

# 時系列解析（5）

– 追加配布用（多変量ARモデルの応用） –

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# 多変量自己回帰モデル (MAR, VAR Model)

$$y_n = (y_n(1), \dots, y_n(\ell))^T \quad \ell \text{ 変量時系列}$$

$$y_n = \sum_{j=1}^m A_j y_{n-j} + v_n \quad A_j = \begin{bmatrix} a_j(1,1) & \cdots & a_j(1,\ell) \\ \vdots & \ddots & \vdots \\ a_j(\ell,1) & \cdots & a_j(\ell,\ell) \end{bmatrix}$$

$v_n$  :  $\ell$  変量正規白色雑音

$$E[v_n] = \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix}, \quad E[v_n v_n^T] = \begin{bmatrix} \sigma_{11} & \cdots & \sigma_{1\ell} \\ \vdots & \ddots & \vdots \\ \sigma_{\ell 1} & \cdots & \sigma_{\ell\ell} \end{bmatrix} = W \quad \sigma_{ij} = \sigma_{ji}$$

$$E[v_n v_m^T] = O \quad n \neq m$$

$$E[v_n y_m^T] = O \quad n > m$$

$v_n$  は自分自身と独立 (白色)

$v_n$  は  $y_n$  の過去と独立

# 相互共分散関数

$$C_k(i, j) = E[y_n(i)y_{n-k}(j)]$$

$$C_k = E[y_n y_{n-k}^T]$$

$$C_k = \begin{bmatrix} C_k(1,1) & \cdots & C_k(1,\ell) \\ \vdots & \ddots & \vdots \\ C_k(\ell,1) & \cdots & C_k(\ell,\ell) \end{bmatrix}$$

多変量ARモデル

$$y_n = \sum_{j=1}^m A_j y_{n-j} + v_n$$

$$A_k = \begin{bmatrix} a_k(1,1) & \cdots & a_k(1,\ell) \\ \vdots & \ddots & \vdots \\ a_k(\ell,1) & \cdots & a_k(\ell,\ell) \end{bmatrix}$$

Yule-Walker方程式

$$C_0 = \sum_{j=1}^m A_j C_{-j} + W$$

$$C_k = \sum_{j=1}^m A_j C_{k-j} \quad (k = 1, 2, \dots)$$

# クロススペクトル

$$\begin{aligned}
 p_{sj}(f) &= \sum_{k=-\infty}^{\infty} C_k(s, j) e^{-2\pi i k f} \\
 &= \sum_{k=-\infty}^{\infty} C_k(s, j) \cos 2\pi i k f - i \sum_{k=-\infty}^{\infty} C_k(s, j) \sin 2\pi i k f
 \end{aligned}$$

$$P(f) = \begin{bmatrix} p_{11}(f) & \dots & p_{1m}(f) \\ \vdots & \ddots & \vdots \\ p_{m1}(f) & \dots & p_{mm}(f) \end{bmatrix}$$

$$\begin{aligned}
 P(f) &= \sum_{k=-\infty}^{\infty} C_k e^{-2\pi i k f} \\
 C_k &= \int_{-\frac{1}{2}}^{\frac{1}{2}} P(f) e^{2\pi i k f} df
 \end{aligned}$$

$y_n$  が多変量ARモデルに従うとき

$$y_n = \sum_{j=1}^m A_j y_{n-j} + v_n, \quad v_n \sim N(0, W)$$

$$P(f) = A(f)^{-1} W (A(f)^{-1})^*$$

$$A(f) = \begin{bmatrix} A_{11}(f) & \dots & A_{1m}(f) \\ \vdots & \ddots & \vdots \\ A_{m1}(f) & \dots & A_{mm}(f) \end{bmatrix}$$

$$A_{jk}(f) = \sum_{j=0}^m a_j(j, k) e^{-2\pi i j f}, \quad a_0 = -I$$

# クロススペクトル

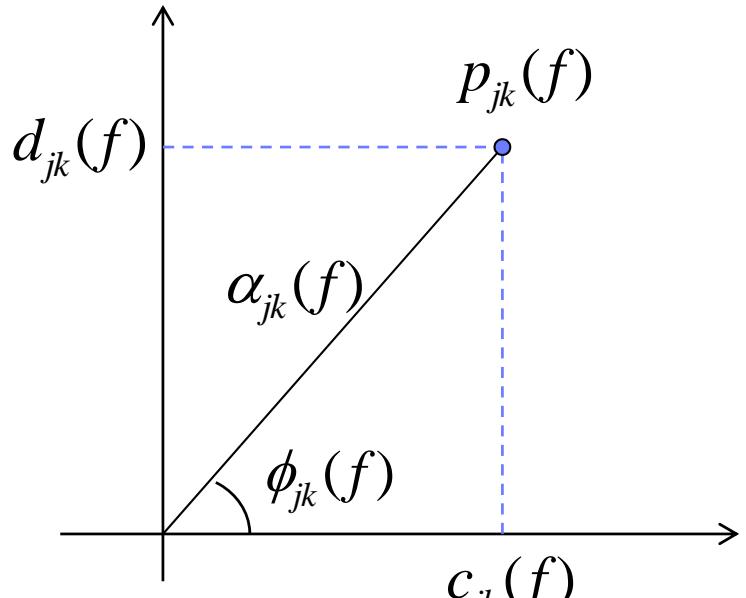
$$p_{jk}(f) = c_{jk}(f) + id_{jk}(f)$$

$$\alpha_{jk}(f) = \sqrt{c_{jk}(f)^2 + d_{jk}(f)^2}$$

$$\phi_{jk}(f) = \arctan d_{jk}(f) / c_{jk}(f)$$

$$p_{jk}(f) = \alpha_{jk}(f) e^{i\phi_{jk}(f)}$$

$$coh_{jk}(f) = \frac{\alpha_{jk}(f)^2}{p_{jj}(f)p_{kk}(f)}$$



```

data(HAKUSAN) # Yaw rate, rolling, pitching and rudder angle for the ship on the open sea
length <- dim(HAKUSAN)[1]
y <- matrix(, length/2, 3)

