

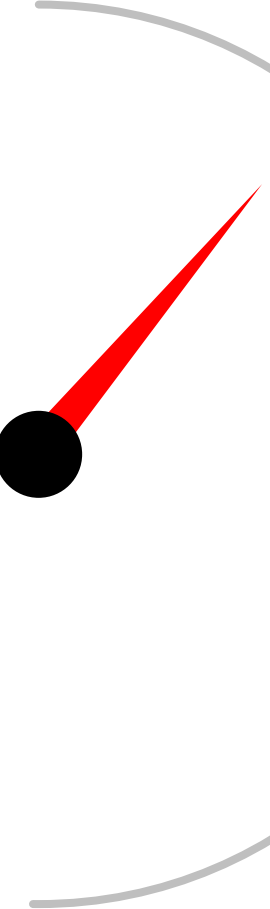
The Introspective Orifice Meter – Uncertainty Improvements

by

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26-29 October 2020

This is what I'm going to talk about

- 
- 1 Recap Maximum Likelihood Uncertainty (MLU)
 - 2 The next dimension - TIME
 - 3 Kalman Filter
 - 4 Thought experiment – theoretical example
 - 5 Real data
 - 6 Conclusions – the introspective orifice meter

Heads and tails sequences



What Consider the two patterns HTH and HTT

Which of the following is true:

A

The average number of tosses until HTH is **larger** than the average number of tosses until HTT

B

The average number of tosses until HTH is the **same** as the average number of tosses until HTT

C

The average number of tosses until HTH is **smaller** than the average number of tosses until HTT



1 Recap Maximum Likelihood Uncertainty (MLU)

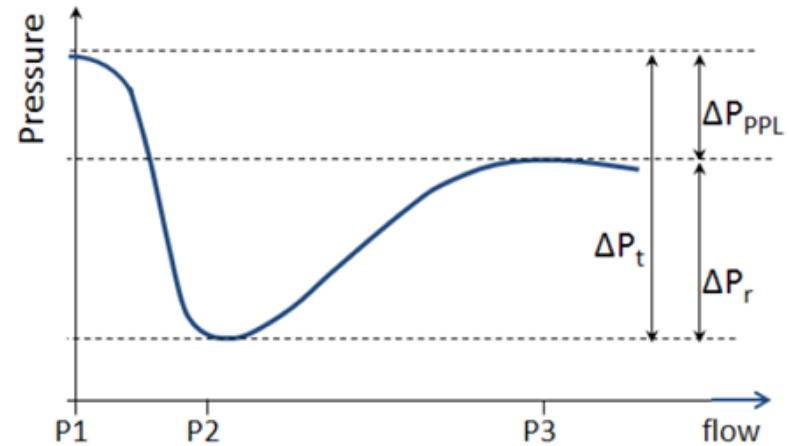
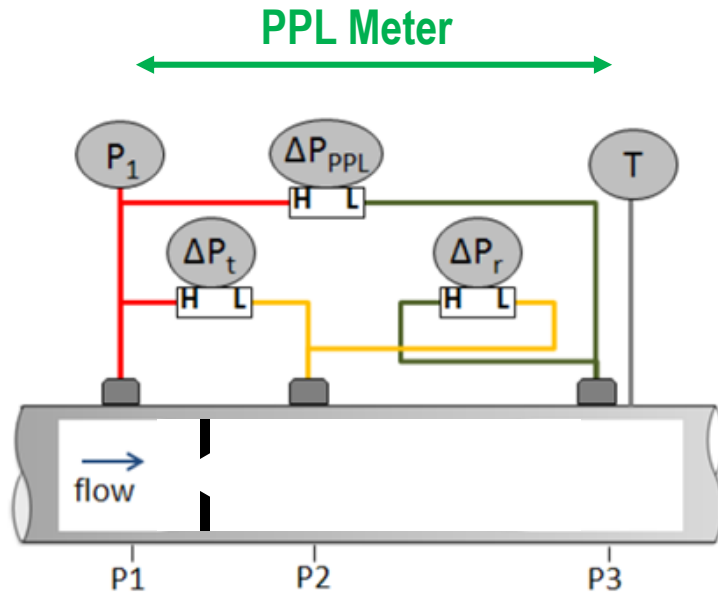
2 The next dimension - TIME

3 Kalman Filter

4 Thought experiment – theoretical example

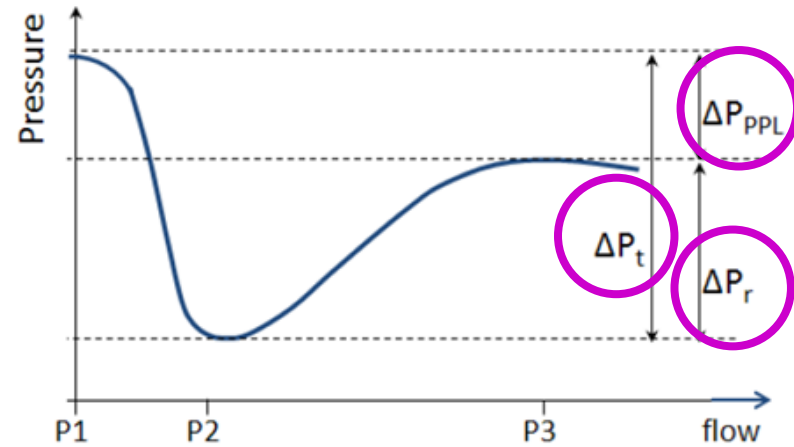
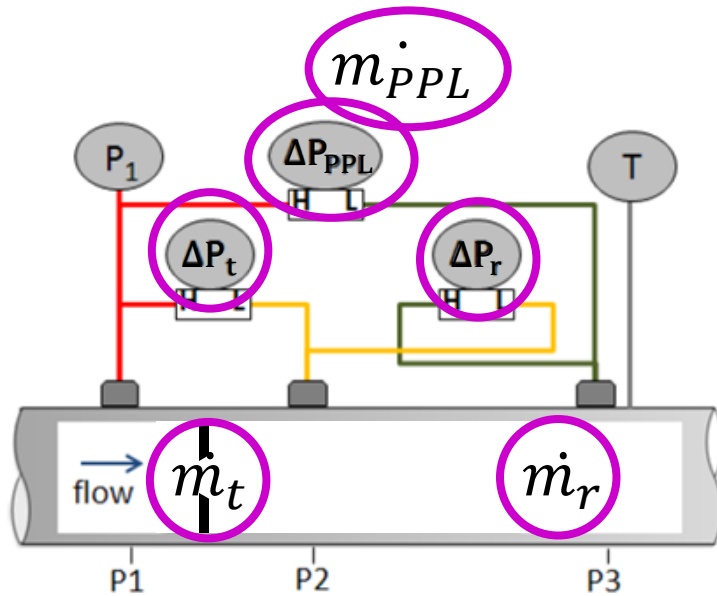
5 Real data

6 Conclusions – the introspective orifice meter

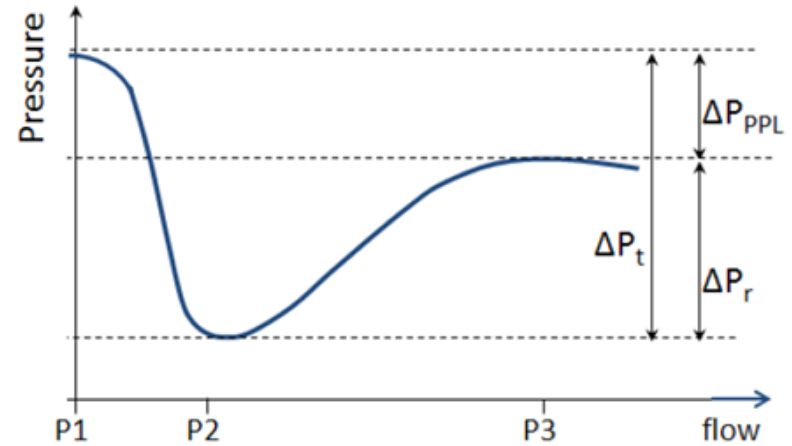
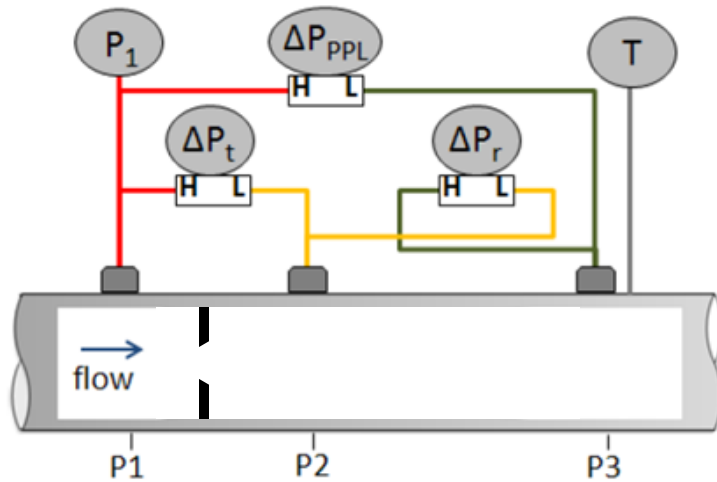


Traditional Expansion

Three DP Meters



Three DP Meters



\dot{m}_t

ΔP_t

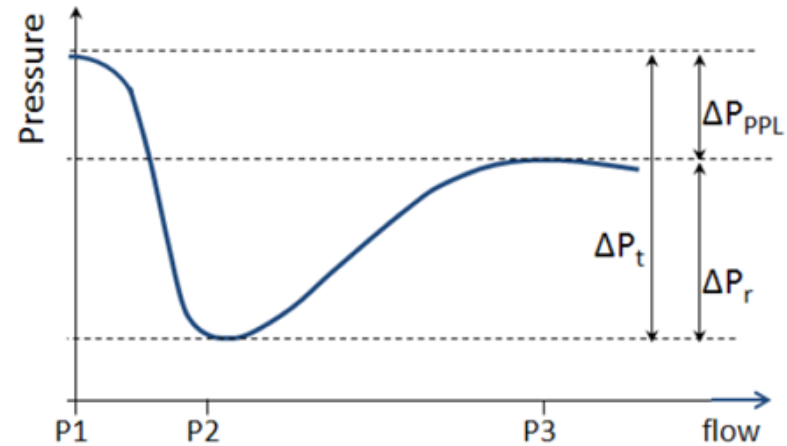
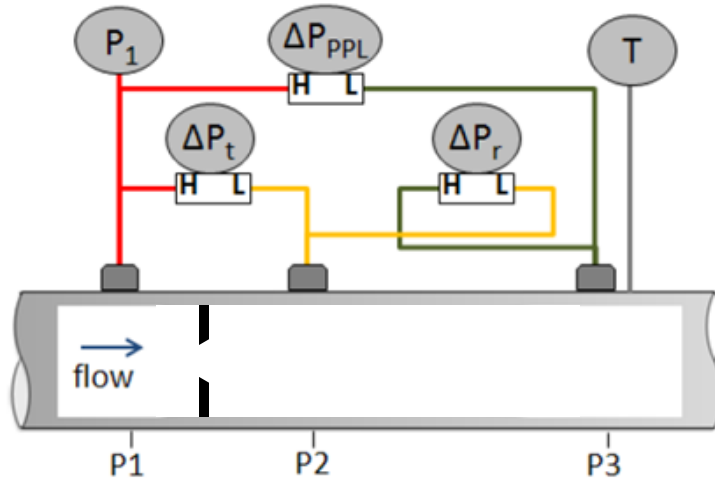
\dot{m}_r

