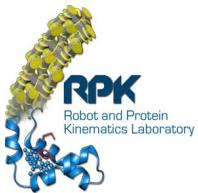


Probabilistic Approaches to the $AXB = YCZ$ Calibration Problem in Multi-Robot Systems

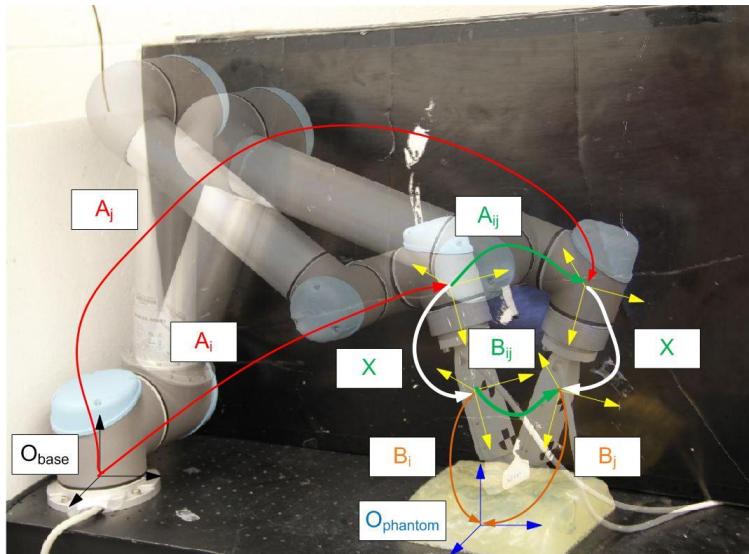
Qianli Ma, Zachariah Goh, Gregory S. Chirikjian

Department of Mechanical Engineering,
Johns Hopkins University, Baltimore, MD, USA





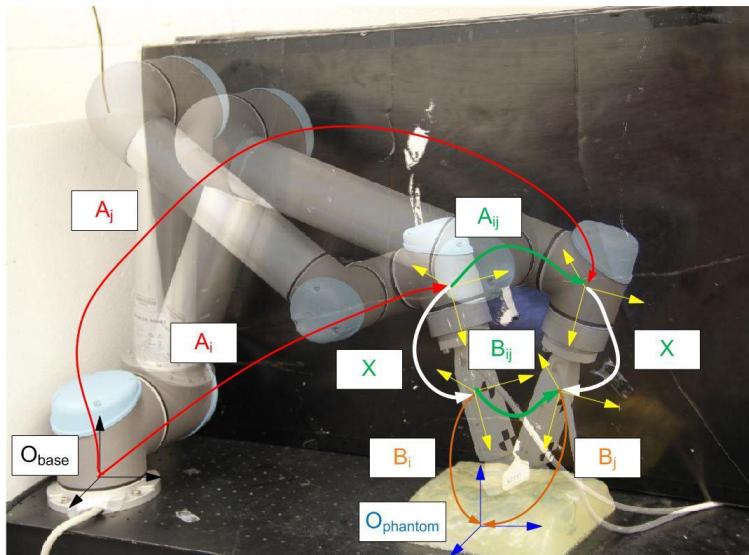
The $AX = XB$ and $AX = YB$ Calibrations