```

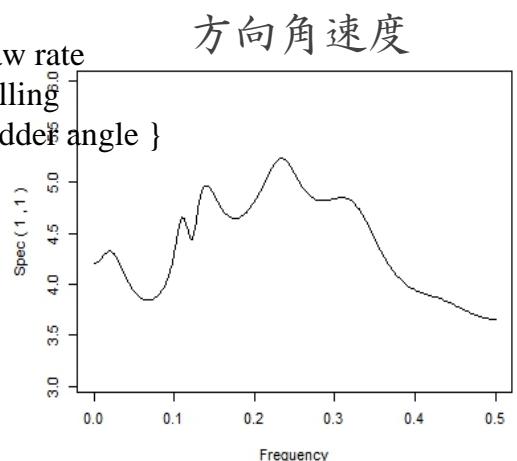
# クロススペクトル

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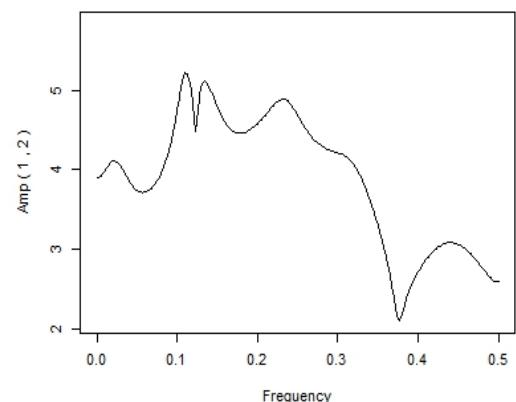
for(i in 1:length/2) {
  y[i,1] <- HAKUSAN[i*2,1] # yaw rate
  y[i,2] <- HAKUSAN[i*2,2] # rolling
  y[i,3] <- HAKUSAN[i*2,4] # rudder angle }
