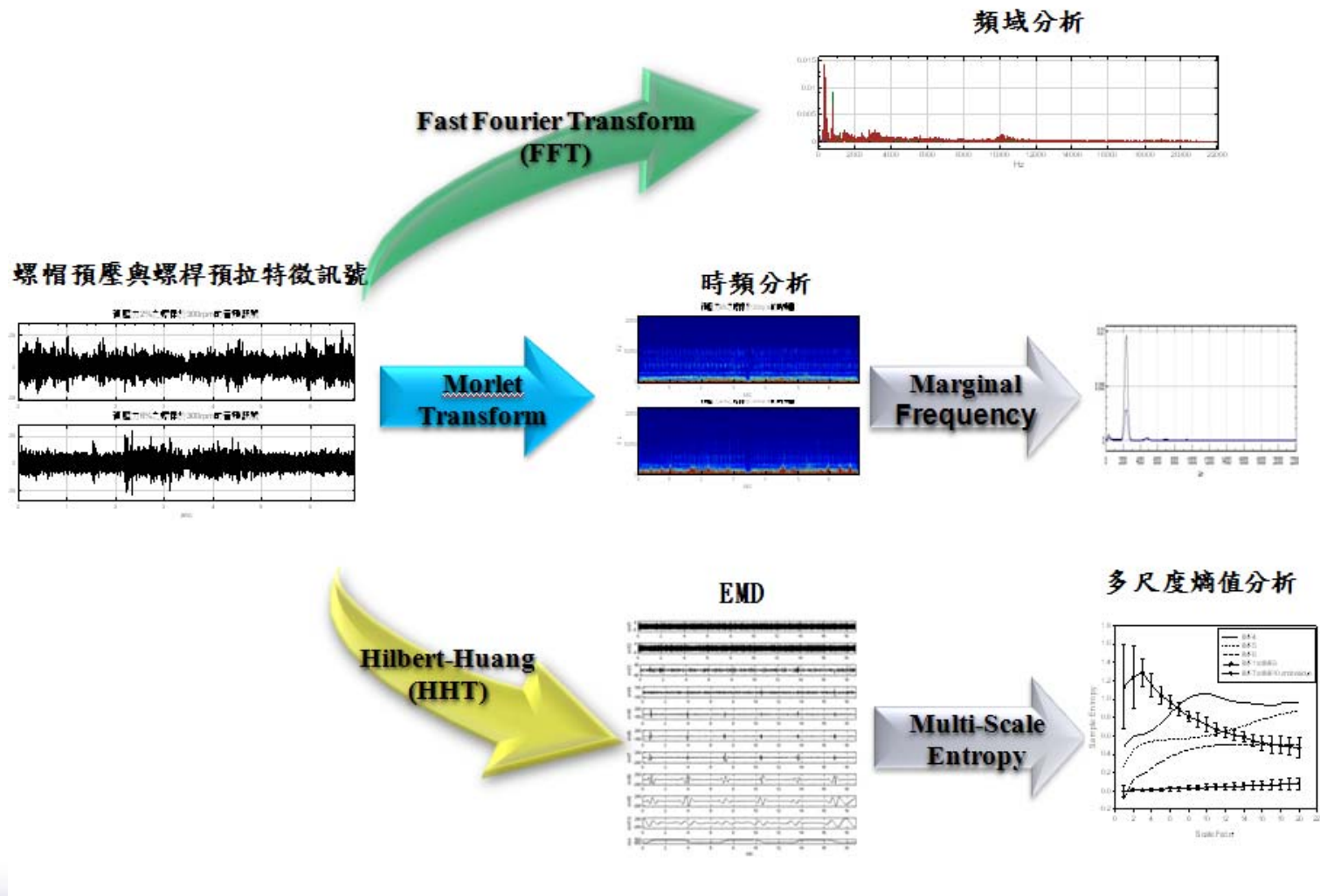


1500rpm



3000rpm





## 智能化螺帽預壓監測

- ❖ 改變不同螺帽預壓量的滾珠導螺桿，透過馬達驅動器，擷取機台往覆運動時馬達的驅動電流訊號，再以經驗模態分析法(EMD)拆解訊號，找出預壓變化的特徵信號，並選擇適當內稟模態函數(IMF)，進行希爾伯特轉換(HHT)，觀測其時頻圖，判定發生預壓失效的時間特徵，最後以多尺度熵(MSE)值來分析滾珠導螺桿預壓失效時訊號複雜度的趨勢。



## Concluding Remarks

- ❖ To summarize, one can drive the ball screw, perform the current data acquisition, and calculate the comparative evaluations of the proposed method before manufacturing.
- ❖ Thus, the machinery status of preload loss or not will be identified and known in advance before operation.



## Concluding Remarks

- ❖ Experimental results are extremely promising for prognostic monitoring on preload loss in industry.
- ❖ Use of this technique for diagnosing a ball screw preload loss through the motor torque current signals is solid and successful.
- ❖ Such results comply with the sensorless framework for future manufacturing processes and a module for smart-sensing the health of ball screws is available.

Thank You !