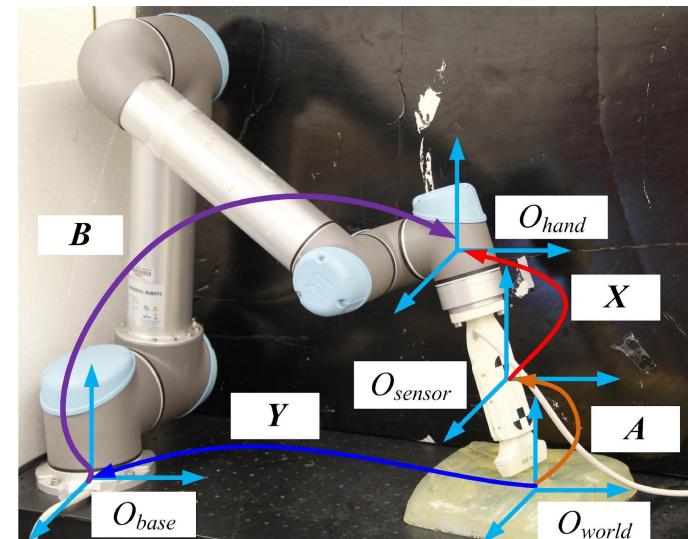
$AX = XB$ Robot Hand-Eye Calibration



The $AX = XB$ and $AX = YB$ Calibrations



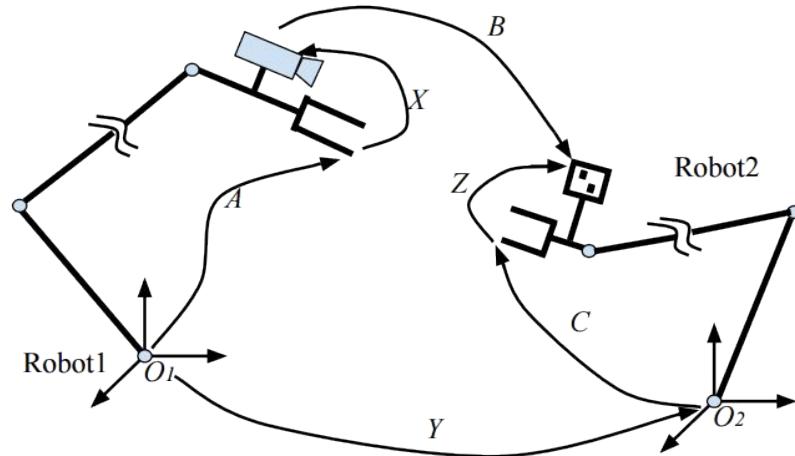
$AX = XB$ Robot Hand-Eye Calibration



$AX = YB$ Hand-Eye and Robot-World Calibration



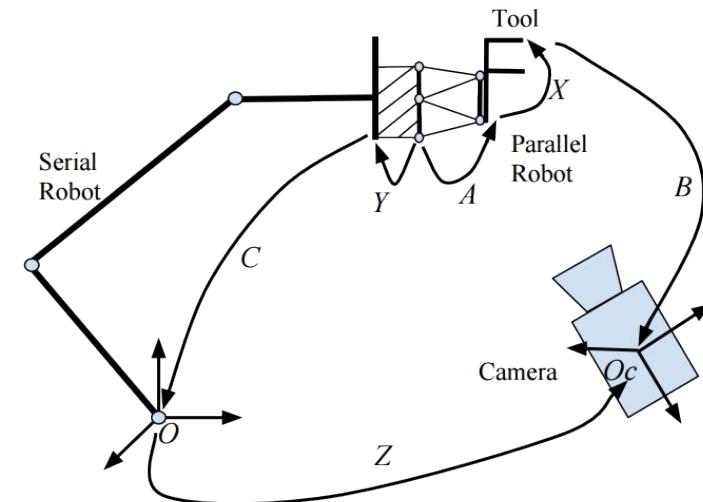
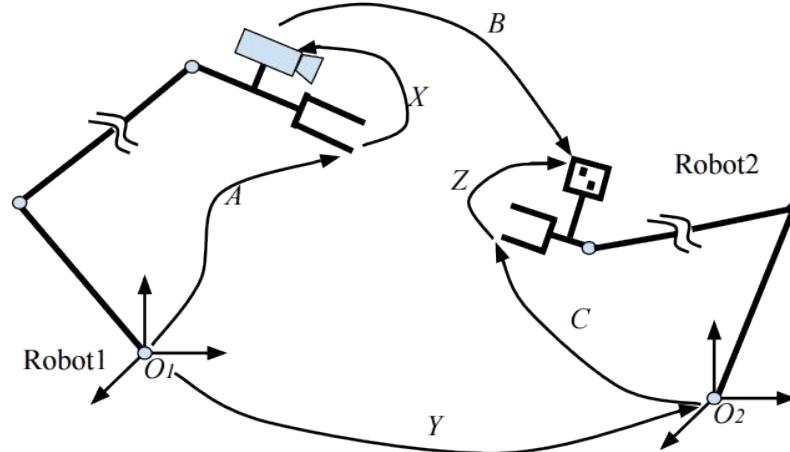
The $\mathbf{AXB} = \mathbf{YCZ}$ Calibration