ΔP_r

\dot{m}_{PPL}

ΔP_{PPL}

Three DP Meters

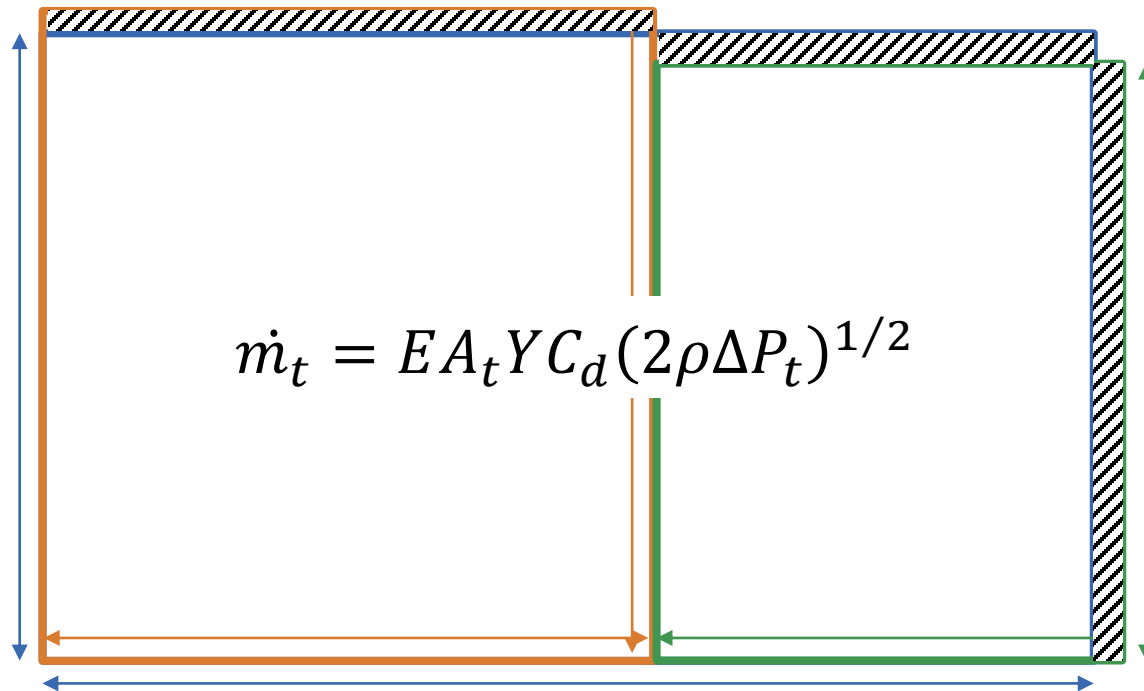


$$\dot{m}_t = EA_t Y C_d (2\rho \Delta P_t)^{1/2}$$

$$\dot{m}_r = EA_t K_r (2\rho \Delta P_r)^{1/2}$$

$$\dot{m}_{PPL} = AK_{PPL} (2\rho \Delta P_{PPL})^{1/2}$$

Mass Flow and Pressure Drop Constraints

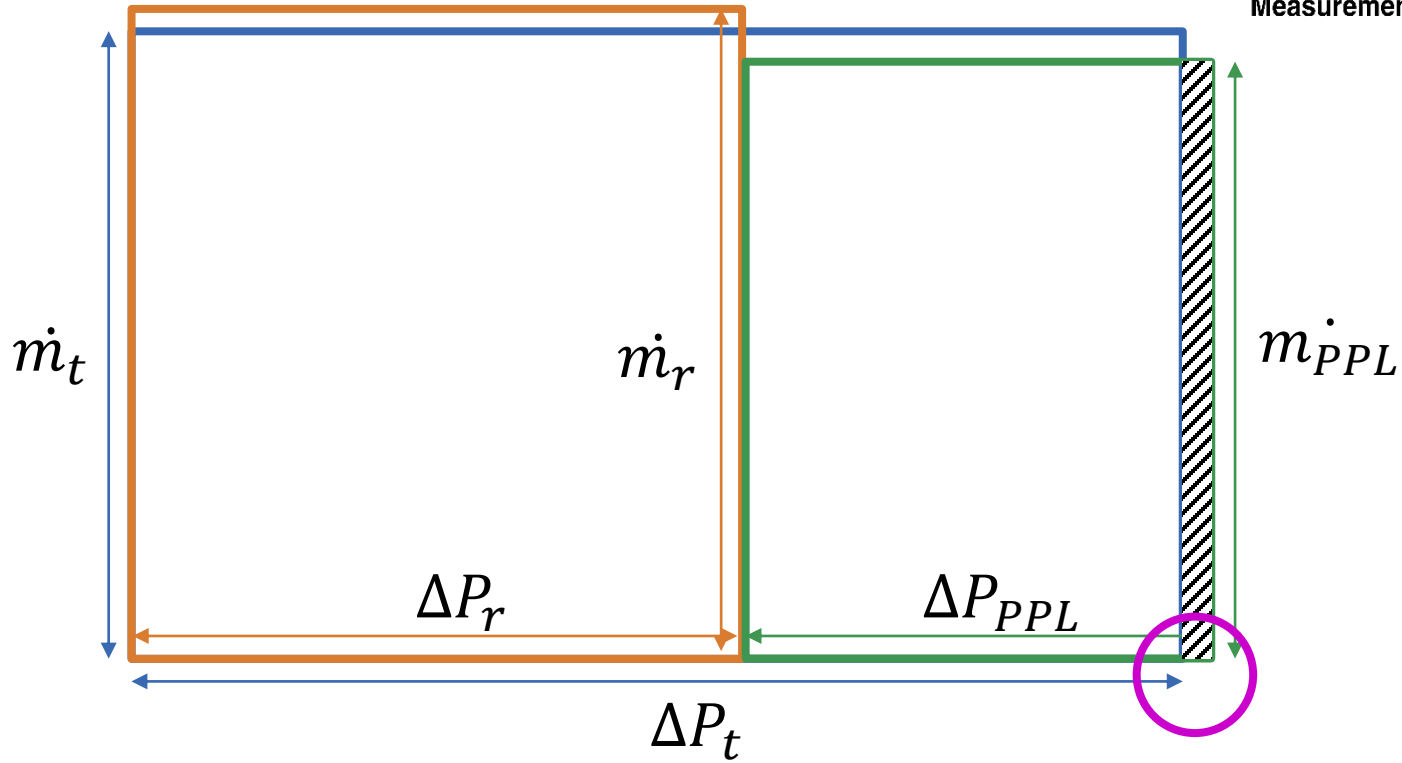


100% certainty – all 3 DP meters are measuring the same flow rate

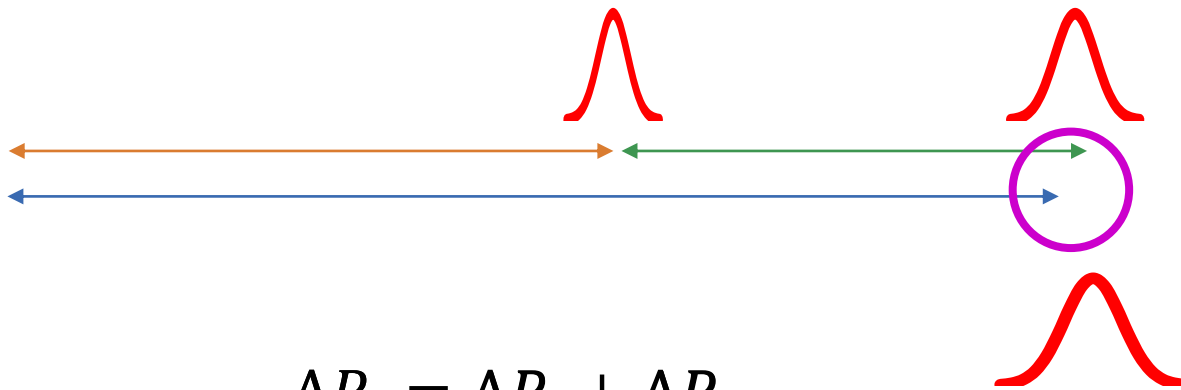


100% certainty – ΔP s must be consistent with each other

Pressure Drop Constraint

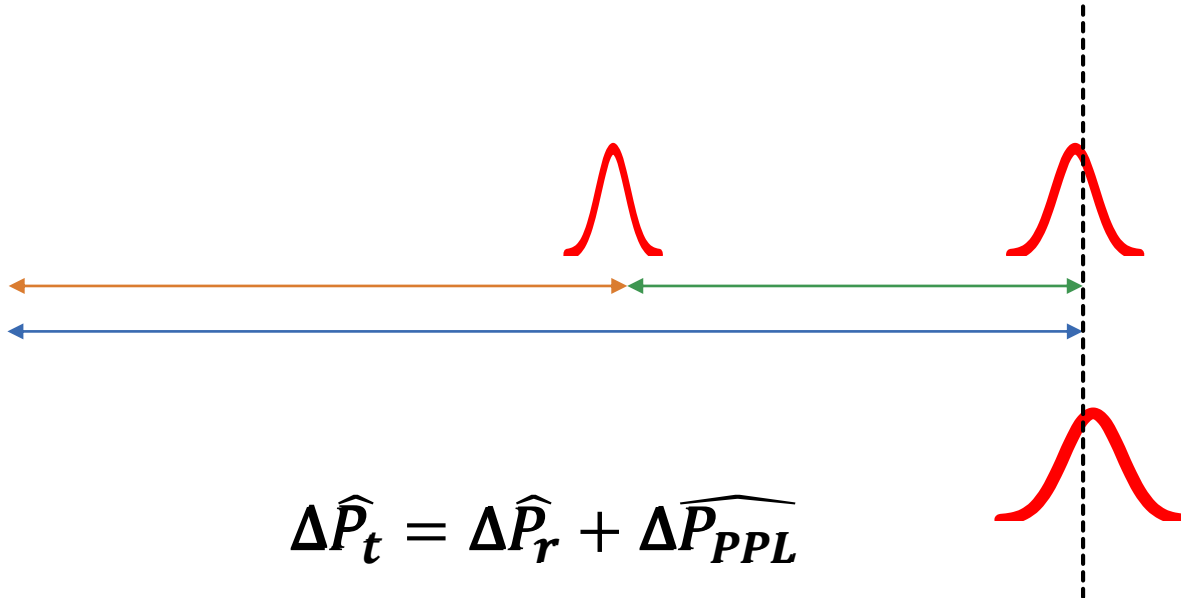


Pressure Drop Constraint

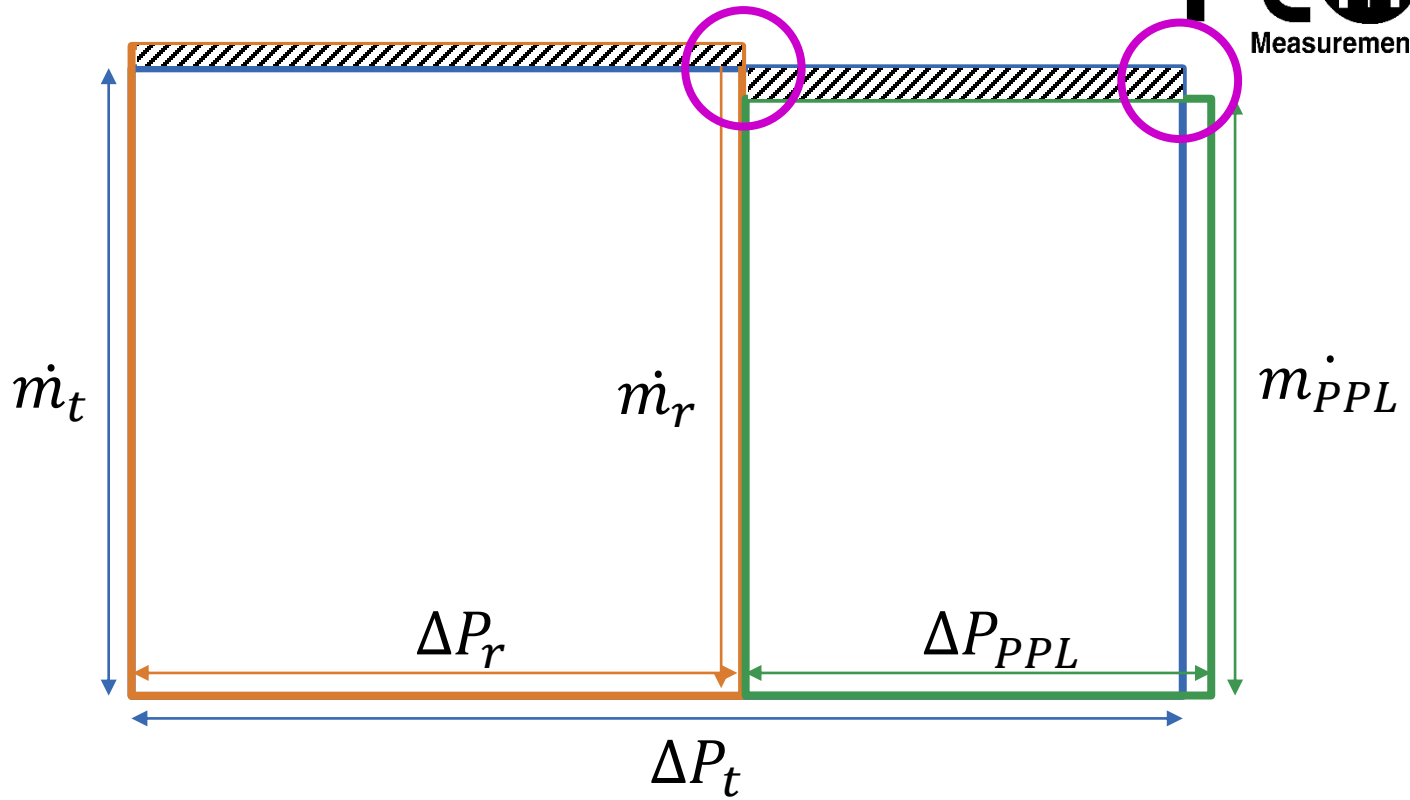


$$\Delta P_t = \Delta P_r + \Delta P_{PPL}$$

Pressure Drop Constraint



Mass Balance



$$\Phi = \hat{m} - \dot{m}_t(\hat{d}, \hat{D}, \hat{Y}, \hat{C}_d, \hat{\rho}, \Delta\hat{P}_t) = 0$$

Mass
Balances

$$\Xi = \hat{m} - \dot{m}_r(\hat{d}, \hat{D}, \hat{K}_r, \hat{\rho}, \Delta\hat{P}_r) = 0$$

$$\Psi = \hat{m} - \dot{m}_{PPL}(\hat{D}, \hat{K}_{PPL}, \hat{\rho}, \Delta\hat{P}_{PPL}) = 0$$

Pressure
Balance

$$\Omega = \Delta\hat{P}_t - \Delta\hat{P}_r - \Delta\hat{P}_{PPL} = 0$$



Measured
Variables

$$x = [\Delta P_t \quad \Delta P_r \quad \Delta P_{PPL} \quad Y \quad C_d \quad K_r \quad K_{PPL} \quad \rho]^t$$

Unmeasured
Variables

$$u = \hat{m}$$

$$S = \sum_i \left(\frac{\hat{x}_i - x_i}{\sigma_{x_i}} \right)^2 \quad \text{Minimise weighted sum}$$



$$u_{i+1} = u_i - (J_u^T (J_x V J_x^T)^{-1} J_u)^{-1} J_u^T (J_x V J_x^T)^{-1} (J_x x_0 + J_x (x_i - x_0))$$


$$x_{i+1} = x_0 - V J_x^T (J_x V J_x^T)^{-1} (J_x x_0 + J_u (u_i - u_0) + J_x (x_i - x_0))$$



Determines MLU mass flow

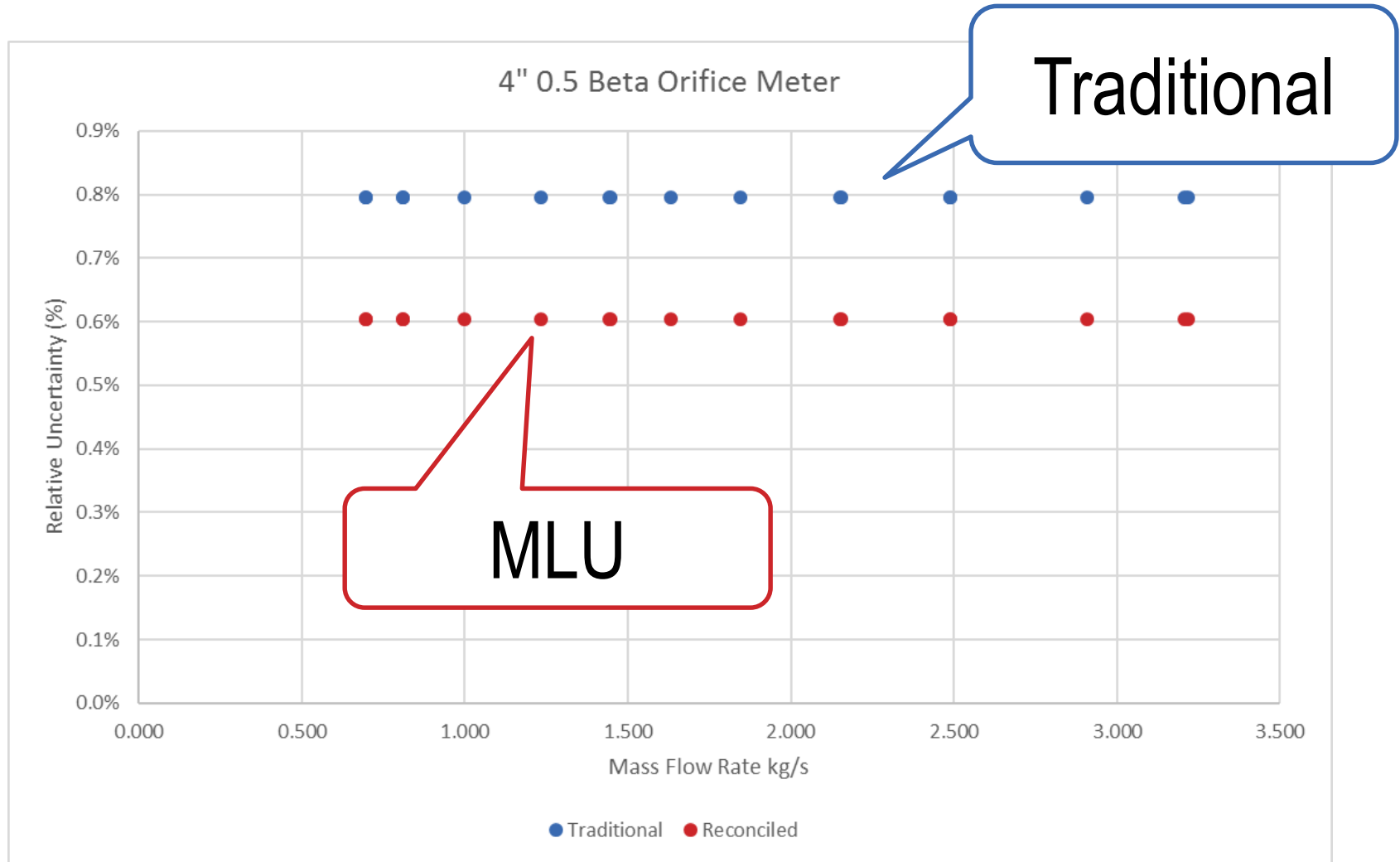
$$u = \hat{m}$$



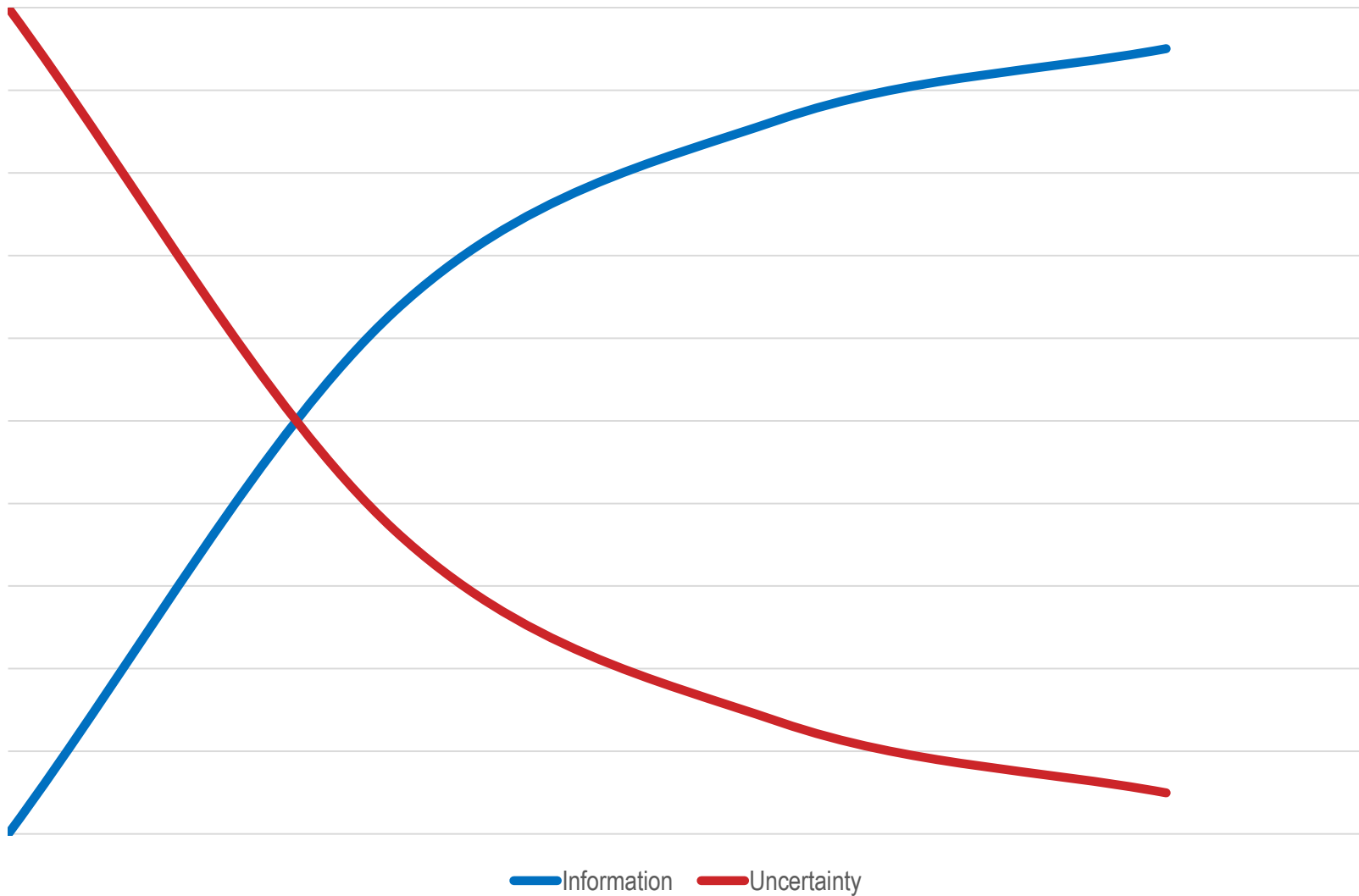
Determines MLU mass flow
uncertainty

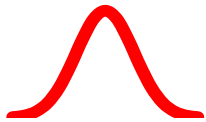
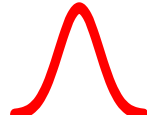
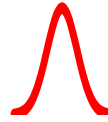
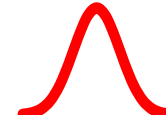
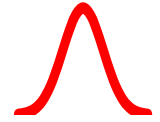
$$\sigma_u = (J_u^T (J_x V J_x^T)^{-1} J_u)^{-1}$$

