

Metabolic Cost Estimation using Dual Unscented Kalman Filter and Gaussian Mixture Model

MECHANICAL AND INDUSTRIAL ENGINEERING COLLEGEOE ENGINEERING

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Introduction



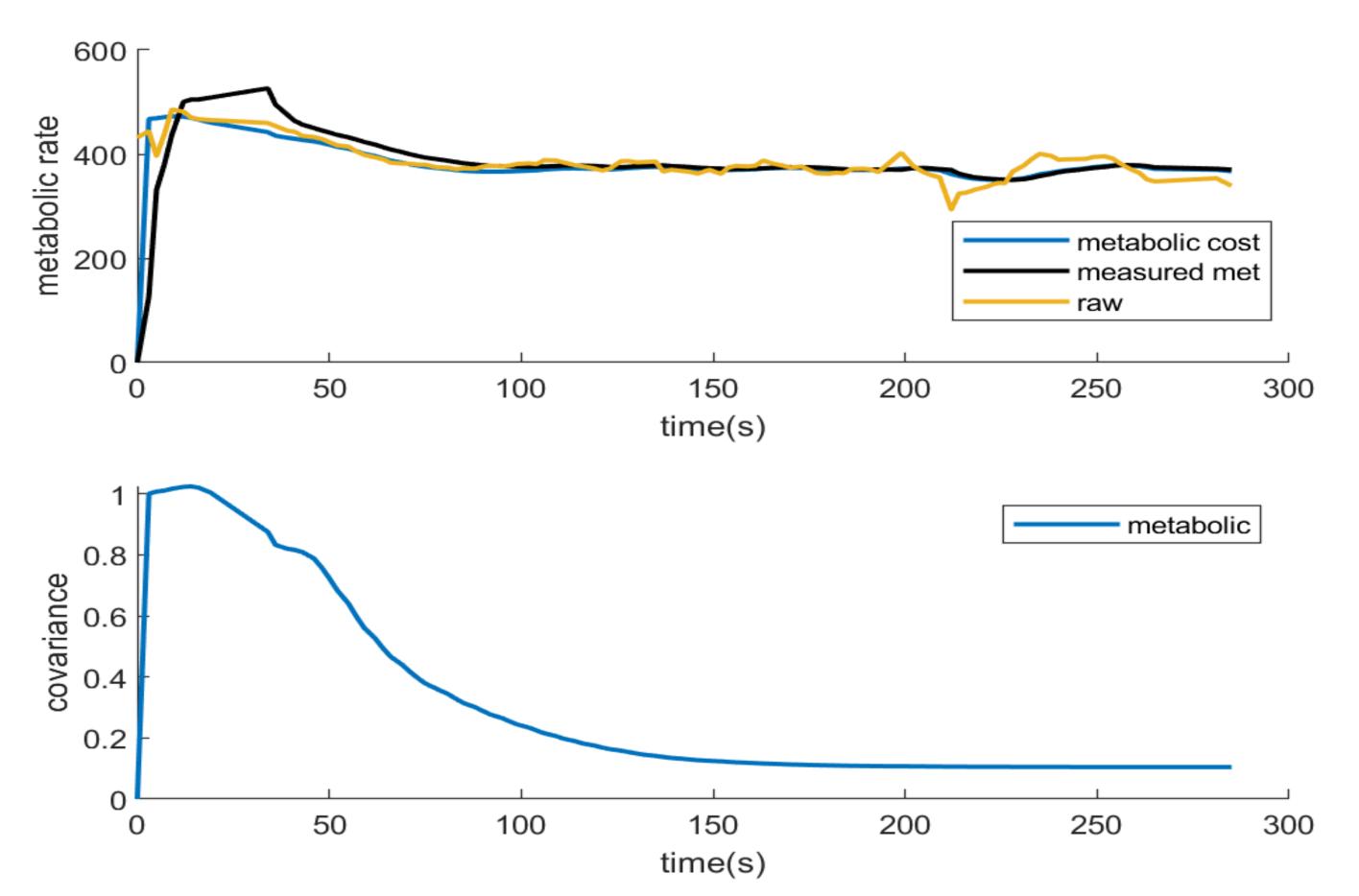
- By assistive increasing robotic devices[1] for human's increasing locomotion, The performance evaluation methods are needed such as metabolic cost [2].
- The metabolic cost is the amount of the energy demand to perform a certain given task [3].