Hand-Eye (X), Robot-Robot (Y) and Tool-Flange (Z)
Calibrations of a Dual Arm System^[1,2]



The $AXB = YCZ$ Calibration



Hand-Eye (X), Robot-Robot (Y) and Tool-Flange (Z)
Calibrations of a Dual Arm System^[1,2]



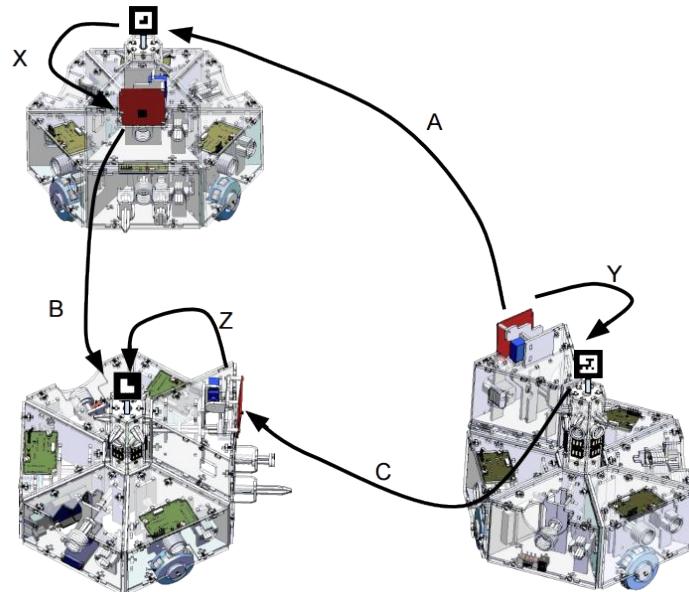
[1] Wang, Jiaole, et al. IROS 2014. [2] Wu, Liao, et al. T-RO 2016.

Flange-Base (Y), Camera-Base (Z) and Tool-
Gripper (X) Calibrations of a Serial-Parallel
Manipulator^[3]

[3] Yan, S.J., et al. Robotica 2015.



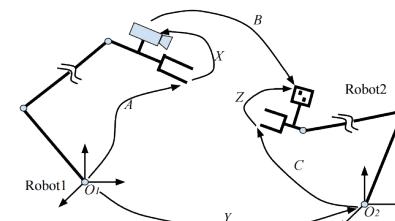
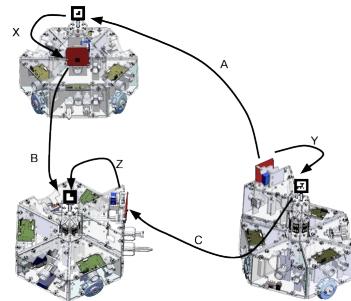
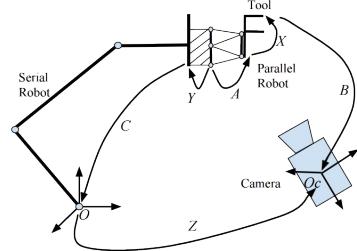
The $AXB = YCZ$ Calibration