Uncertainty over a Range of Flows

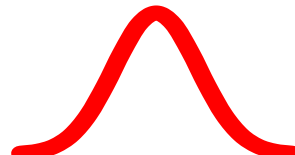


Information and Uncertainty



ΔP_t  ΔP_r  ΔP_{PPL}  E  A_t  A  Y  ρ 

Information
Pressure balance constraint
Mass balance constraint

 $\widehat{\dot{m}_{MLU}}$ 



1 Recap Maximum Likelihood Uncertainty (MLU)

2 **The next dimension - TIME**

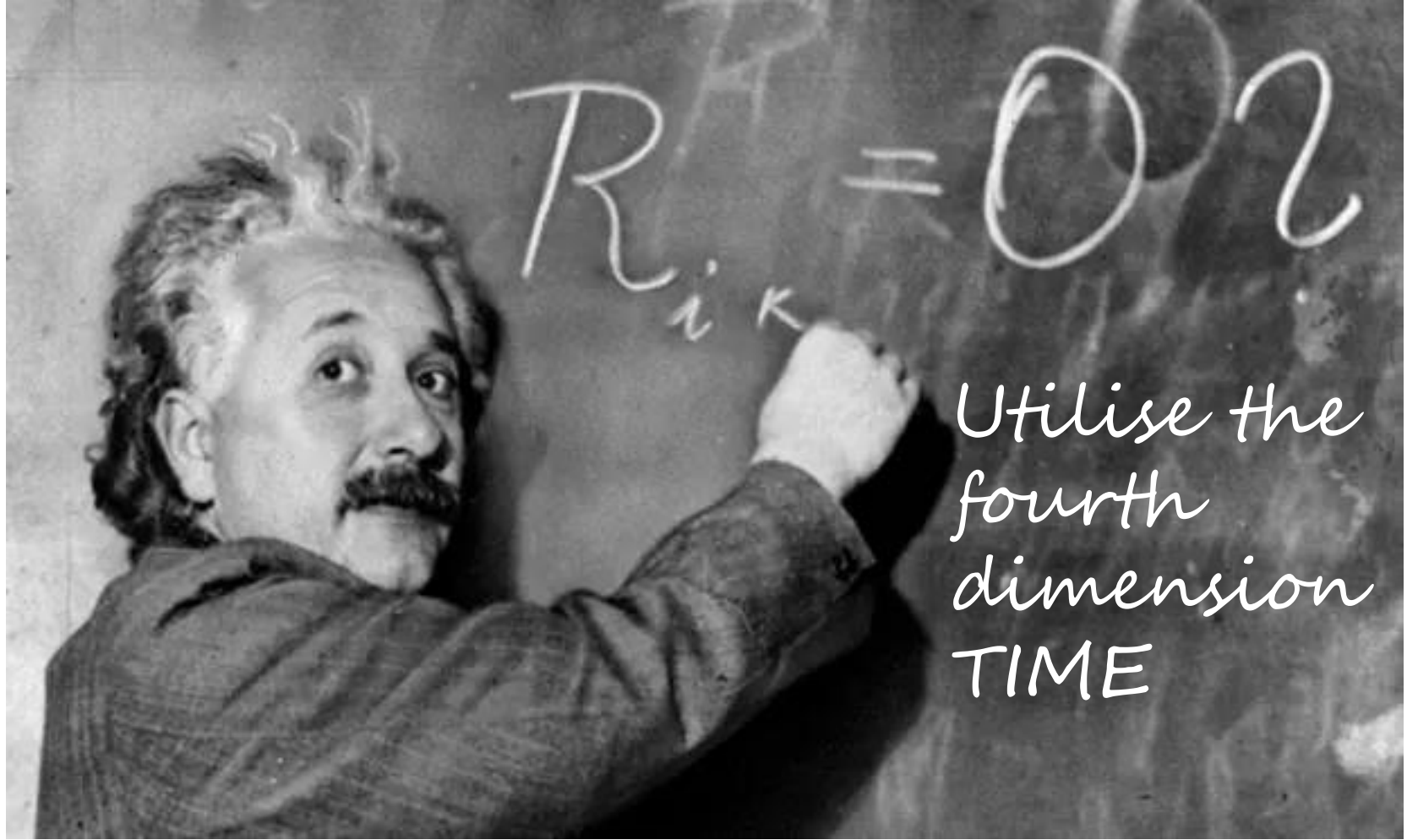
3 Kalman Filter

4 Thought experiment – theoretical example

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6 Conclusions – the introspective orifice meter

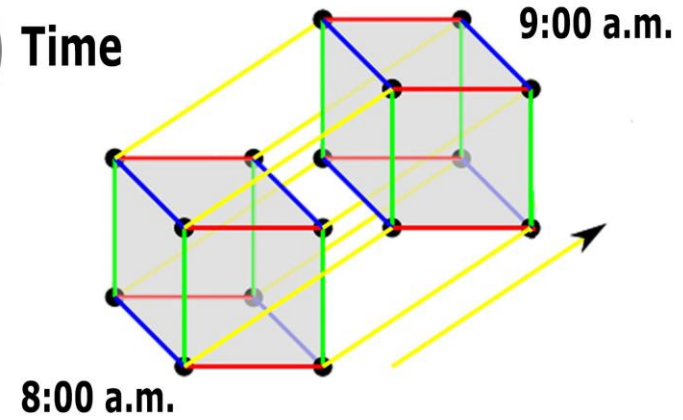
The next dimension



Special Theory of Relativity 1905



Time



Return to the Mass Balance



$$\dot{m}_t = EA_t Y C_d (2\rho \Delta P_t)^{1/2}$$



$$\dot{m}_r = EA_t K_r (2\rho \Delta P_r)^{1/2}$$



$$\dot{m}_{PPL} = AK_{PPL} (2\rho \Delta P_{PPL})^{1/2}$$

$$\dot{m}_t = \dot{m}_r = \dot{m}_{PPL}$$

$$EA_t Y C_d (2\rho \Delta P_t)^{1/2} = EA_t K_r (2\rho \Delta P_r)^{1/2} = AK_{PPL} (2\rho \Delta P_{PPL})^{1/2}$$

$$\dot{m}_t = \dot{m}_r = \dot{m}_{PPL}$$

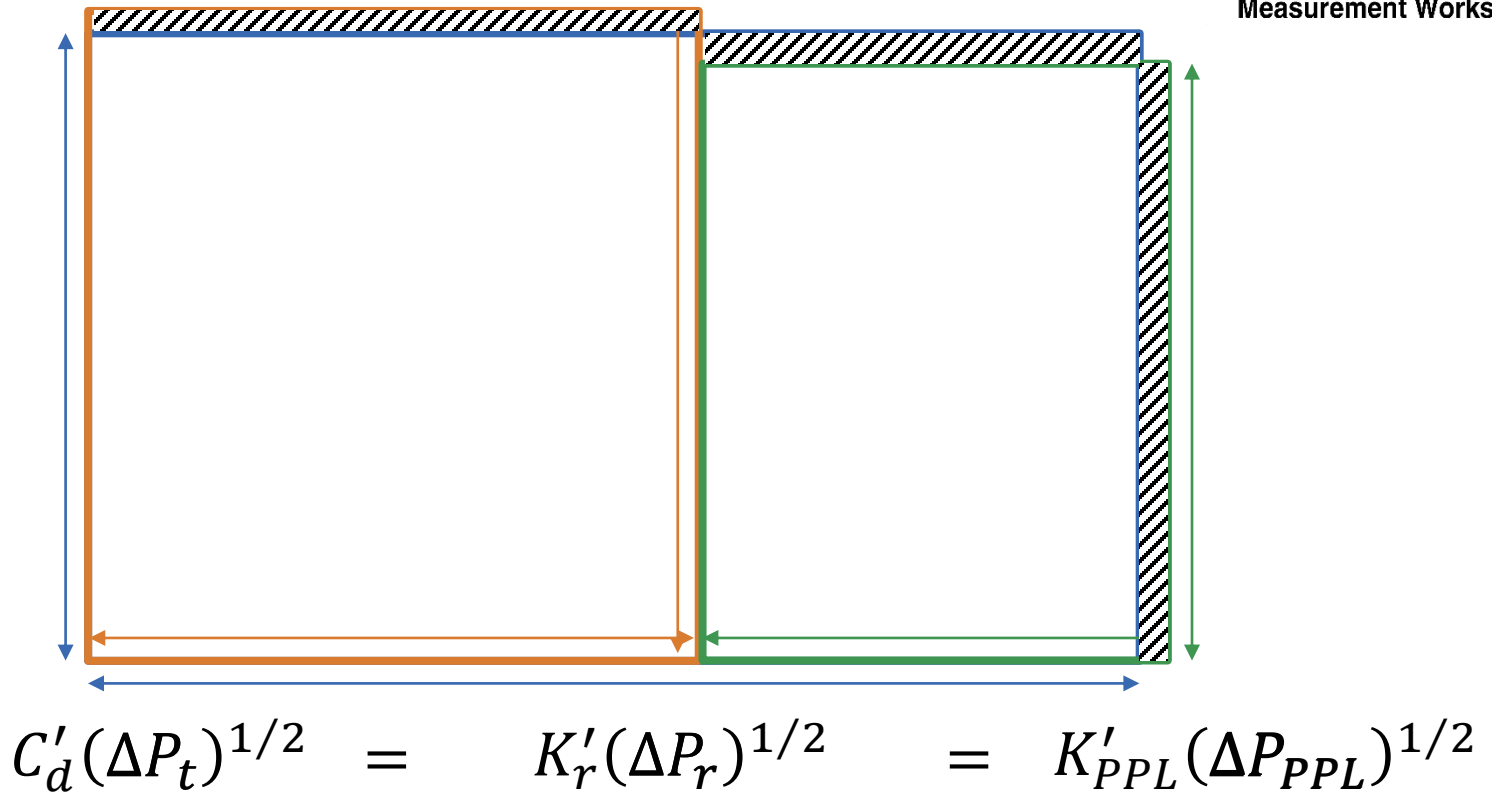
$$EA_t Y C_d (\cancel{2\rho} \Delta P_t)^{1/2} = EA_t K_r (\cancel{2\rho} \Delta P_r)^{1/2} = AK_{PPL} (\cancel{2\rho} \Delta P_{PPL})^{1/2}$$

$$\dot{m}_t = \dot{m}_r = \dot{m}_{PPL}$$

$$\underbrace{EA_t Y C_d}_{C'_d} (\Delta P_t)^{1/2} = \underbrace{EA_t K_r}_{K'_r} (\Delta P_r)^{1/2} = \underbrace{AK_{PPL}}_{K'_{PPL}} (\Delta P_{PPL})^{1/2}$$

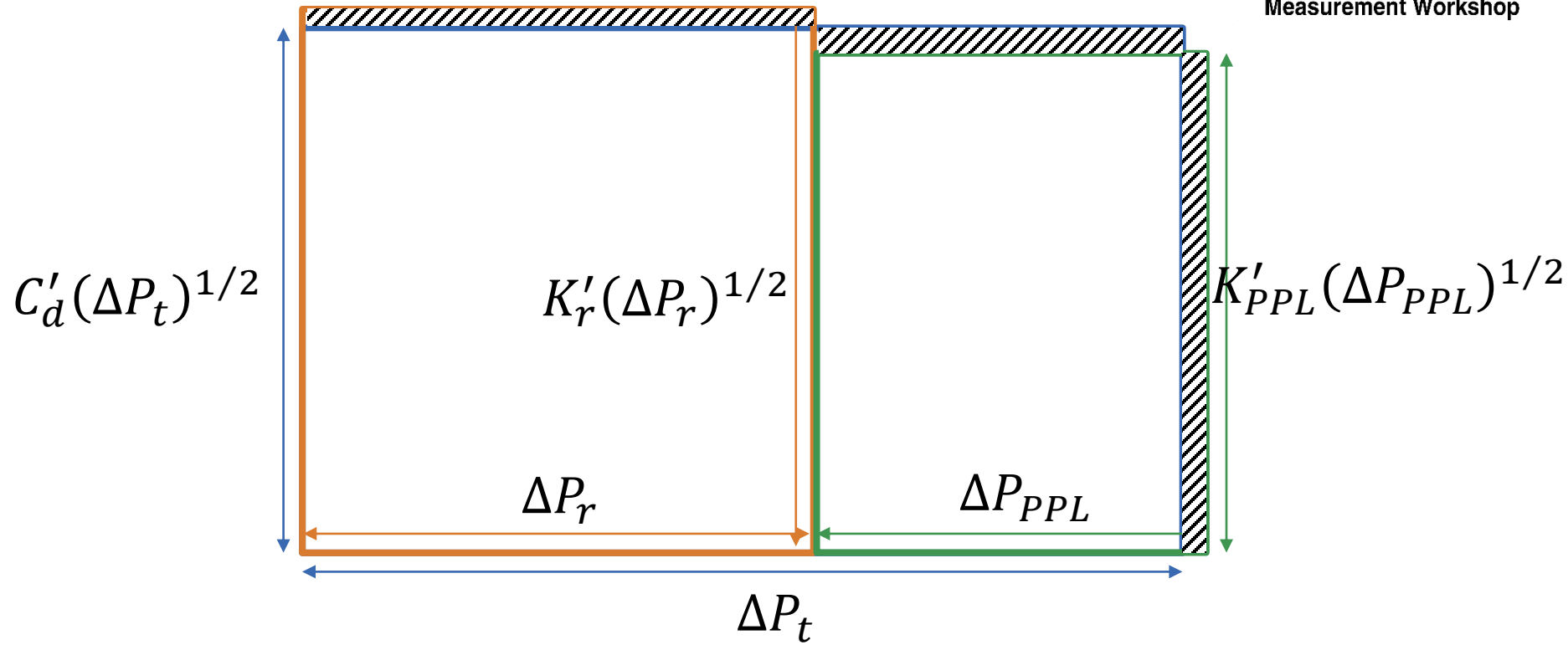
$$\dot{m}_t = \dot{m}_r = \dot{m}_{PPL}$$