z <- marfit(y,20)

```

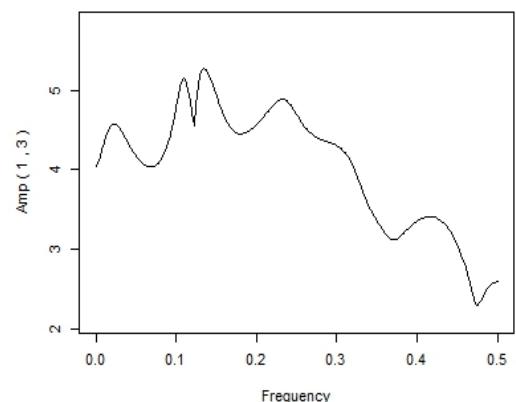
方向角速度



縦揺れ

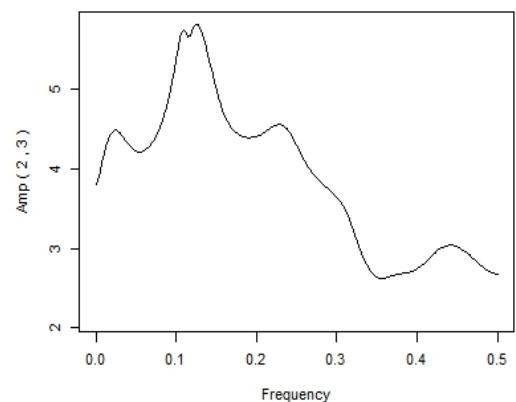
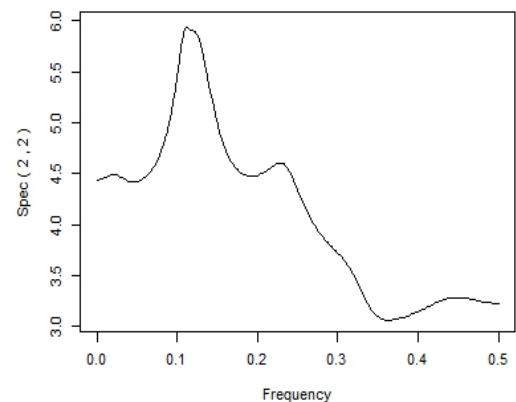
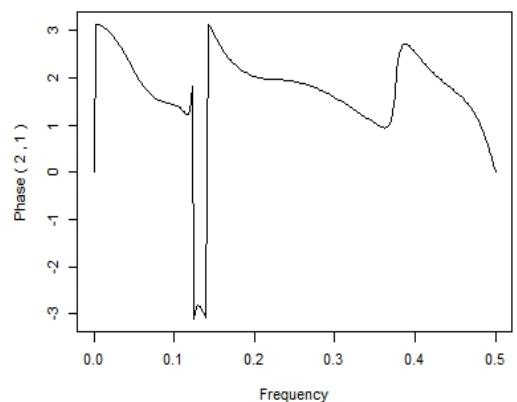


舵角

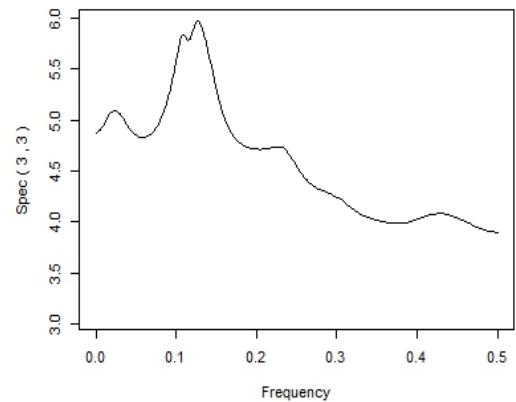
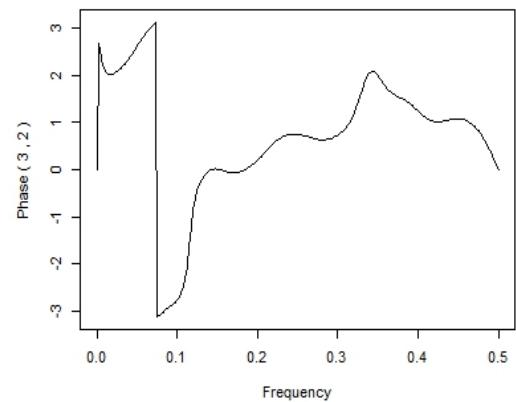