Triple Hand-Eye (X, Y, Z) Calibrations of a Multi-Robot System

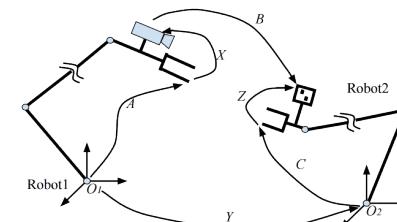
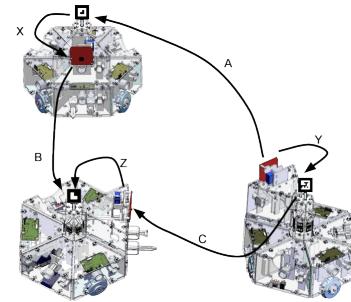
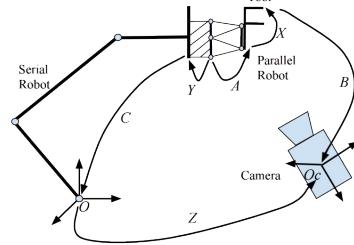


Mathematical Background



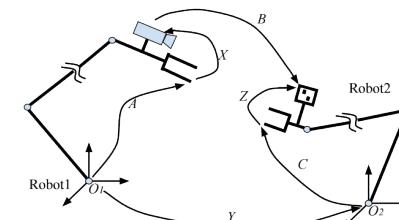
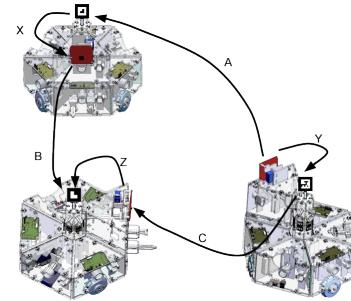
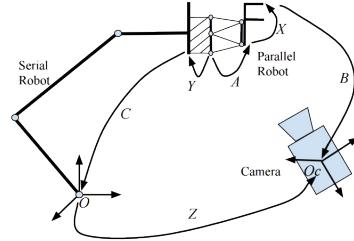


Mathematical Background


$$A = \{A_i\} \rightarrow f_A(H) \text{ with mean } M \text{ and covariance } \Sigma$$
$$X \rightarrow \delta_X(H)$$



Mathematical Background



$A = \{A_i\} \rightarrow f_A(H)$ with mean M and covariance Σ
 $X \rightarrow \delta_X(H)$

$$\int_{SE(3)} \log(M^{-1}H) f(H) dH = \mathbb{O}$$

Mean Def.

$$\Sigma = \int_{SE(3)} \log^\vee(M^{-1}H) [\log^\vee(M^{-1}H)]^T f(H) dH$$

Covariance Def.



Mathematical Formulation

$$A_i X B_i = Y C_i Z$$



Mathematical Formulation

$$A_i X B_i = Y C_i Z$$



$(f_A * \delta_X * f_B)(H) \approx (\delta_Y * f_C * \delta_Z)(H)$. Independent of correspondence



Mathematical Formulation

$$A_i X B_i = Y C_i Z$$



$(f_A * \delta_X * f_B)(H) \approx (\delta_Y * f_C * \delta_Z)(H)$. Independent of correspondence



$$M_A X M_B \approx Y M_C Z \quad (1) \text{ Mean Equation on } SE(3)$$

$$Ad(B^{-1})Ad(X^{-1})\Sigma_A Ad^T(X^{-1})Ad^T(B^{-1}) + \Sigma_B \approx Ad(Z^{-1})\Sigma_C Ad^T(Z^{-1})$$

(2) Covariance Equation on $\mathbb{R}^{6 \times 6}$



Mathematical Formulation

$$A_i X B_i = Y C_i Z$$



$(f_A * \delta_X * f_B)(H) \approx (\delta_Y * f_C * \delta_Z)(H)$. Independent of correspondence



$$M_A X M_B \approx Y M_C Z \quad (1) \text{ Mean Equation on } SE(3)$$

$$Ad(B^{-1})Ad(X^{-1})\Sigma_A Ad^T(X^{-1})Ad^T(B^{-1}) + \Sigma_B \approx Ad(Z^{-1})\Sigma_C Ad^T(Z^{-1})$$