Mass Flow and Pressure Drop Constraints

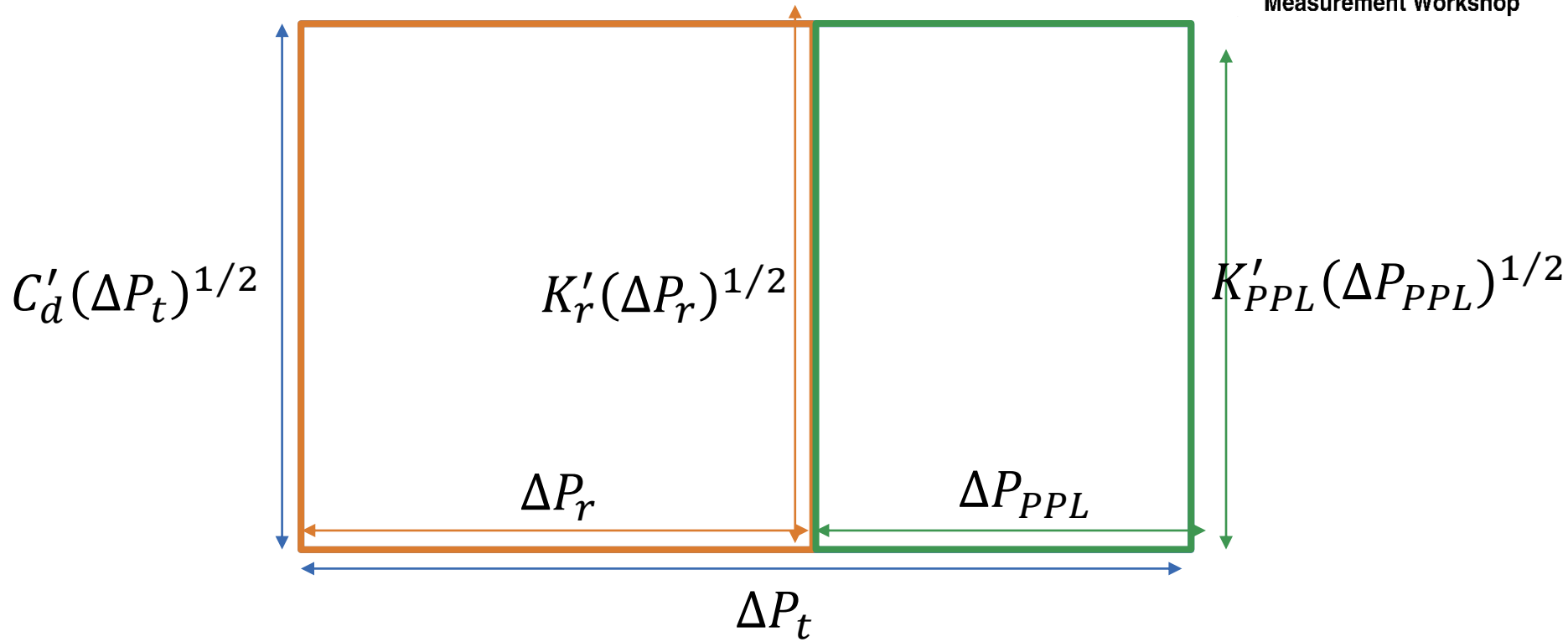


$$\dot{m}_t = \dot{m}_r = \dot{m}_{PPL}$$

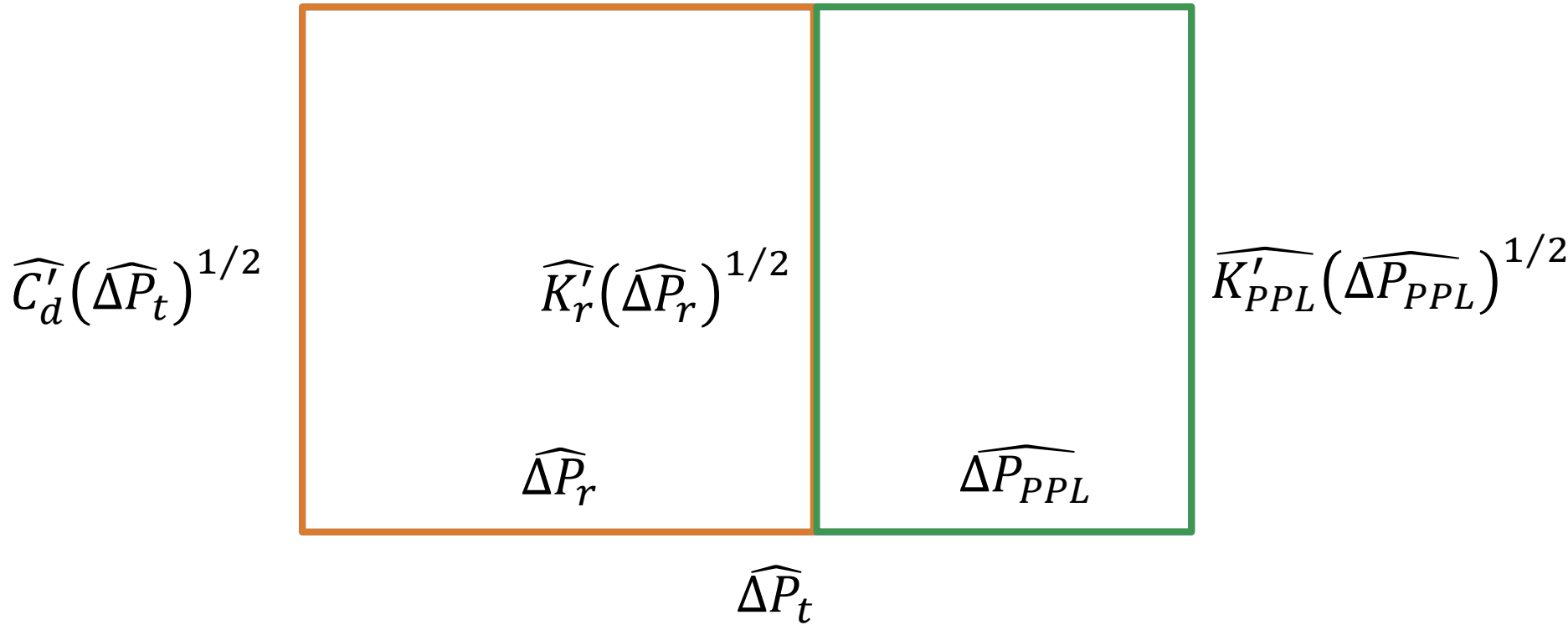
Mass Flow and Pressure Drop Constraints

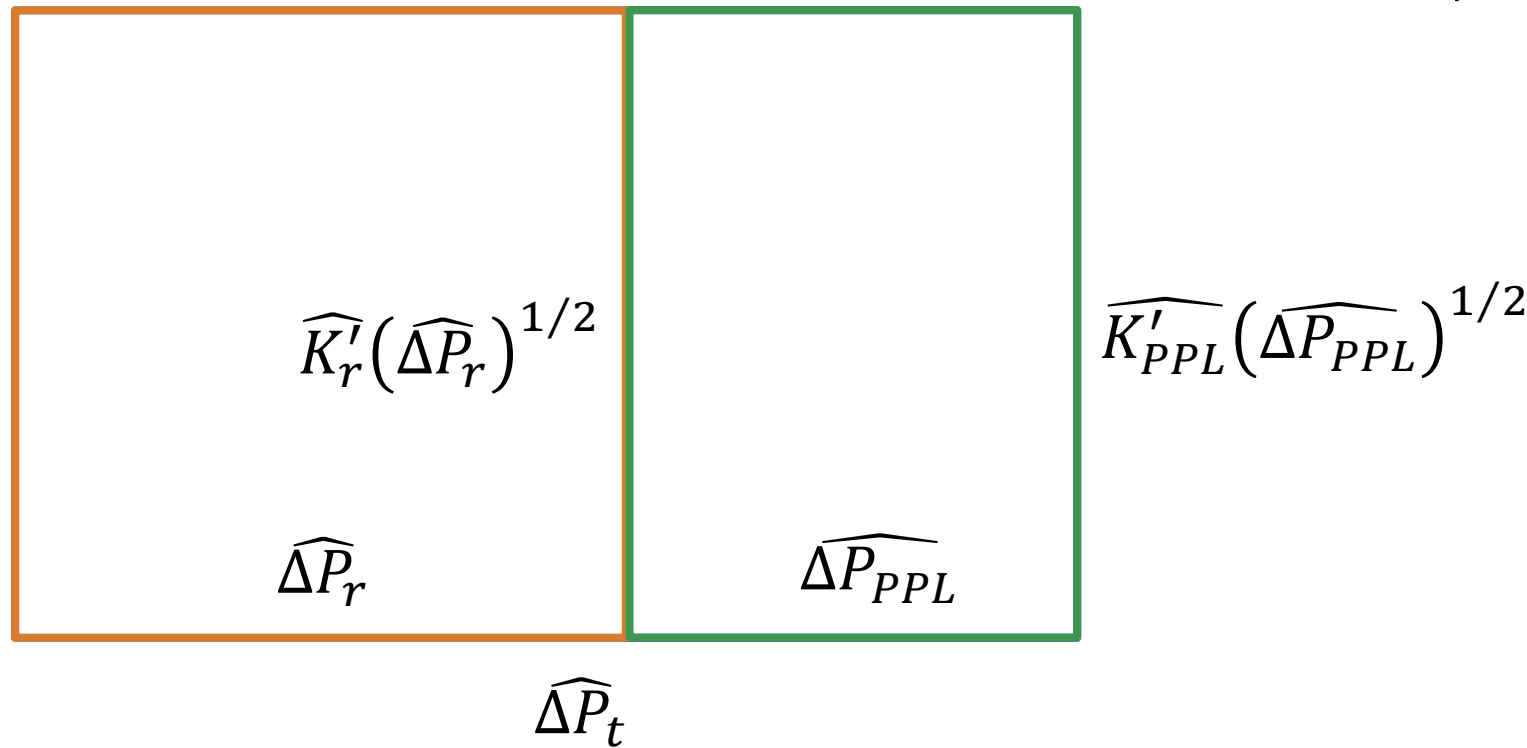


Maximum Likelihood Uncertainty



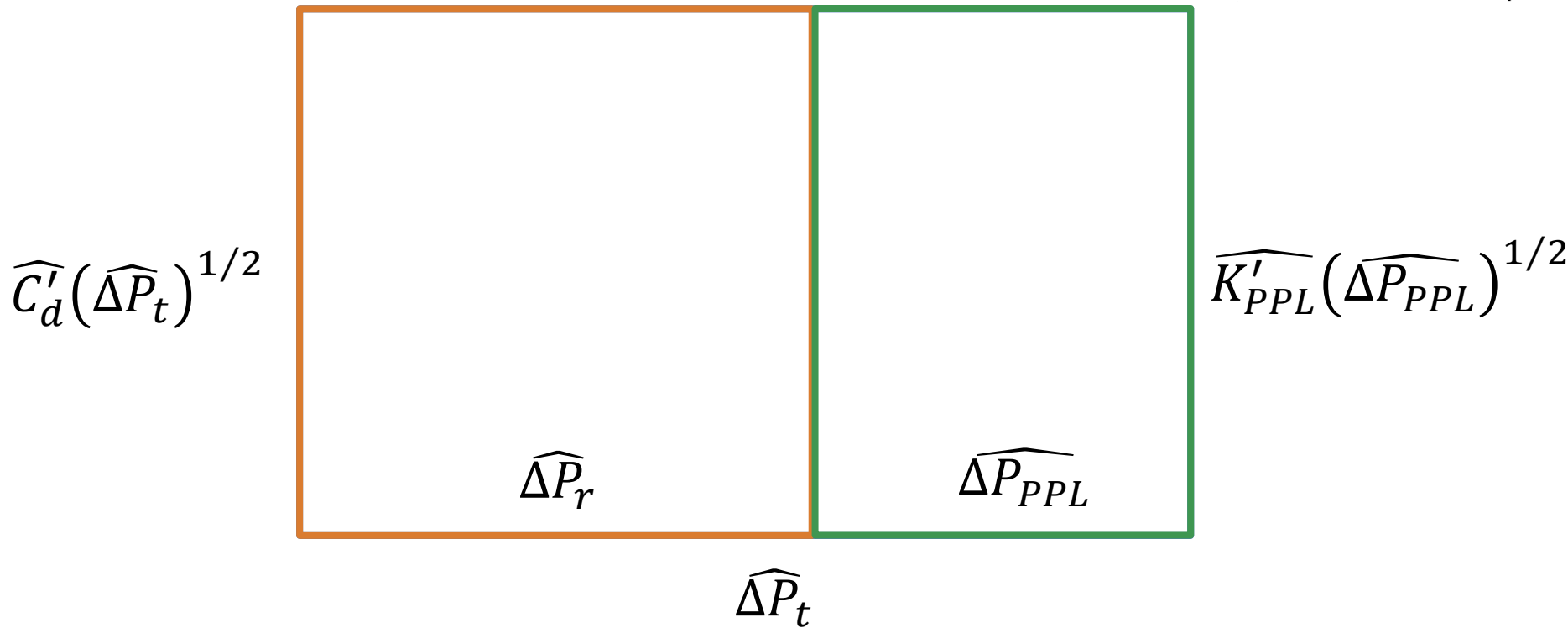
Maximum Likelihood Uncertainty





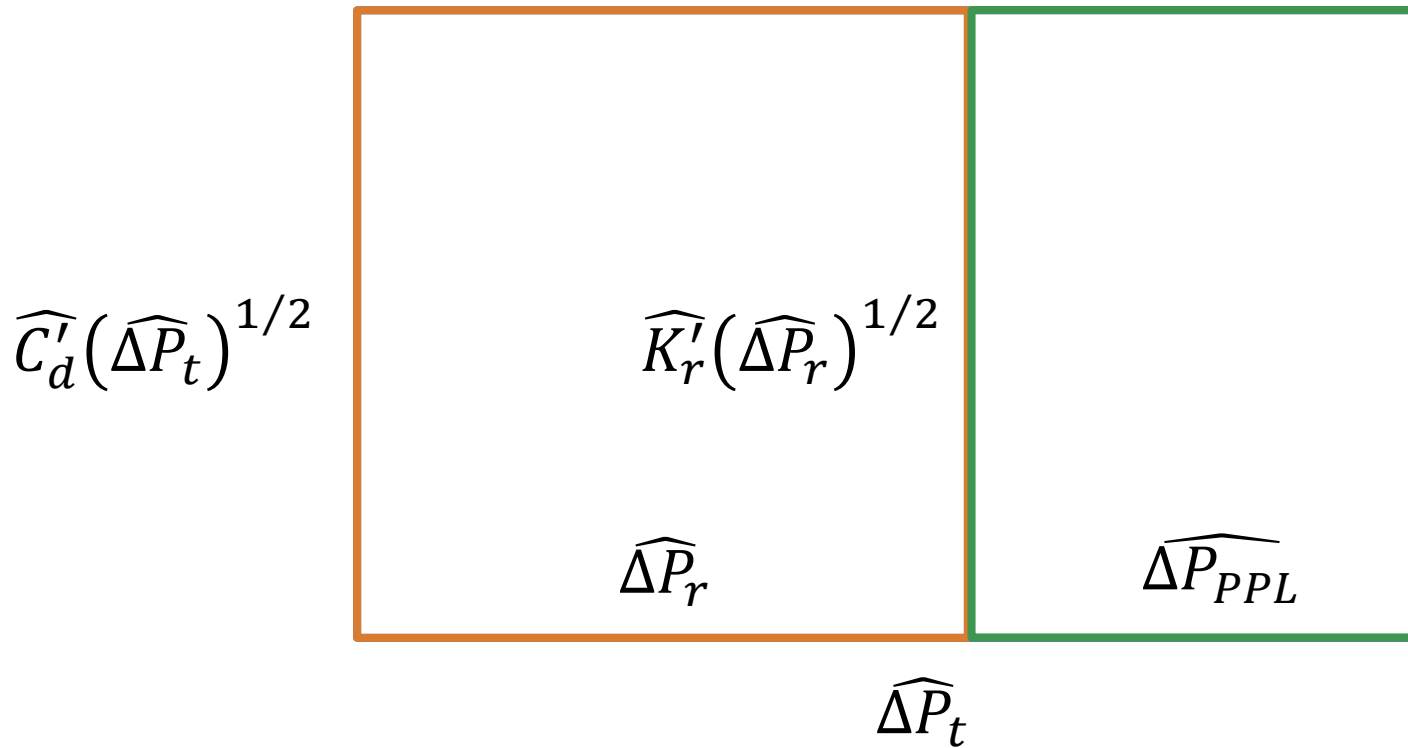
$$\widehat{\dot{m}_{MLU}} = \widehat{C}'_d(\widehat{\Delta P}_t)^{1/2}(2\rho)^{1/2}$$

MLU mass flow,
reduced
uncertainty



$$\widehat{\dot{m}_{MLU}} = \widehat{K}'_r(\widehat{\Delta P}_r)^{1/2} (2\rho)^{1/2}$$

MLU mass flow,
reduced
uncertainty



$$\widehat{\dot{m}}_{MLU} = \widehat{K}'_{PPL}(\widehat{\Delta P}_{PPL})^{1/2}(2\rho)^{1/2}$$

MLU mass flow,
reduced
uncertainty

$$\widehat{C}_d'(\widehat{\Delta P}_t)^{1/2}$$

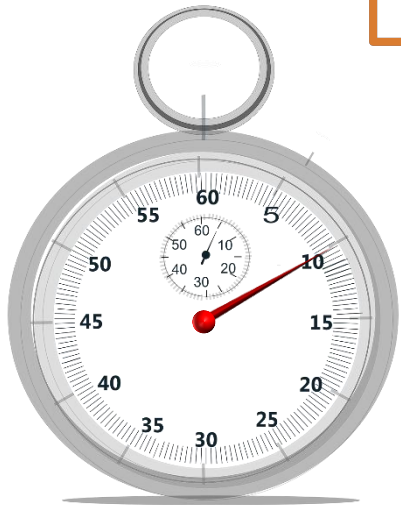
$$\widehat{K}_r'(\widehat{\Delta P}_r)^{1/2}$$

$$\widehat{K}_{PPL}'(\widehat{\Delta P}_{PPL})^{1/2}$$

$$\widehat{\Delta P}_r$$

$$\widehat{\Delta P}_{PPL}$$

$$\widehat{\Delta P}_t$$



$$\widehat{C'_{d,1}}(\widehat{\Delta P_{t,1}})^{1/2}$$

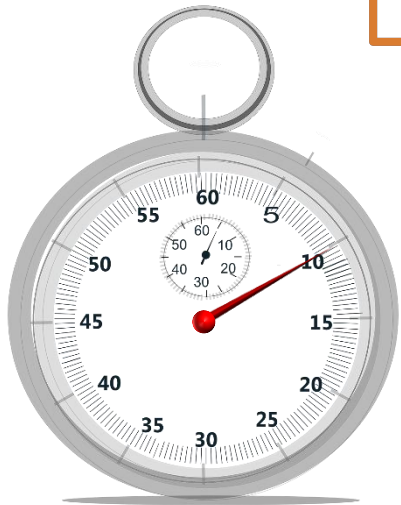
$$\widehat{K'_{r,1}}(\widehat{\Delta P_{r,1}})^{1/2}$$

$$\widehat{K'_{PPL,1}}(\widehat{\Delta P_{PPL,1}})^{1/2}$$

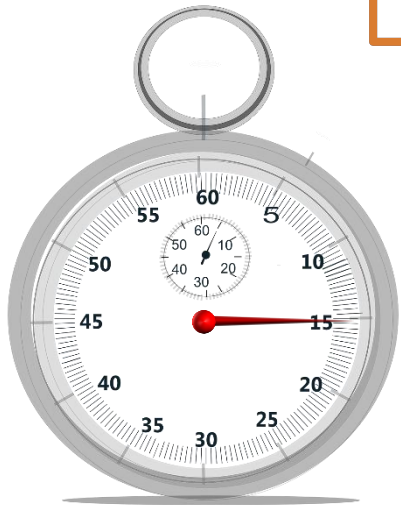
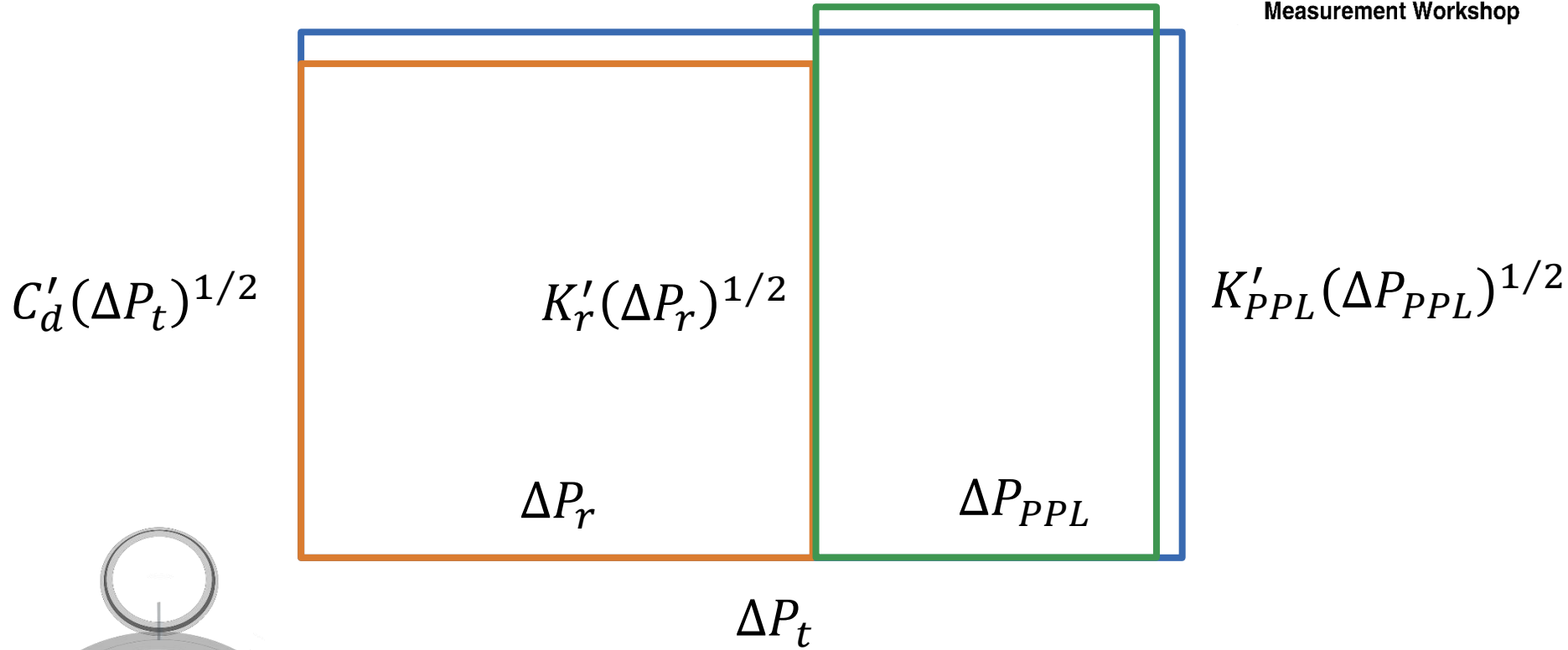
$$\widehat{\Delta P_{r,1}}$$

$$\widehat{\Delta P_{PPL,1}}$$

$$\widehat{\Delta P_{t,1}}$$



MLU time, $t=1$



MLU time, $t=2$

$$\widehat{C'_d}(\widehat{\Delta P_t})^{1/2}$$

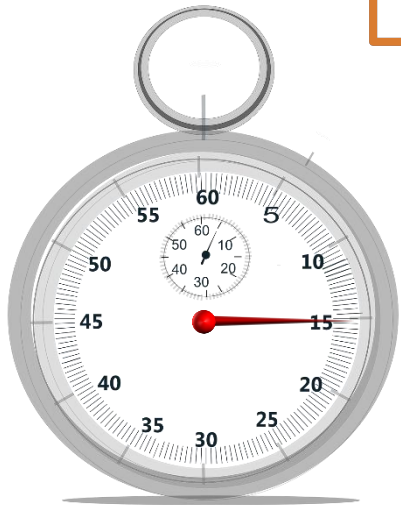
$$\widehat{K'_r}(\widehat{\Delta P_r})^{1/2}$$

$$\widehat{K'_{PPL}}(\widehat{\Delta P_{PPL}})^{1/2}$$

$$\widehat{\Delta P_r}$$

$$\widehat{\Delta P_{PPL}}$$

$$\widehat{\Delta P_t}$$



MLU time, $t=2$

$$\widehat{C'_{d,2}}(\widehat{\Delta P_{t,2}})^{1/2}$$

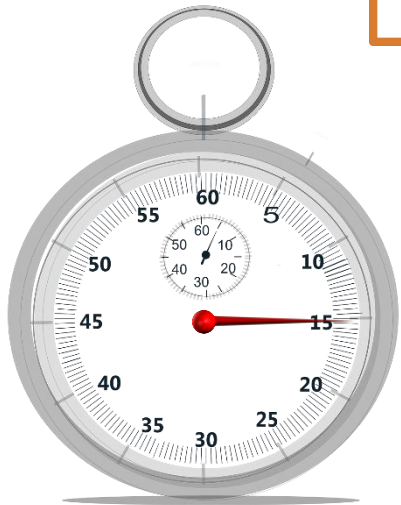
$$\widehat{K'_{r,2}}(\widehat{\Delta P_{r,2}})^{1/2}$$