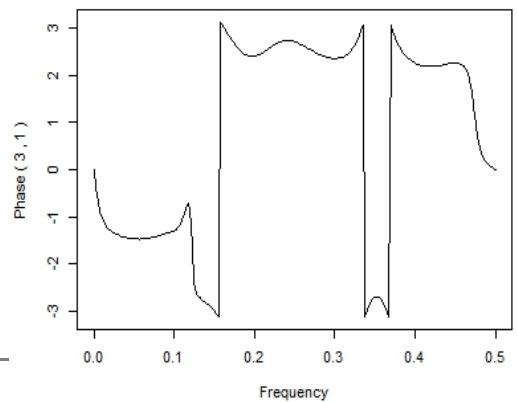


対角線：スペクトル  
上：振幅スペクトル  
下：位相スペクトル

縦揺れ

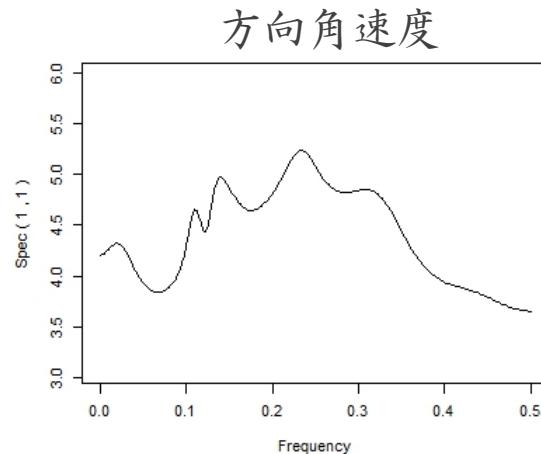


舵角

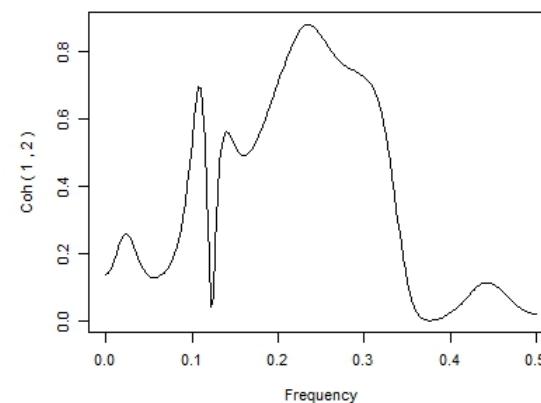


# パワースペクトルとコヒーレンシー

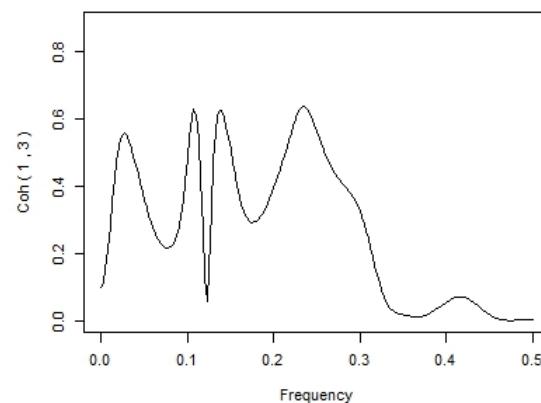
方向角速度



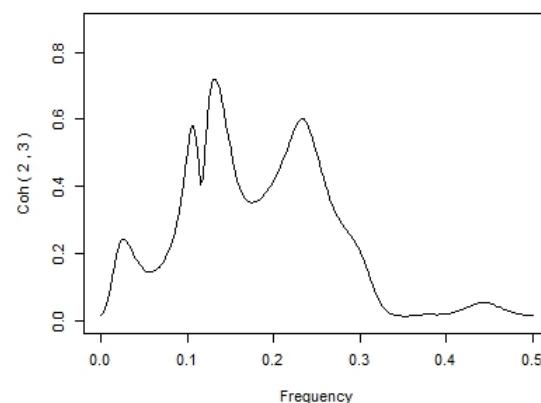
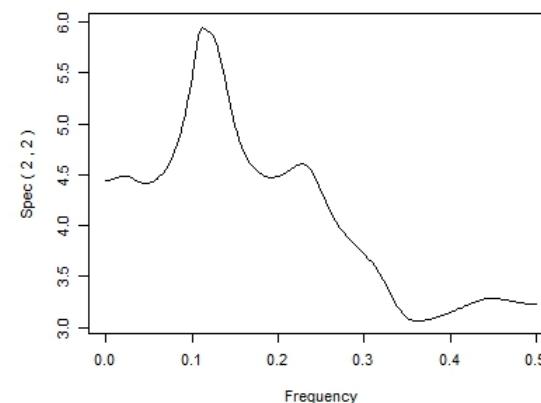