A. Dual Unscented Kalman Filter

Table 1. Maximum error results with different time and data points using Dual UKF

	All	2min	1.5min	30 samples	20 samples
Max Err	1.6%	4.3%	4.7%	6.1%	8.4%

Results



- The challenges of measuring metabolic cost
- ✓ **Time delay** [4] because of the mitochondrial dynamics
- ✓ **Noisy** in respiratory measurement [5]
- \checkmark Long measurement time [5] : at least five minutes to obtain a reasonable estimate per testing condition.
- The purpose of the paper is to minimize the metabolic cost estimation time as well as provide high confidence in the measurement.
- We proposed two algorithms, <u>Dual Unscented Kalman Filter(Dual</u> UKF) and Gaussian Mixture Model(GMM)

Fig. 1 Metabolic estimation of Subject 1 in 4th condition with 5min data based on Dual UKF

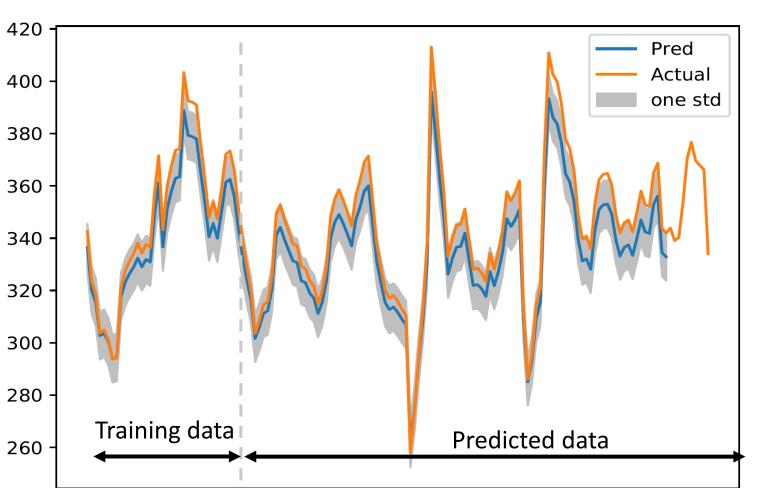
Methodology

- A. Dual Unscented Kalman Filter (Dual UKF)
 - Dual UKF
 - ✓ No need linearization (improves stability)
 - Simultaneously estimates model parameters and state \checkmark
 - Dynamic model

$$\begin{pmatrix} & -\frac{t}{2} \end{pmatrix}$$

B. Gaussian Mixture Model (GMM)

- Table 2. Maximum error results with different data points using GMM
- 50 30 20 samples samples samples 3.0% 7.7% Max Err 2.2%



 $Y = a (1 - e^{-\tau} J + a_0)$

- τ : the rate of change in time constant
- a_0 : metabolic cost in standing position
- *a* : initialized underlying metabolic cost using ten initial data

B. Gaussian Mixture Model (GMM)

- GMM \bullet
 - ✓ Data driven and model-free algorithm
 - ✓ Based on prior modeling and Normal-Inverse-Wishart(NIW) distribution
- Linear-Gaussian dynamics \bullet

 $P(x_{t+1}|x_t, u_t) = N(f_{xt}x_t + f_{ut}u_t + f_{ct}, F_t)$

- \checkmark Use GMM to add the posterior information about the system to estimate
- \checkmark Trained the model using the limited number of respiratory data points.
- \checkmark The model is used to predict complete metabolic estimation for a single mode

C. Evaluation

- Used previously collected metabolic data(N=7), 4 conditions
- Calculated the maximum error between the true metabolic cost(two minutes average of five minutes data) and the estimated metabolic cost using a limited data set for each algorithm.

20 40 100 120 140 0 Number of data points

Fig. 2 Metabolic estimation results with 30 given data points using GMM

Discussion

We could reduce metabolic cost estimation time using both Dual UKF and GMM algorithm.

	Advantages	Disadvantages	
Dual UKF	No need training data	Larger error	
GMM	Smaller error	Need training data	

- Future work UKF : optimizing model uncertainty, noise, covariance matrix
- Both algorithms could be used to optimize assistance method in real-time with further improvement

References

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