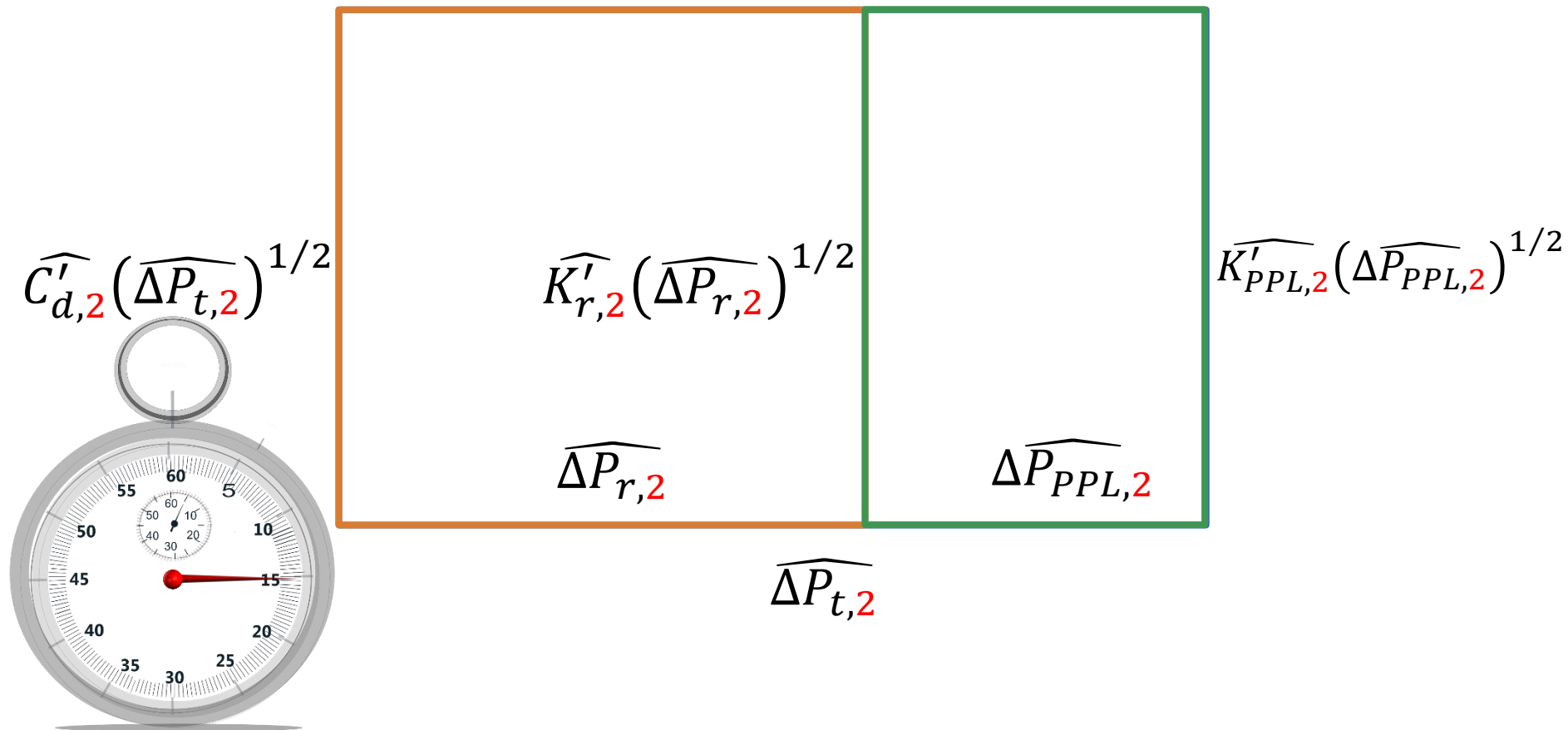
$$\widehat{K'_{PPL,2}}(\widehat{\Delta P_{PPL,2}})^{1/2}$$

$$\widehat{\Delta P_{r,2}}$$

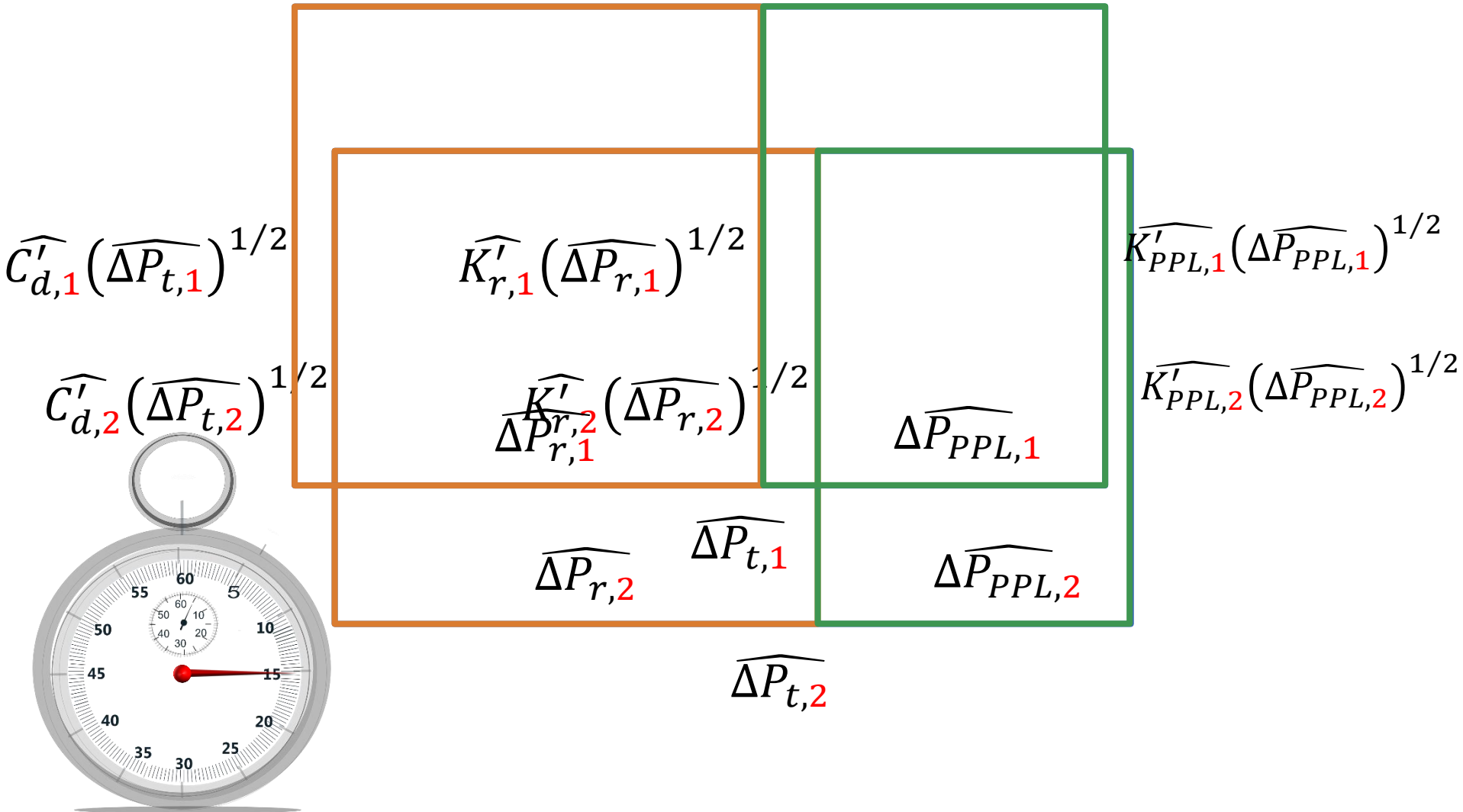
$$\widehat{\Delta P_{PPL,2}}$$

$$\widehat{\Delta P_{t,2}}$$



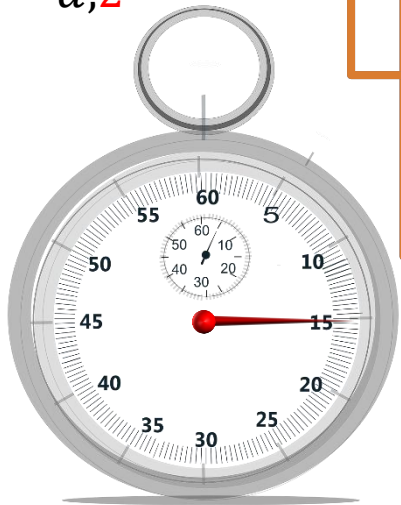


MLU time, t=1 and t=2



MLU time, t=1 and t=2

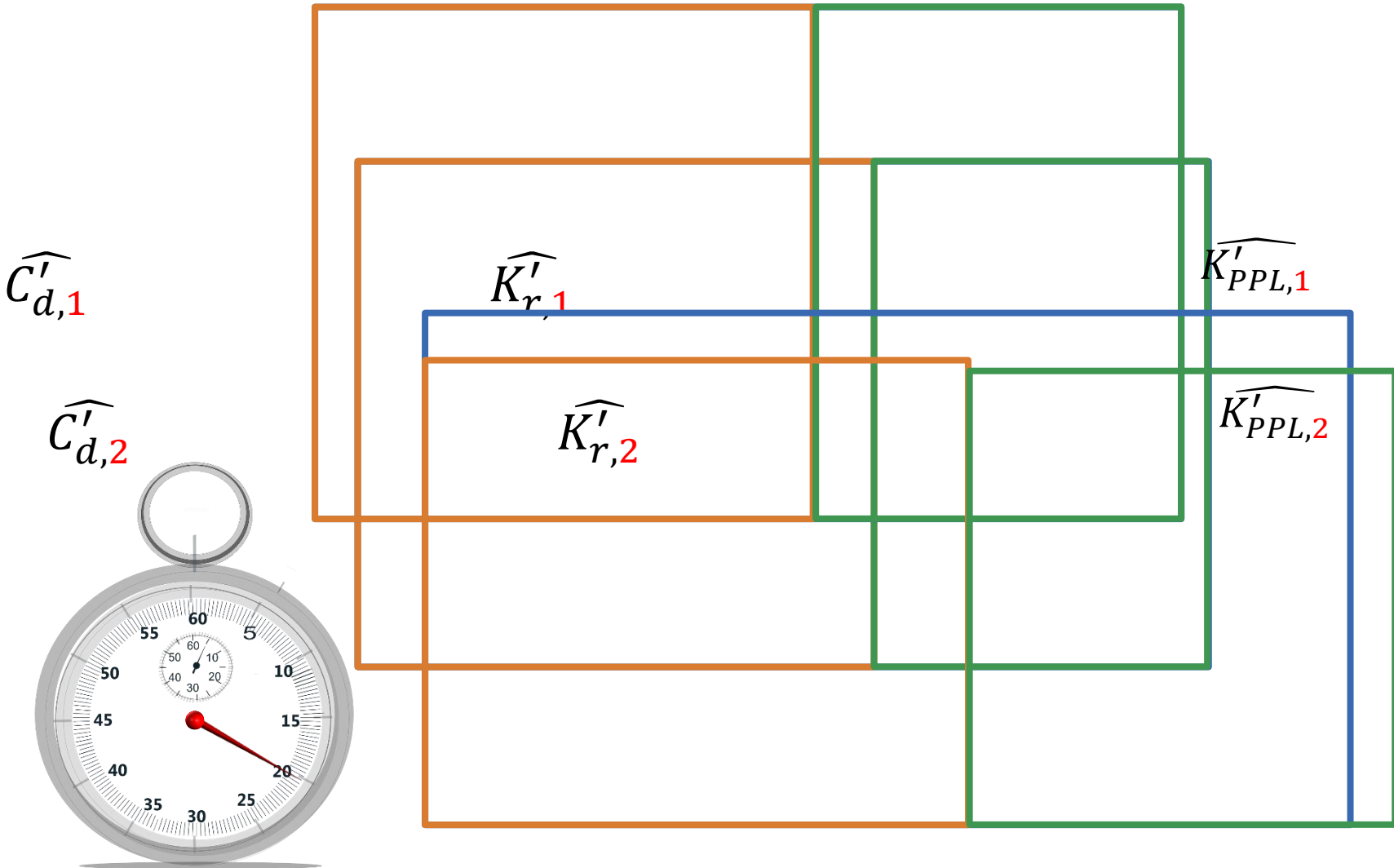
$$\widehat{C'_{d,1}} = \widehat{C'_{d,2}}$$



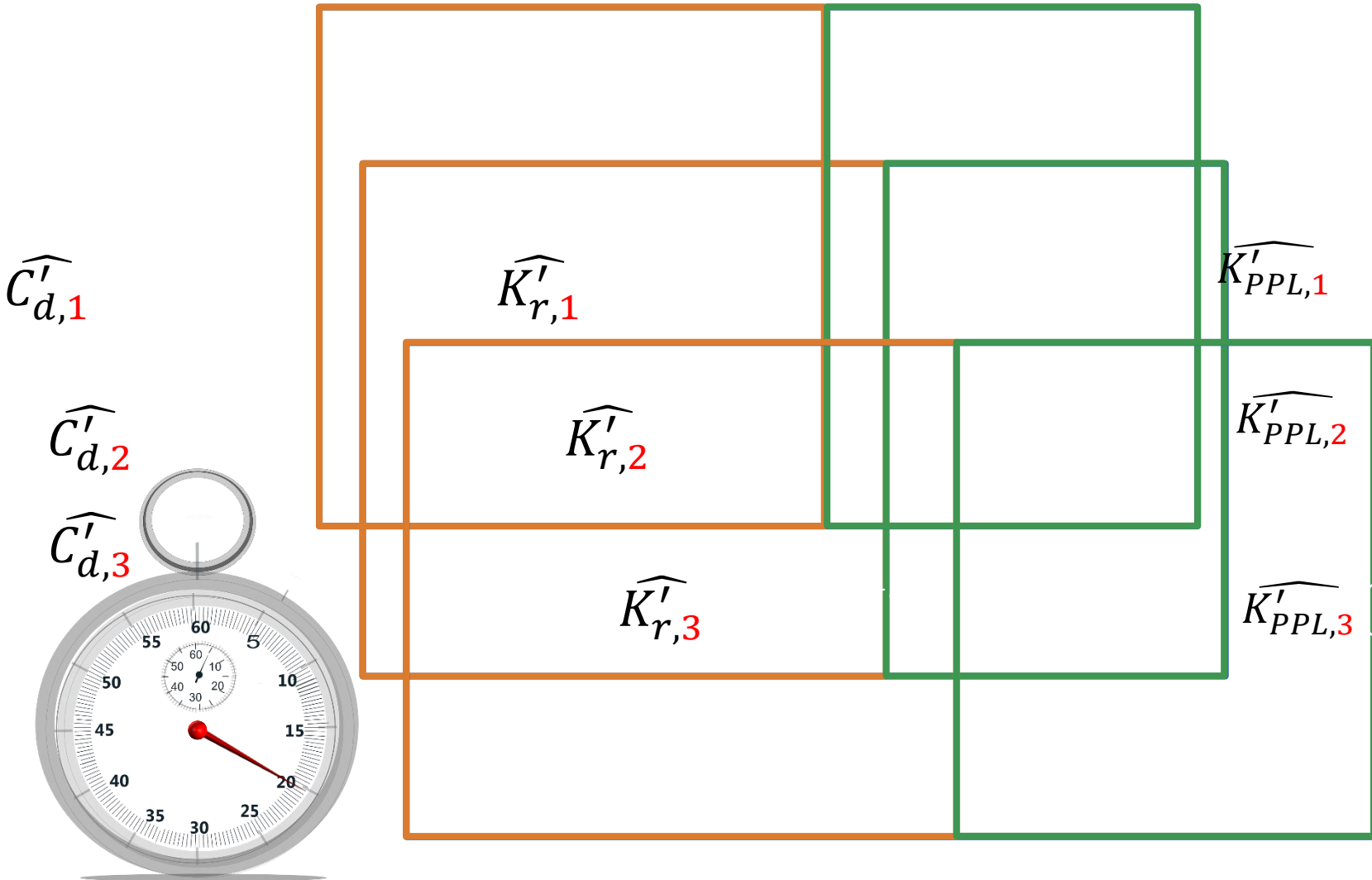
$$\widehat{K'_{r,1}} = \widehat{K'_{r,2}}$$