(2) Covariance Equation on $\mathbb{R}^{6 \times 6}$

Lost Y in the covariance equation



Case Study with Completely Scrambled Data

$$AXB = YCZ \quad (1)$$

$$C^{-1}Y^{-1}A = ZB^{-1}X^{-1} \quad (2)$$



Case Study with Completely Scrambled Data

$$AXB = YCZ \quad (1)$$



Covariance Eq. (1)

Mean Eq. (1)

$$C^{-1}Y^{-1}A = ZB^{-1}X^{-1} \quad (2)$$



Mean Eq. (2)

Covariance Eq. (2)



Case Study with Completely Scrambled Data

$$AXB = YCZ \quad (1)$$



Covariance Eq. (1)

Mean Eq. (1)



$\{Z_k\}$

$$C^{-1}Y^{-1}A = ZB^{-1}X^{-1} \quad (2)$$



Mean Eq. (2)

Covariance Eq. (2)



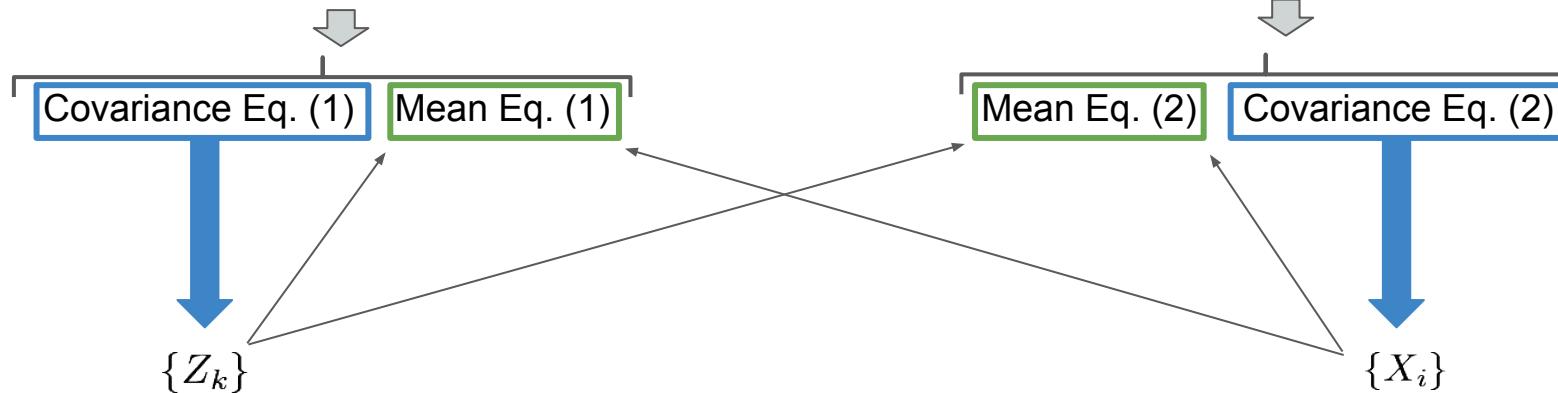
$\{X_i\}$



Case Study with Completely Scrambled Data

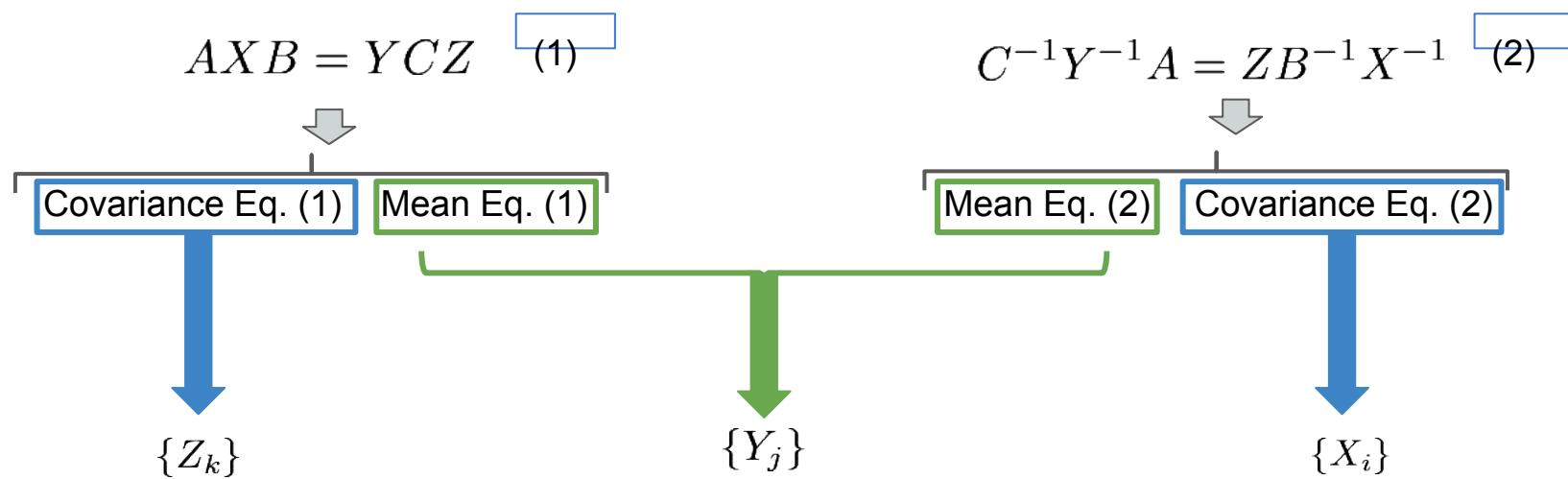