縦揺れ



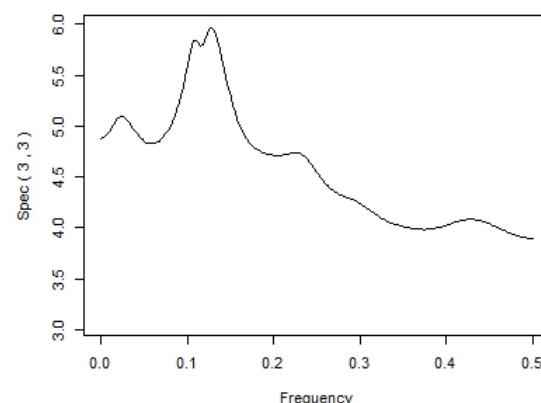
舵角



縦揺れ



舵角



対角線：パワースペクトル  
上： コヒーレンシー

# パワー寄与率

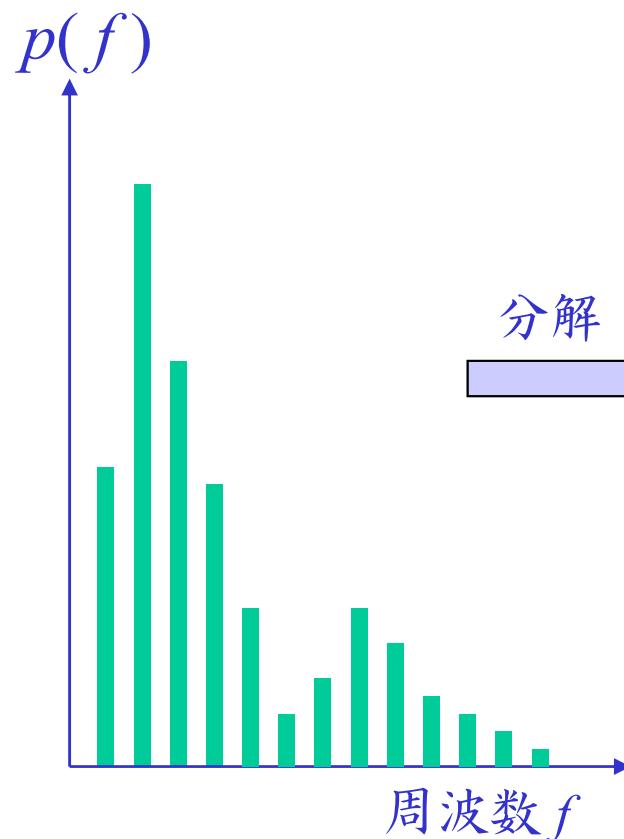
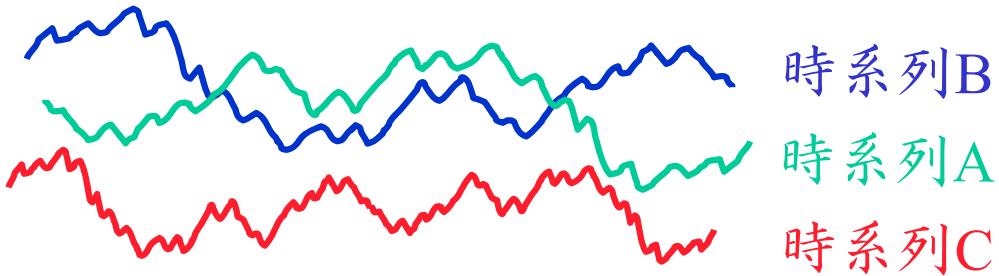
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$$p_{ii}(f) = \sum_{j=1}^{\ell} b_{ij}(f) \sigma_j^2 