$$\widehat{K'_{PPL,1}} = \widehat{K'_{PPL,2}}$$

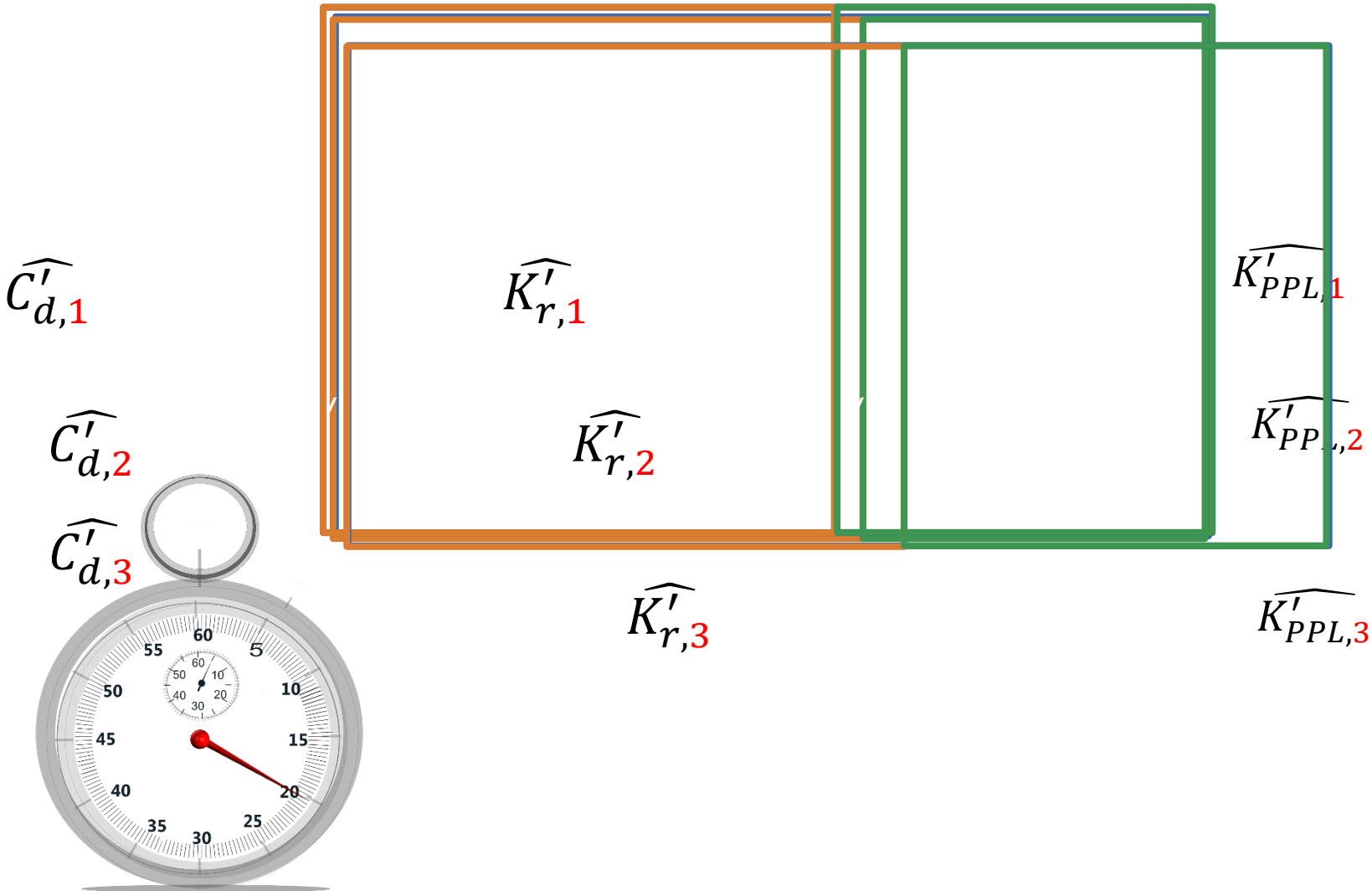
MLU time, $t=1$, $t=2$ and $t=3$



MLU time, $t=1$, $t=2$ and $t=3$



Flow Coefficient Constraints

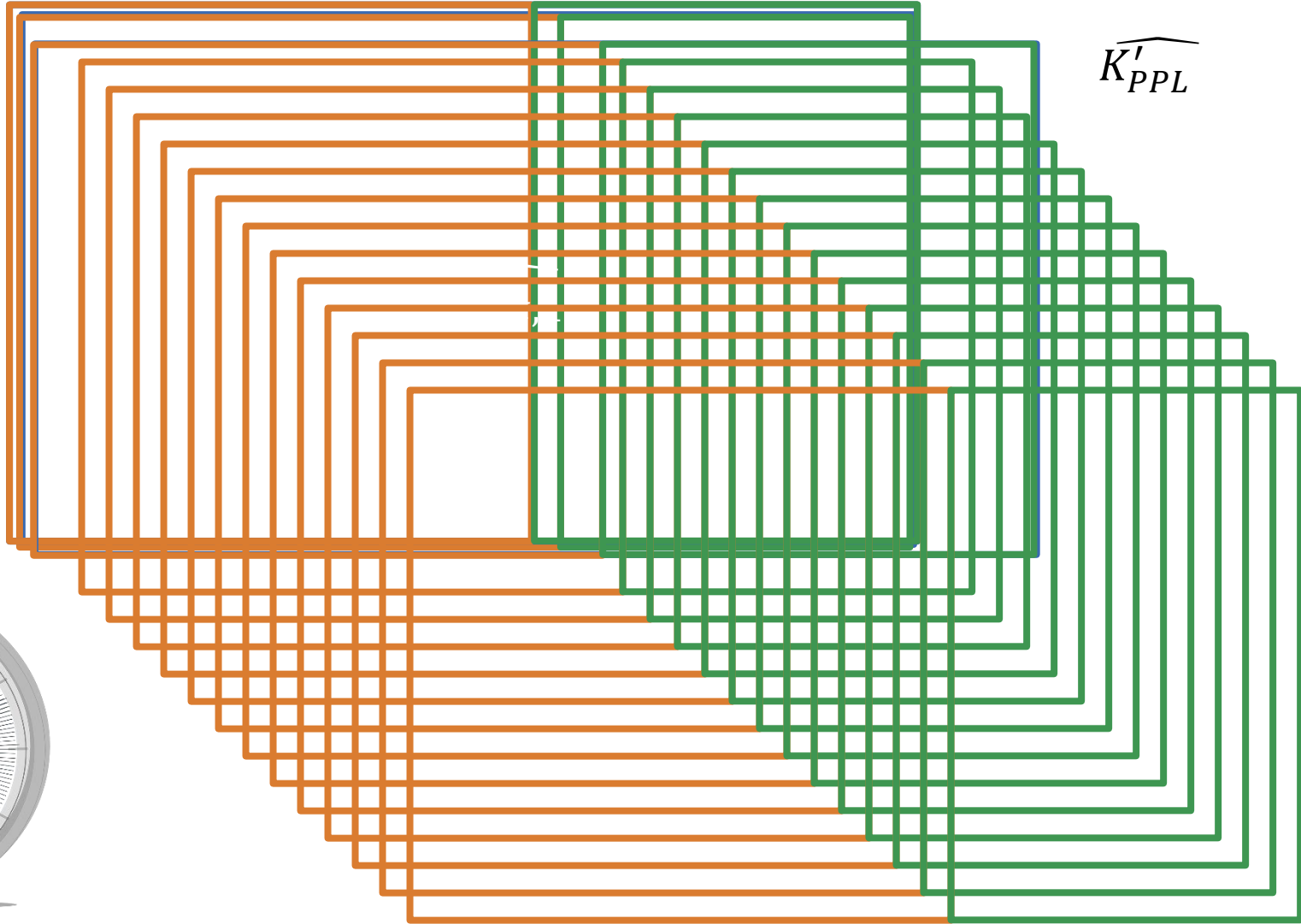


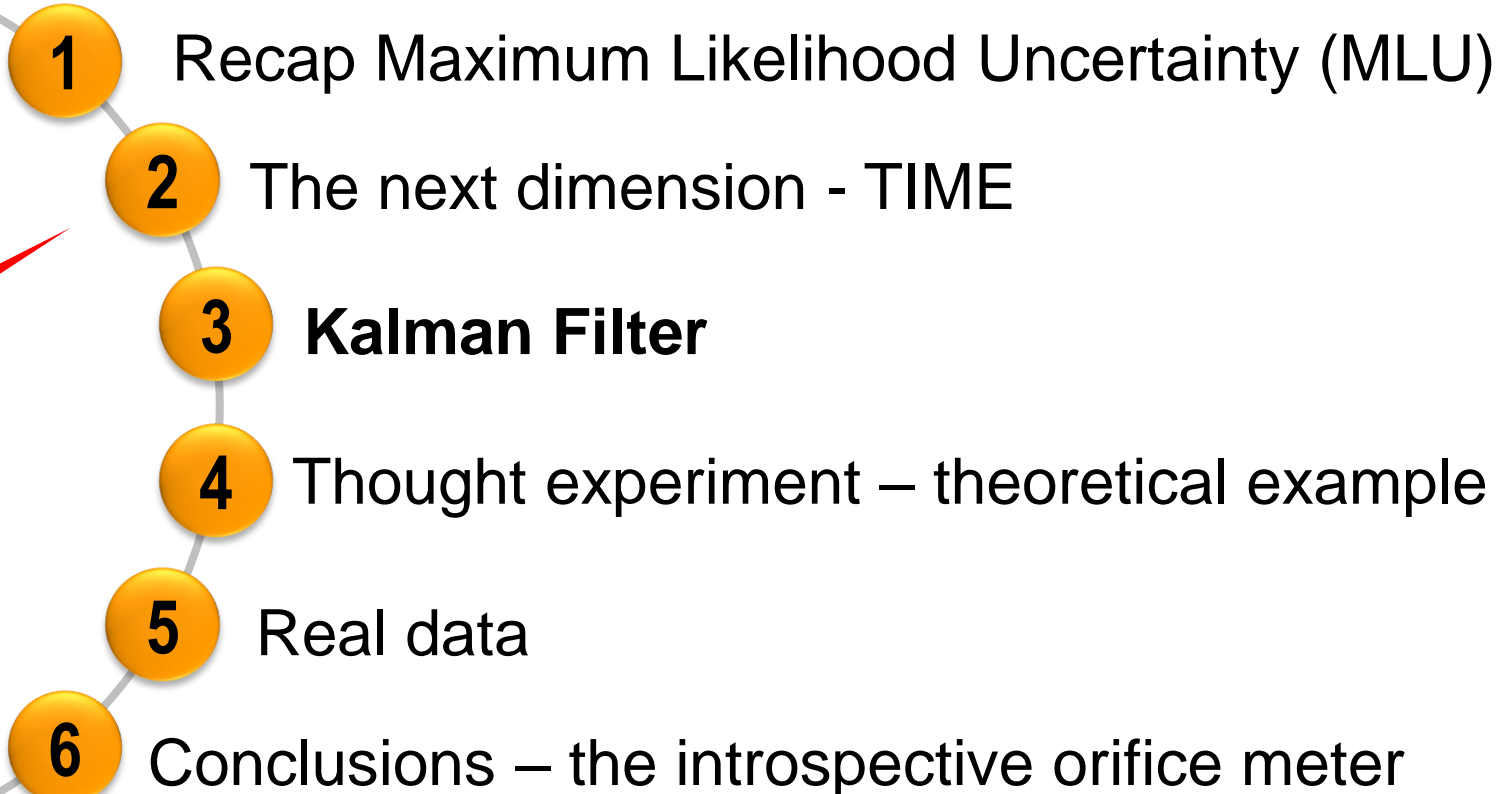
Flow Coefficient Constraints

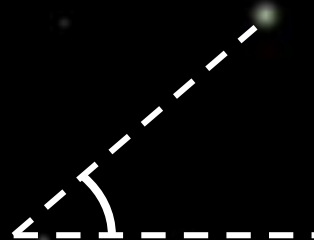
$$\widehat{K'_r}$$

$$\widehat{K'_{PPL}}$$

$$\widehat{C'_d}$$

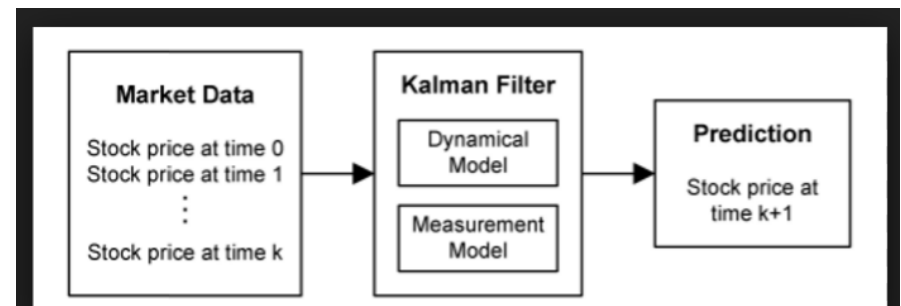
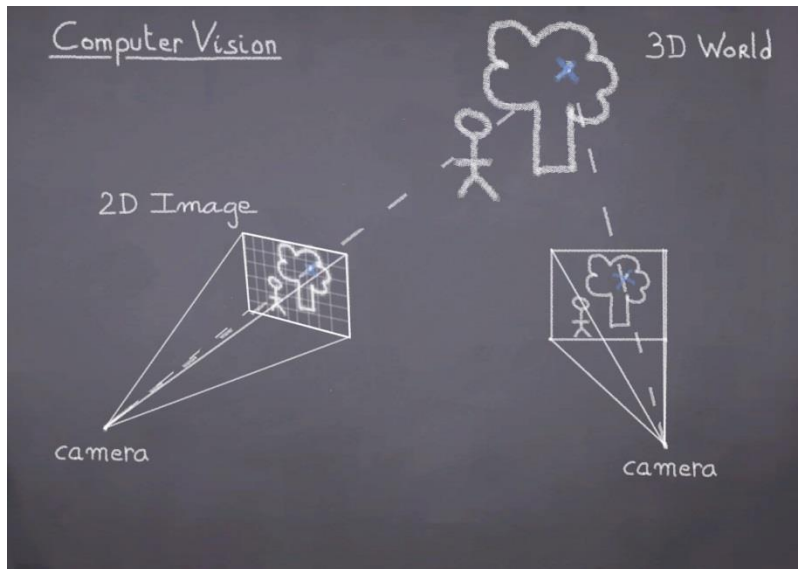
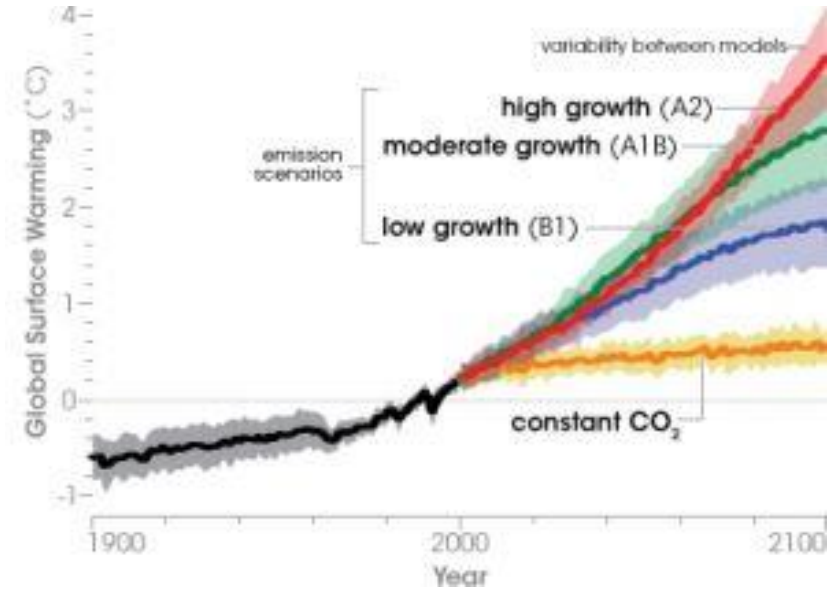