$$AXB = YCZ \quad (1)$$

$$C^{-1}Y^{-1}A = ZB^{-1}X^{-1} \quad (2)$$



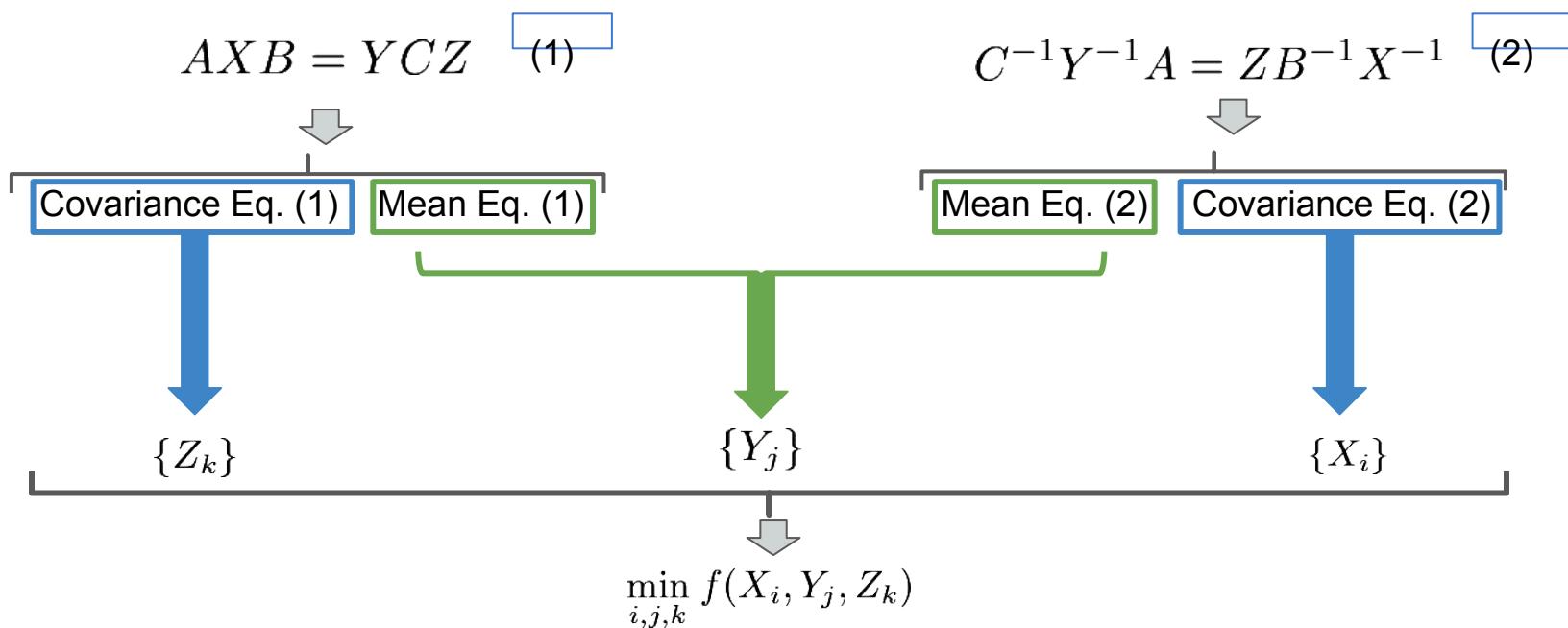


Case Study with Completely Scrambled Data





Case Study with Completely Scrambled Data

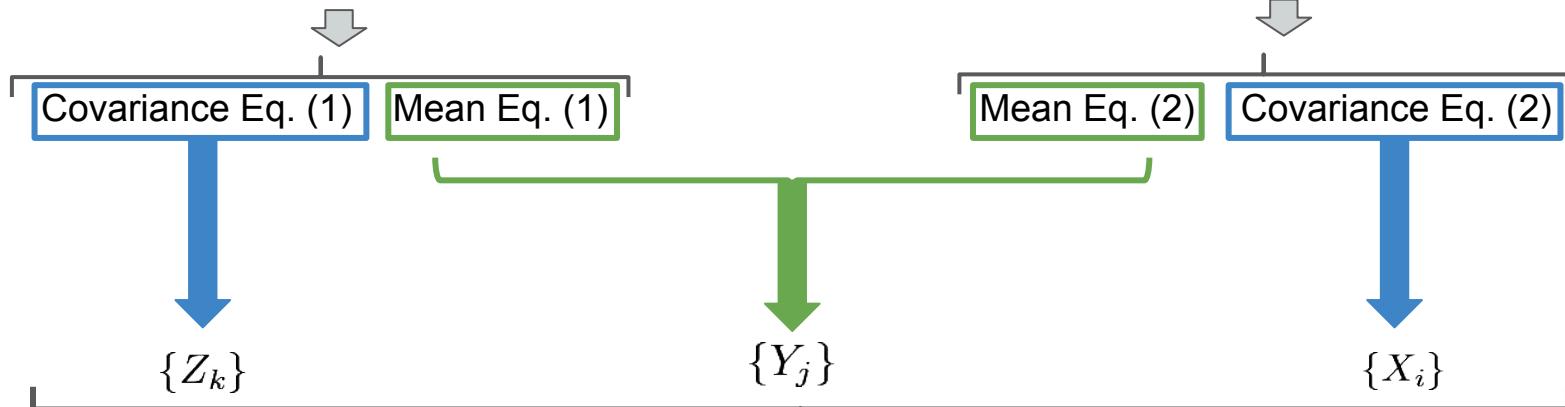




Case Study with Completely Scrambled Data

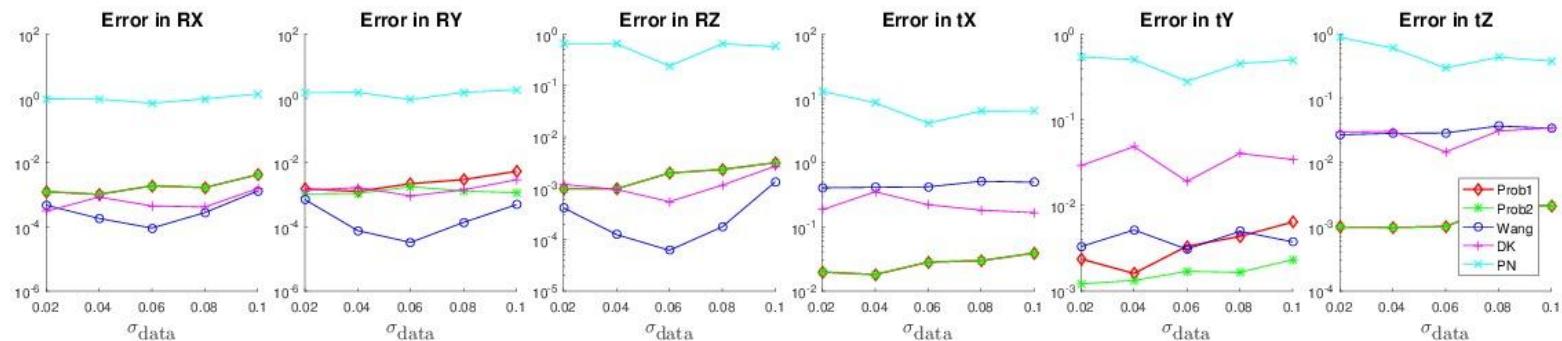