b_{ij}(f)^* \equiv \sum_{j=1}^{\ell} |b_{ij}(f)|^2 \sigma_j^2$$

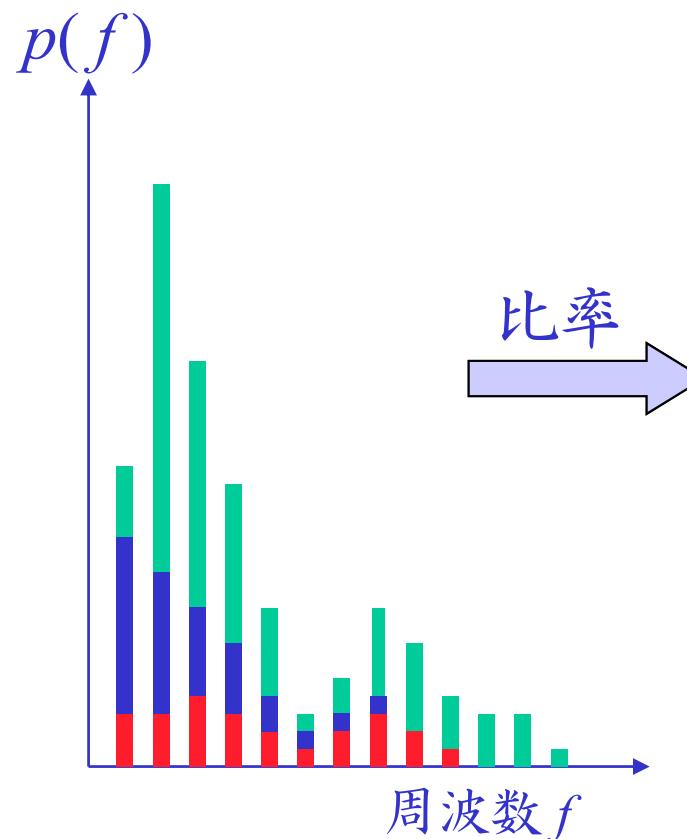
$$r_{ij}(f) = \frac{|b_{ij}(f)|^2 \sigma_j^2}{p_{ii}(f)}$$

$$s_{ij}(f) = \sum_{k=1}^j r_{ik}(f) = \frac{\sum_{k=1}^j |b_{ik}(f)|^2 \sigma_k^2}{p_{ii}(f)}$$