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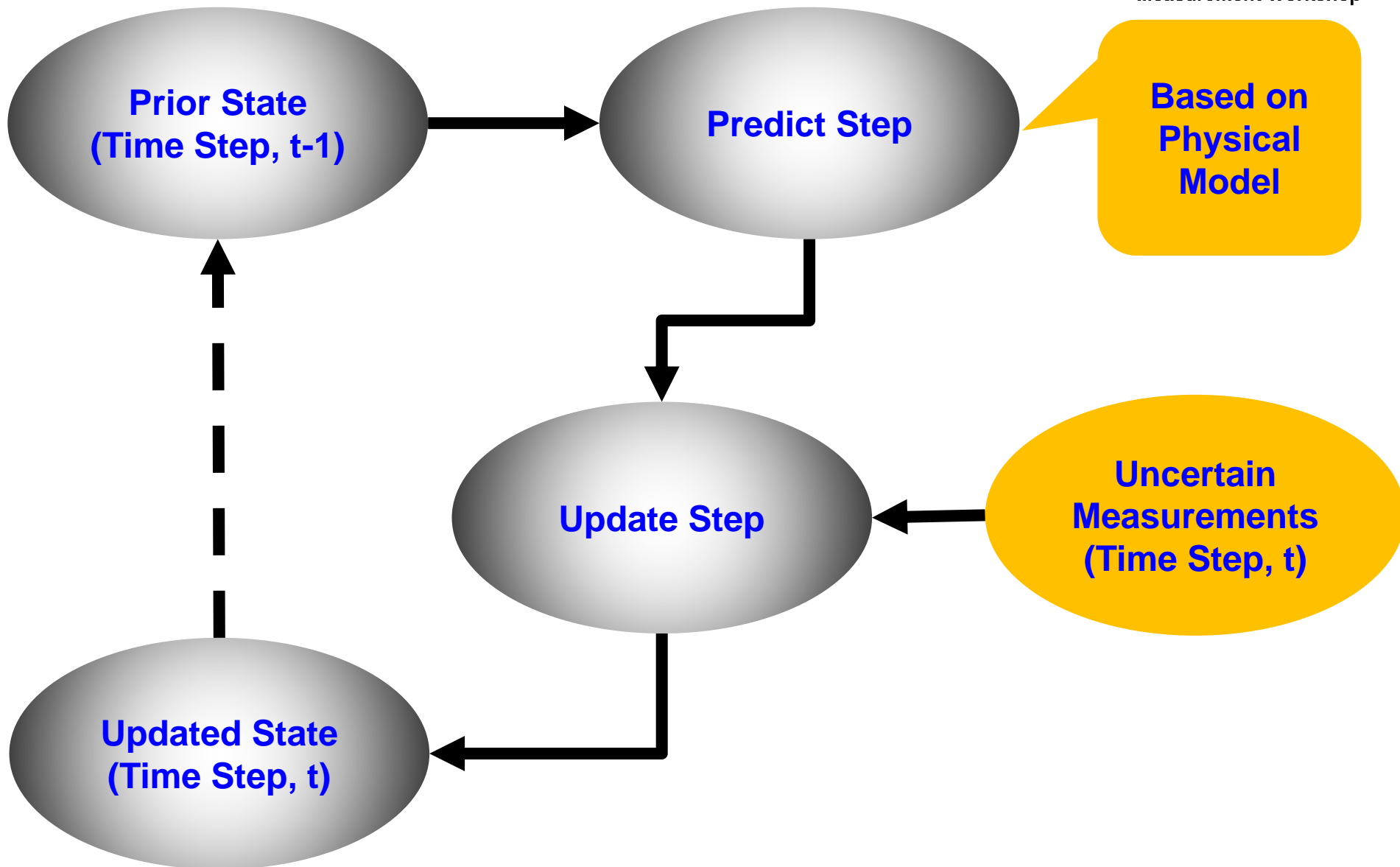




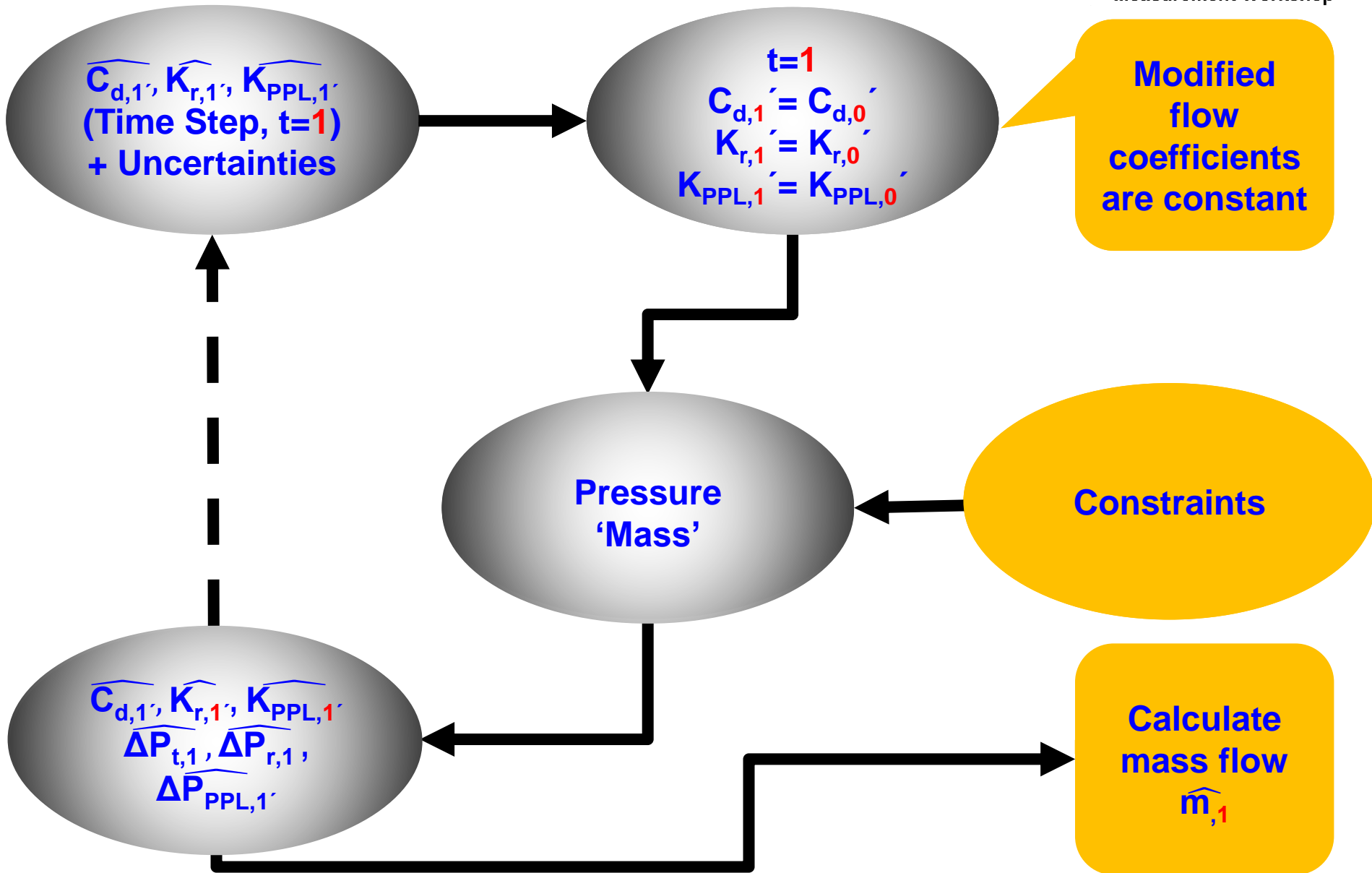
Kalman Filter Applications

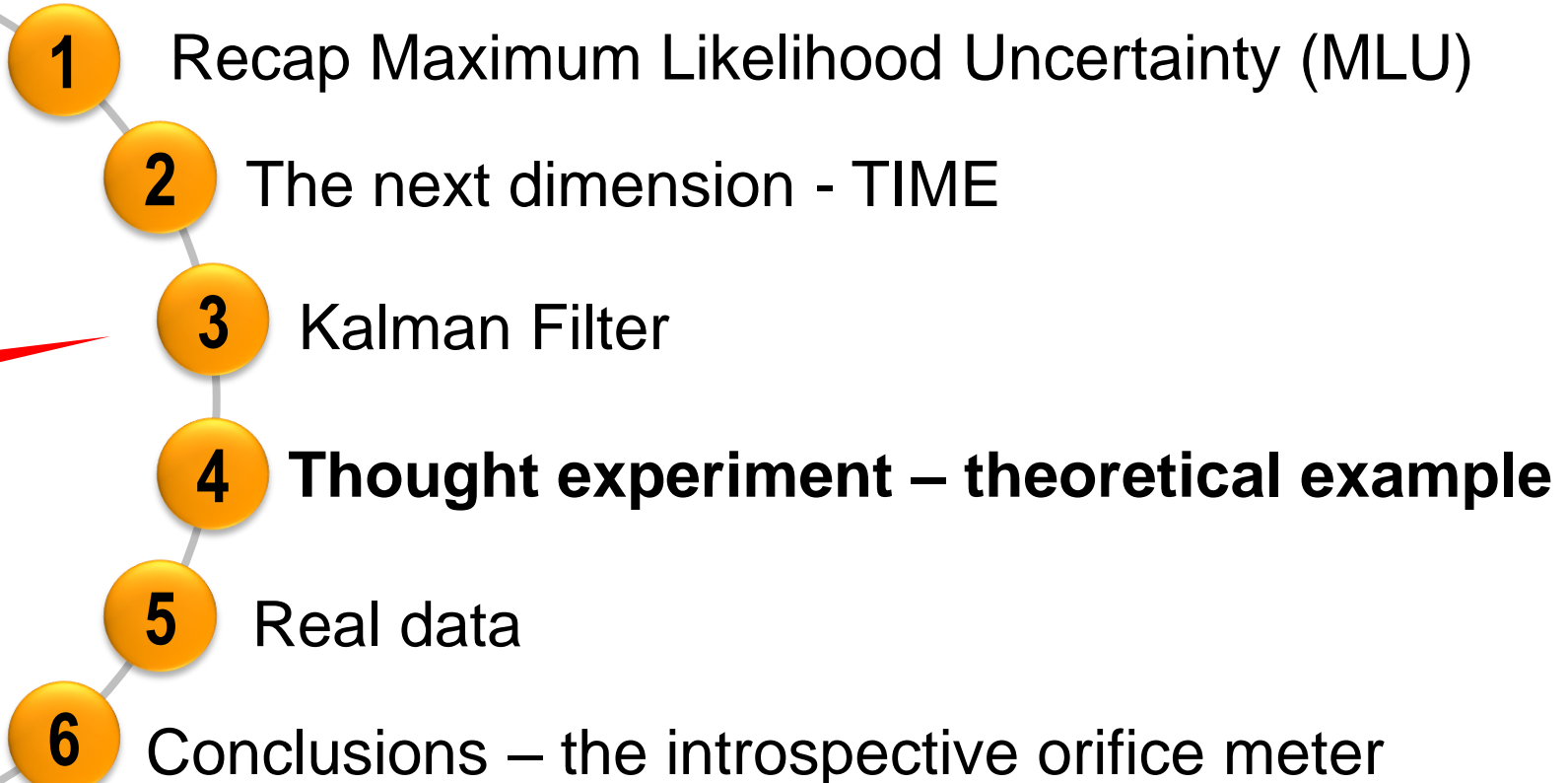


Kalman Filter Algorithm

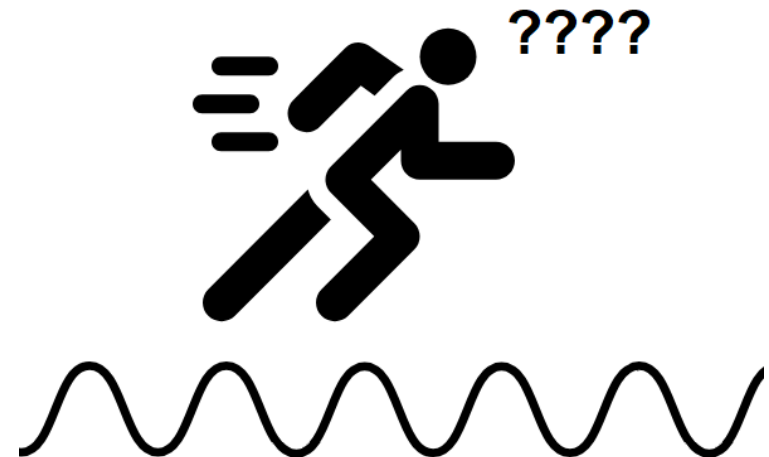
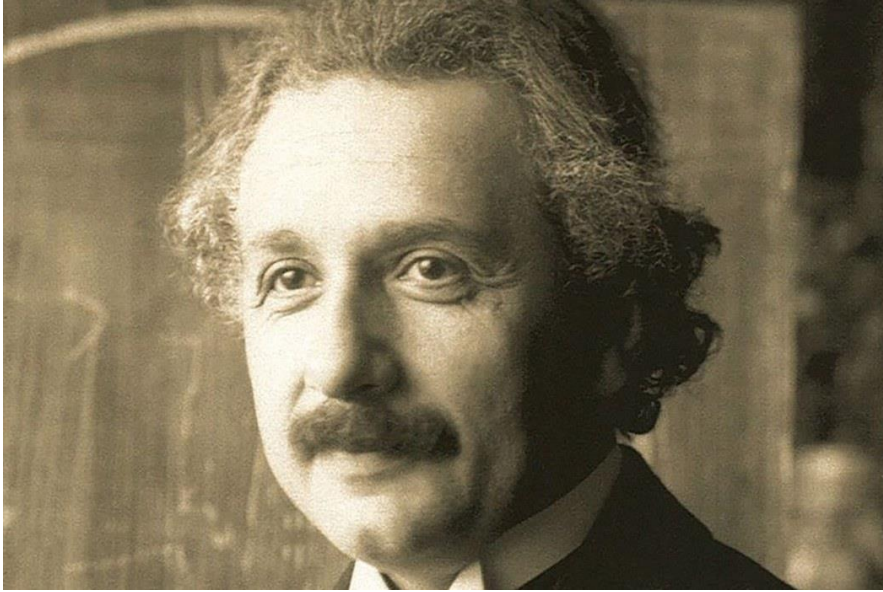


Kalman Filter Algorithm - MLU

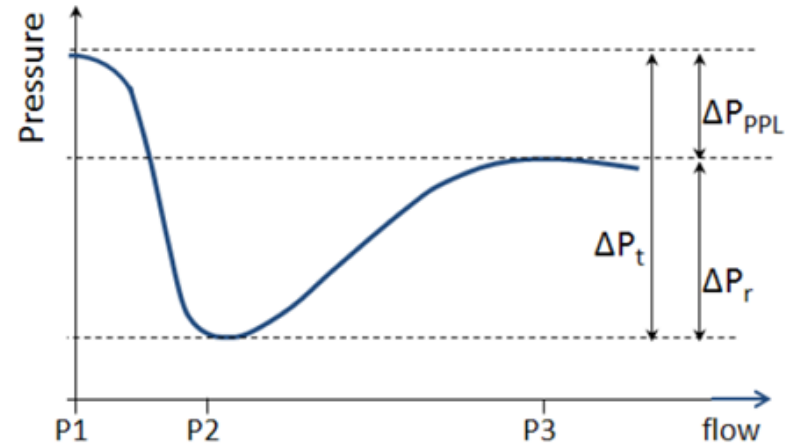
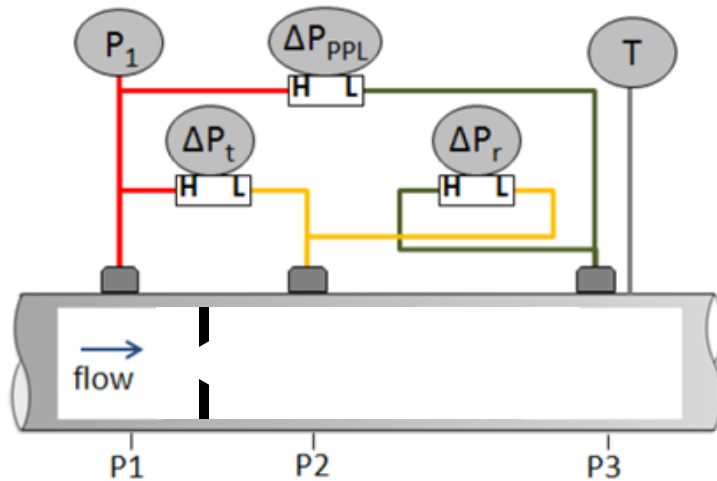