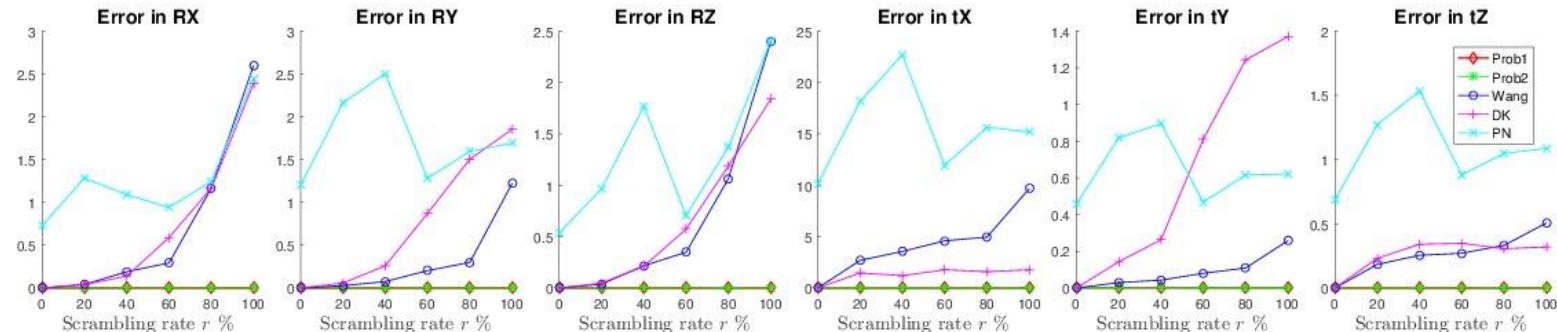
$$AXB = YCZ \quad (1)$$

$$C^{-1}Y^{-1}A = ZB^{-1}X^{-1} \quad (2)$$





Comparison between Probabilistic Methods and Traditionals Methods



Prob1 and *Prob2* are probabilistic methods; *Wang*, *DK* and *PN* are traditional methods



Acknowledgements

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Thank you for listening!