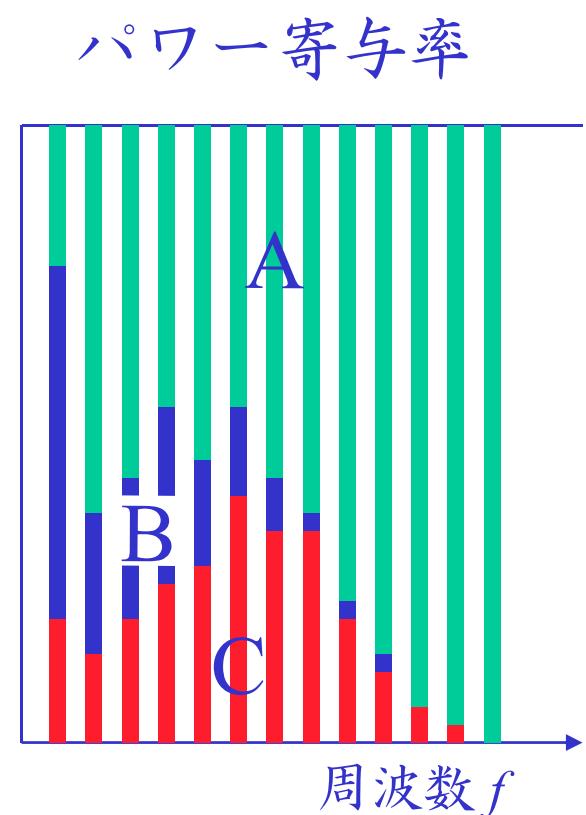
# パワー寄与率



分解

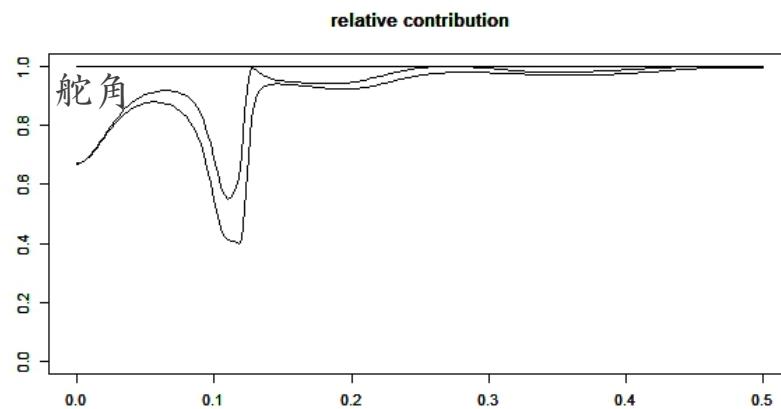
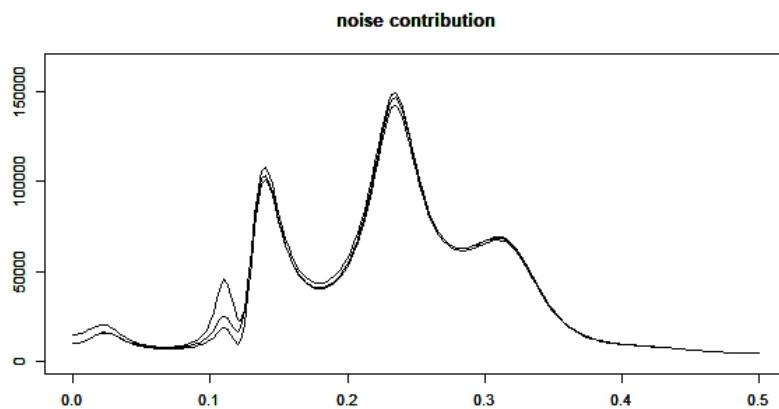


比率

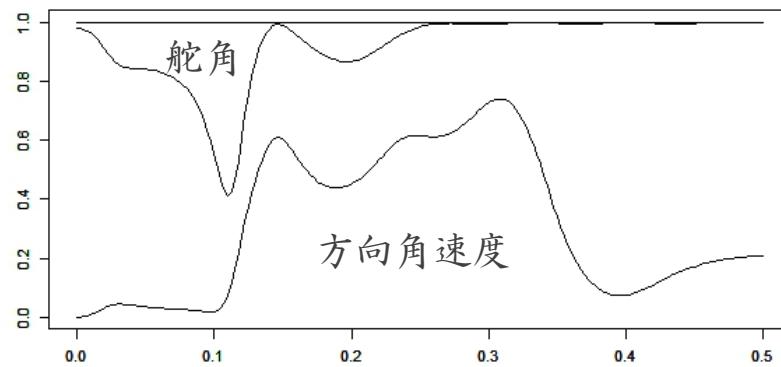
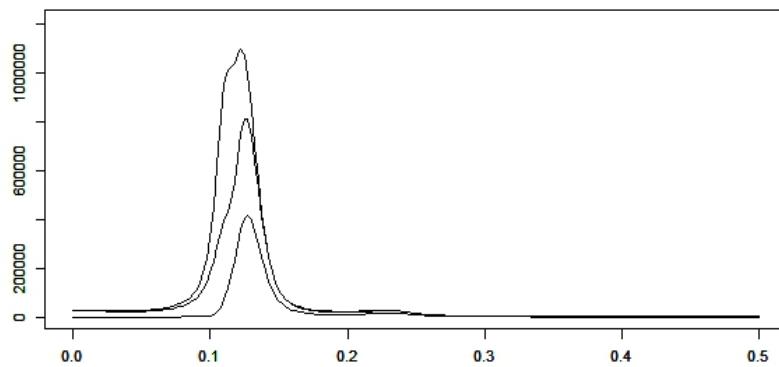


# パワー寄与率

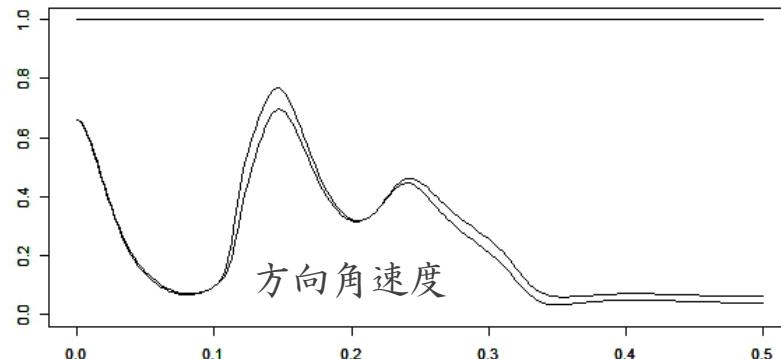
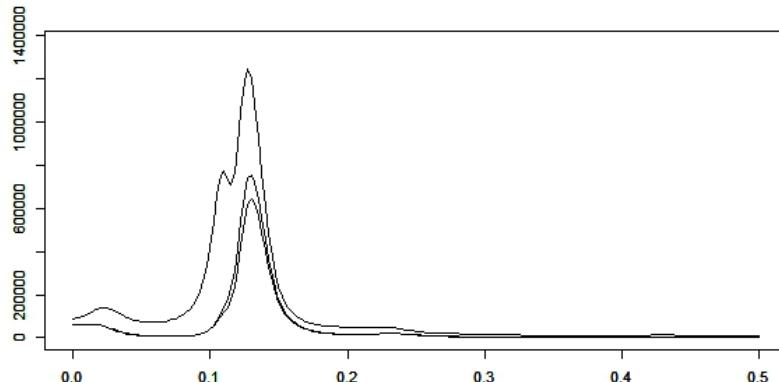
方向角速度



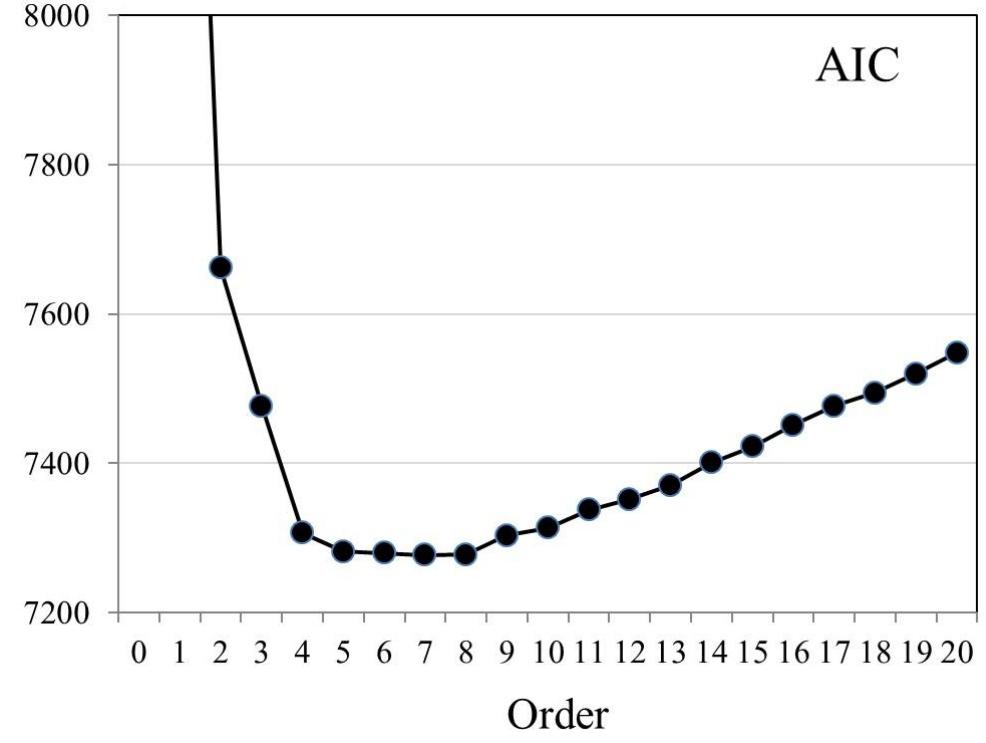
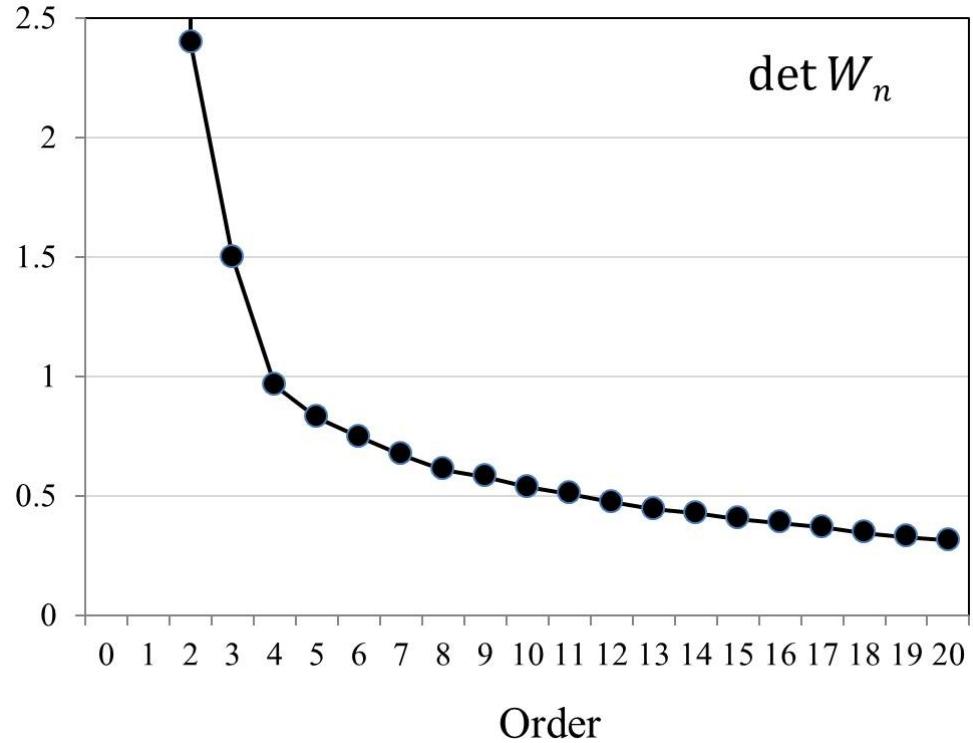
縦揺れ



舵角

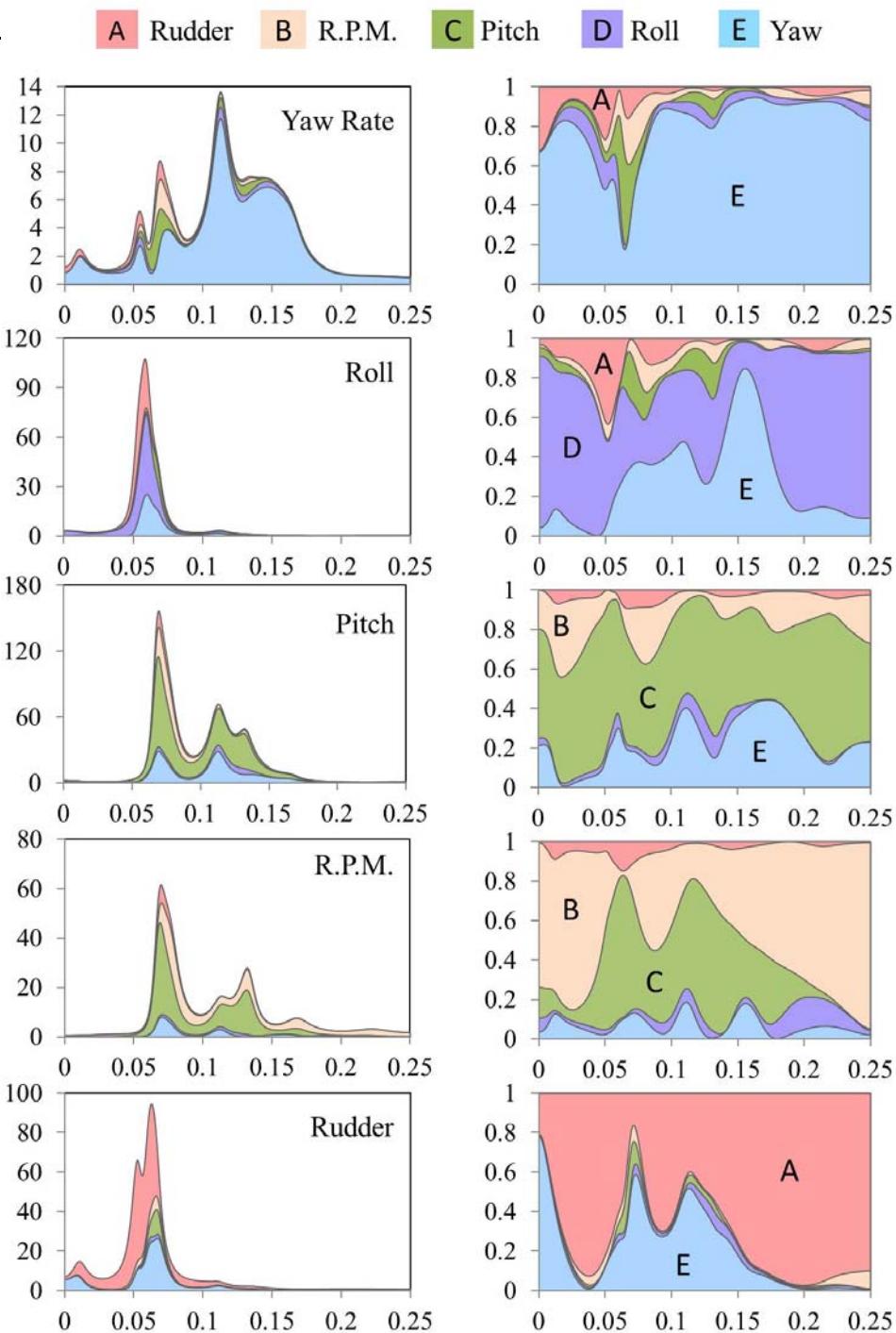


# 船体運動データ

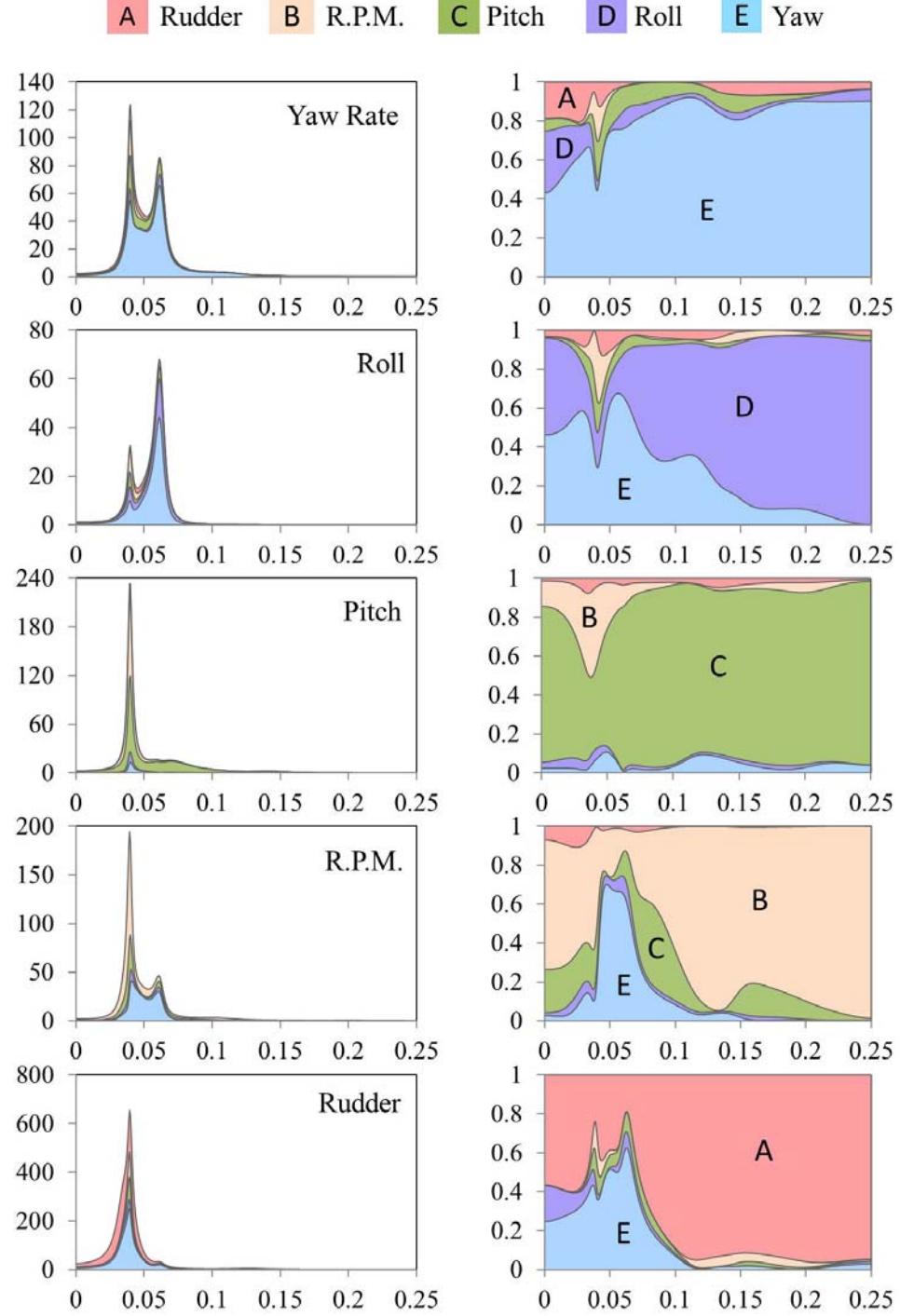


5変量 (yaw rate, roll, pitch,rpm,rudder)

# Autopilot Control



# Manual Control



# パワー寄与率