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Thought Experiment



Three DP Meters – Perfect Information

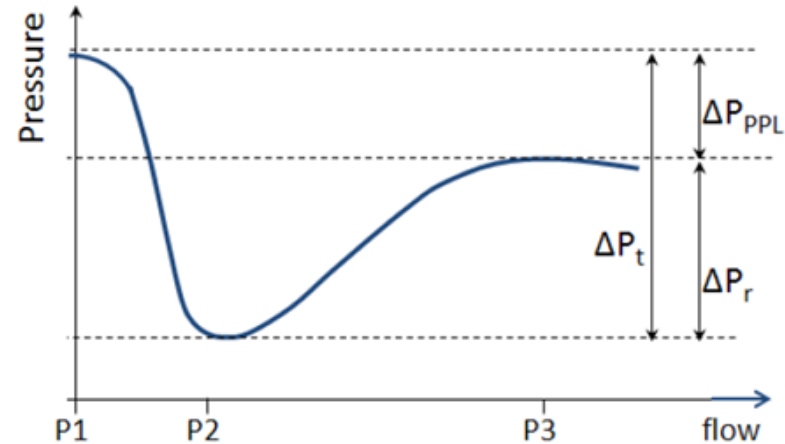
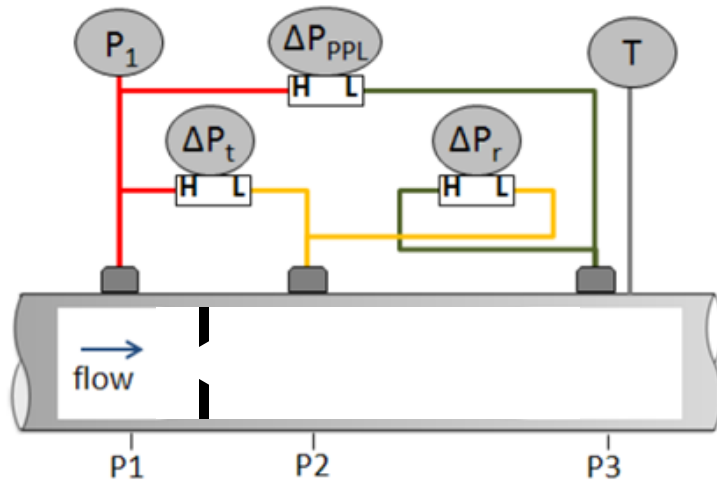


$$\dot{m}_t = EA_t Y C_d (2\rho \Delta P_t)^{1/2}$$

$$\dot{m}_r = EA_t K_r (2\rho \Delta P_r)^{1/2}$$

$$\dot{m}_{PPL} = AK_{PPL} (2\rho \Delta P_{PPL})^{1/2}$$

Three DP Meters – Perfect Information



$$\dot{m} = EA_t Y C_d (2\rho \Delta P_t)^{1/2}$$

$$\dot{m} = EA_t K_r (2\rho \Delta P_r)^{1/2}$$

$$\dot{m} = AK_{PPL} (2\rho \Delta P_{PPL})^{1/2}$$

4", 0.5 Beta Orifice DP Meter

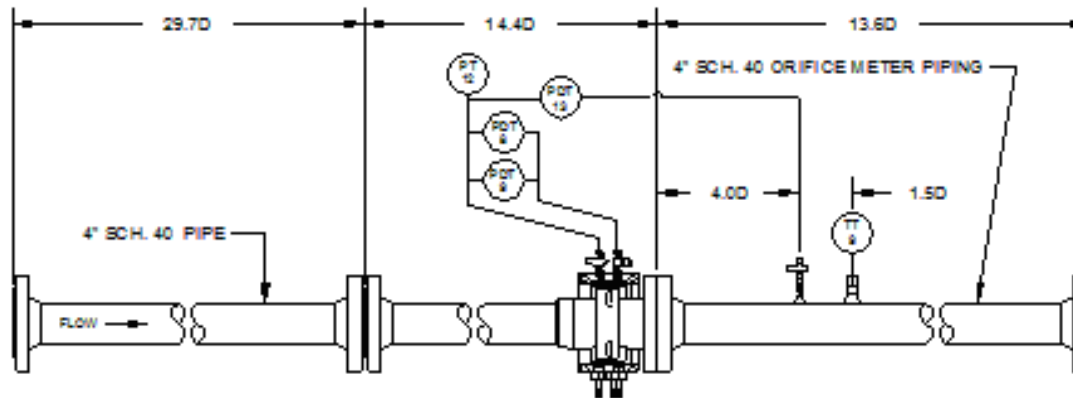


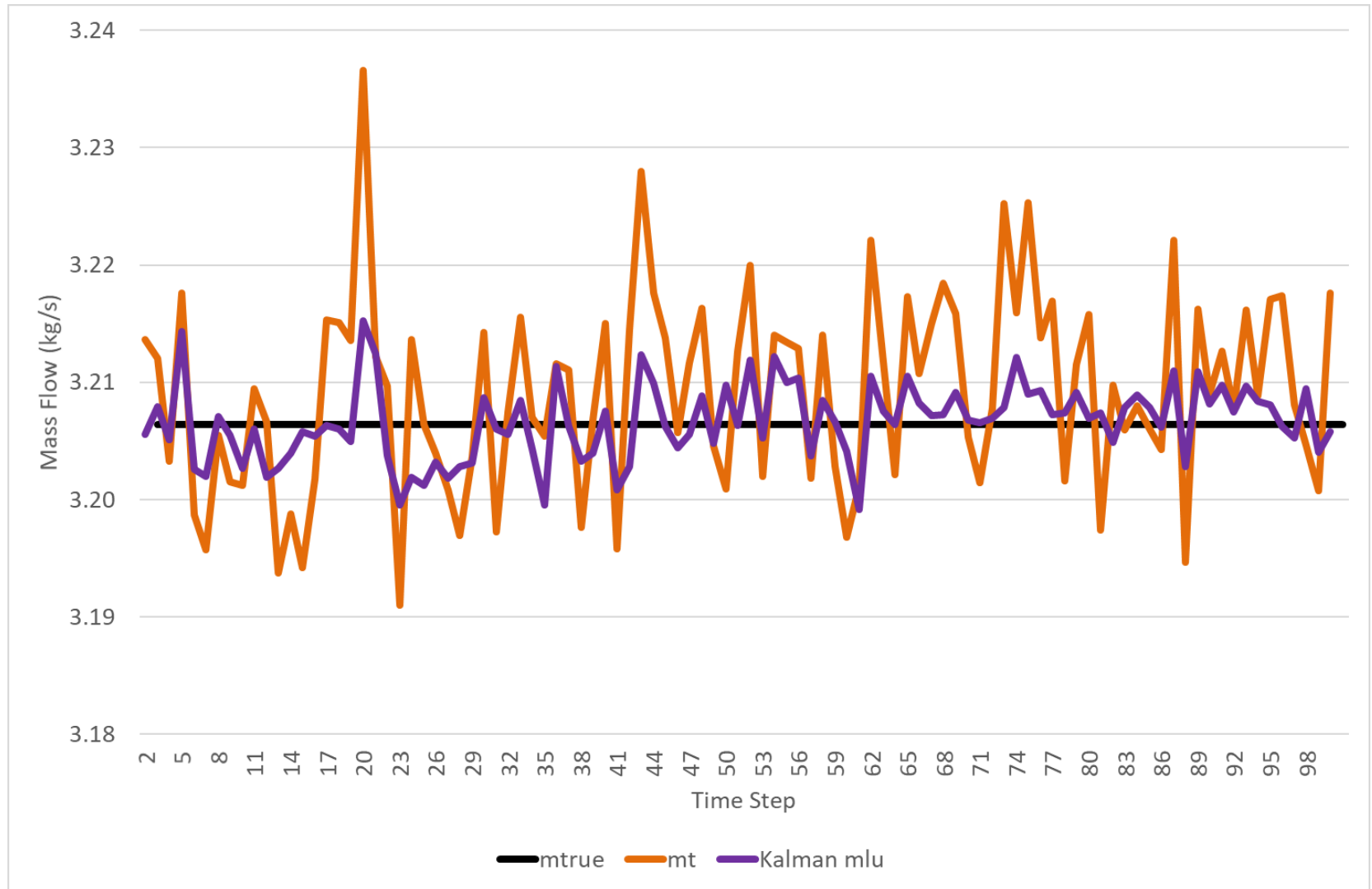
Table 1 – Hypothetical 4", 0.5 Beta Orifice DP Meter Variable and Parameter Uncertainties

Variable / Parameter	Unit	'True' Value	Percent Uncertainty	Absolute Uncertainty
Mass Flow	kg/s	3.2064		
DP_t	Pa	90,796	1.0%	908
DP_r	Pa	23,931	1.0%	239
DP_{PPL}	Pa	66,866	1.0%	669
d	M	0.0508	0.05%	0.000025
D	m	0.102	0.25%	0.00026
Y	Dimensionless	0.991	0.30%	0.0030
C_d	Dimensionless	0.602	0.5%	0.003
K_r	Dimensionless	1.163	2.9%	0.017
K_{PPL}	Dimensionless	0.177	1.2%	0.002
ρ	kg/m ³	36.304	0.27%	0.098

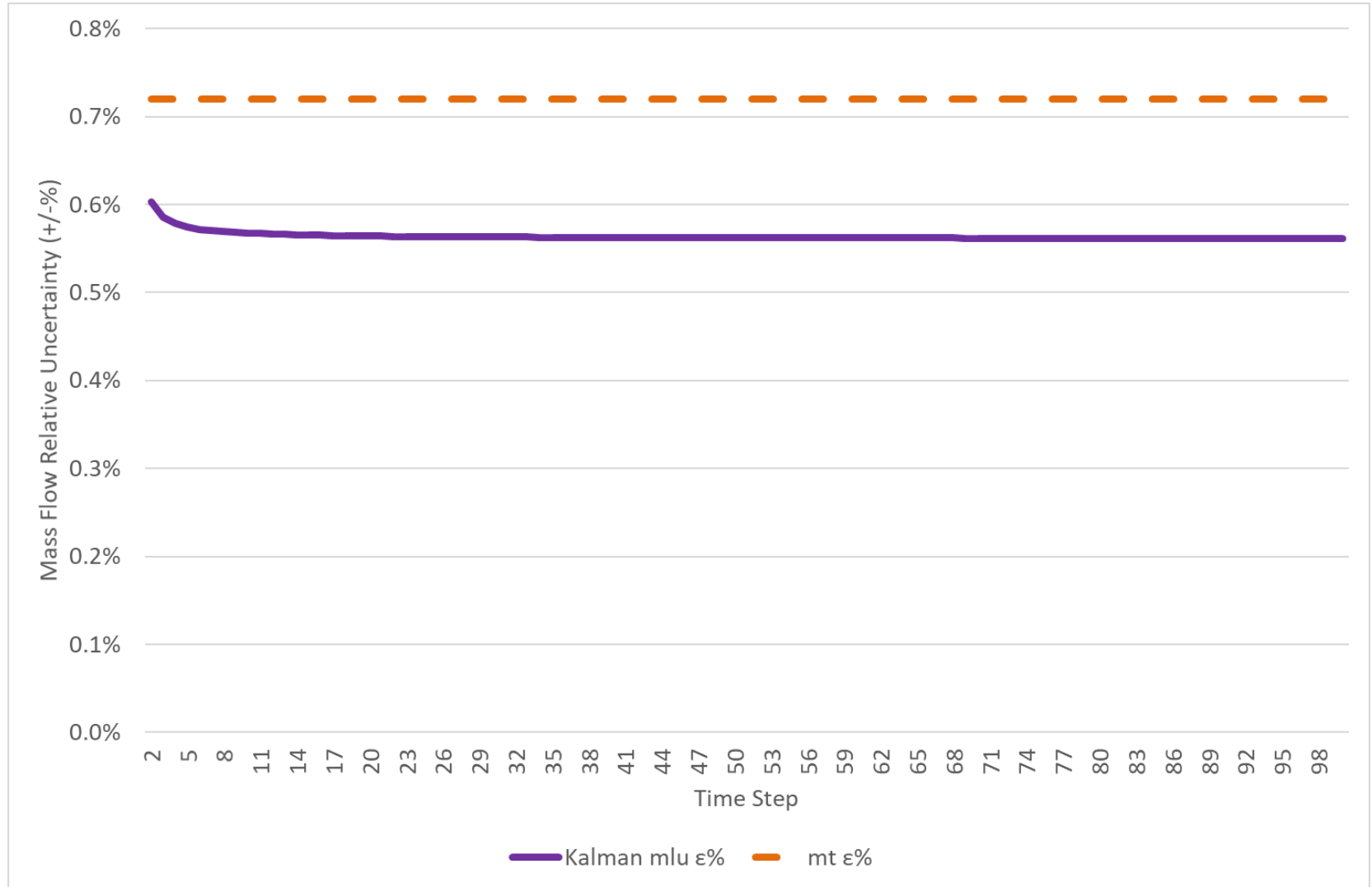
North Sea
FLOW
Measurement Workshop

[illegible]

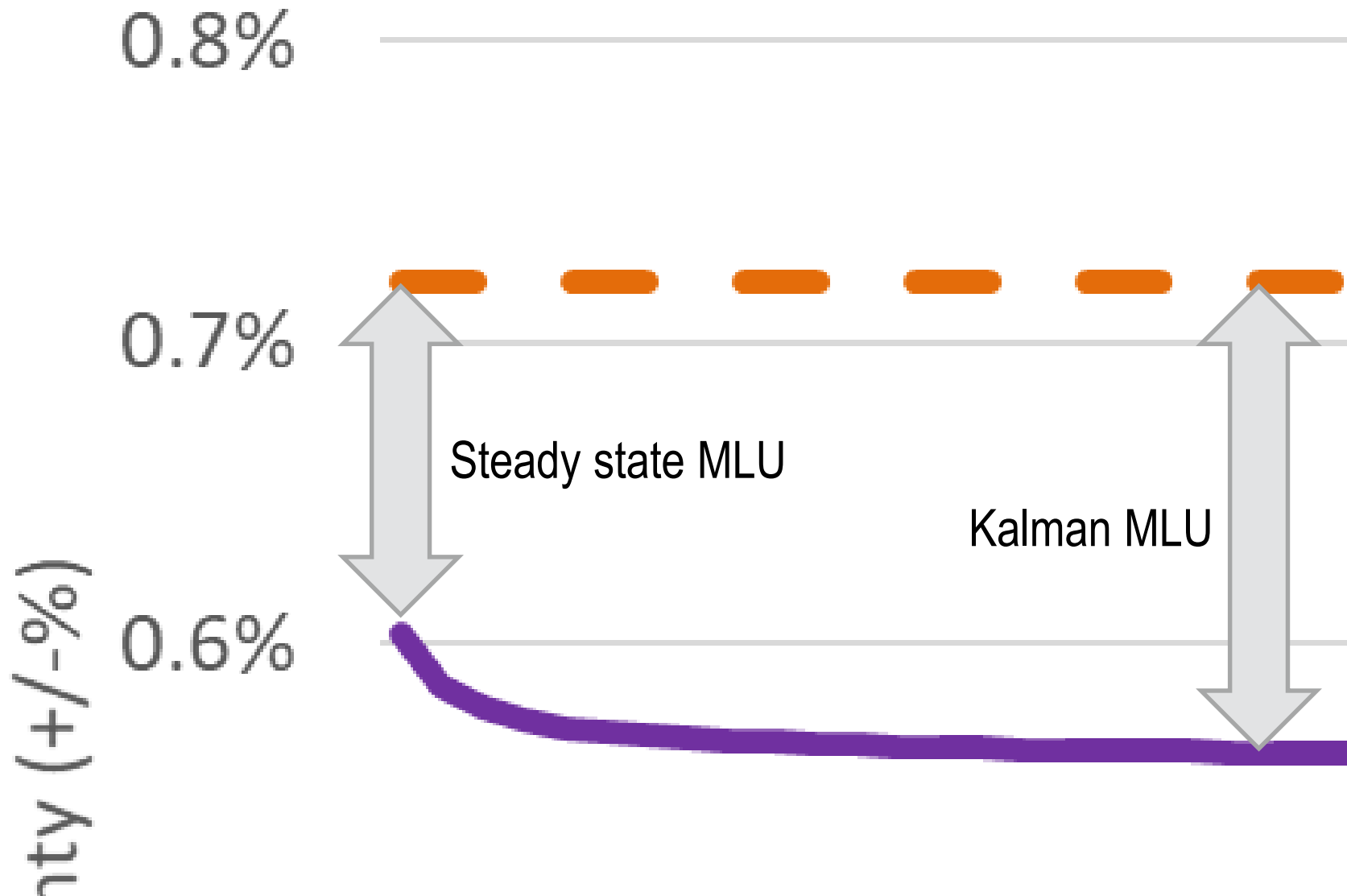
Hypothetical Example – Mass Flow Rate



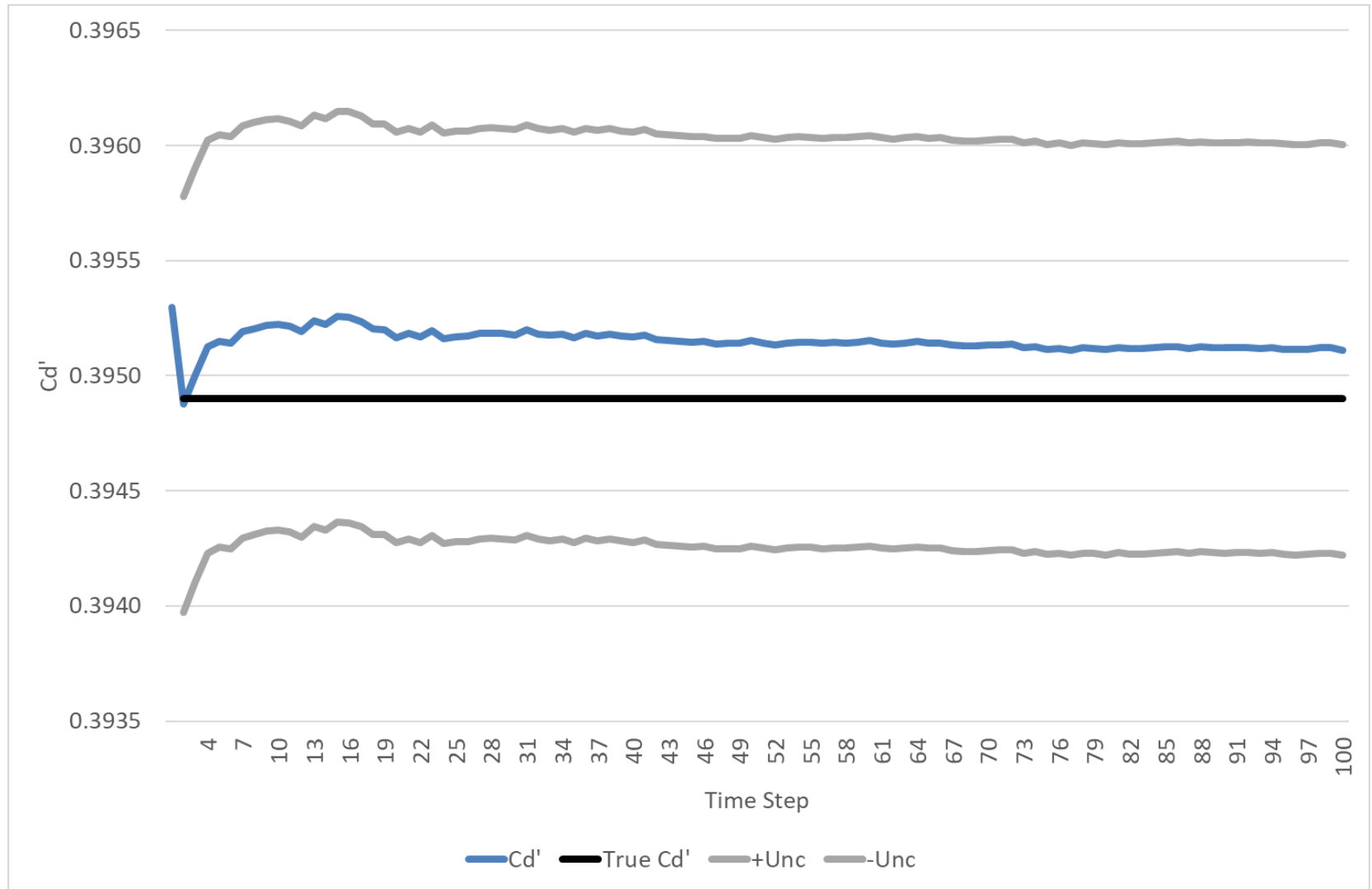
Hypothetical Example – Mass Flow Rate Uncertainty



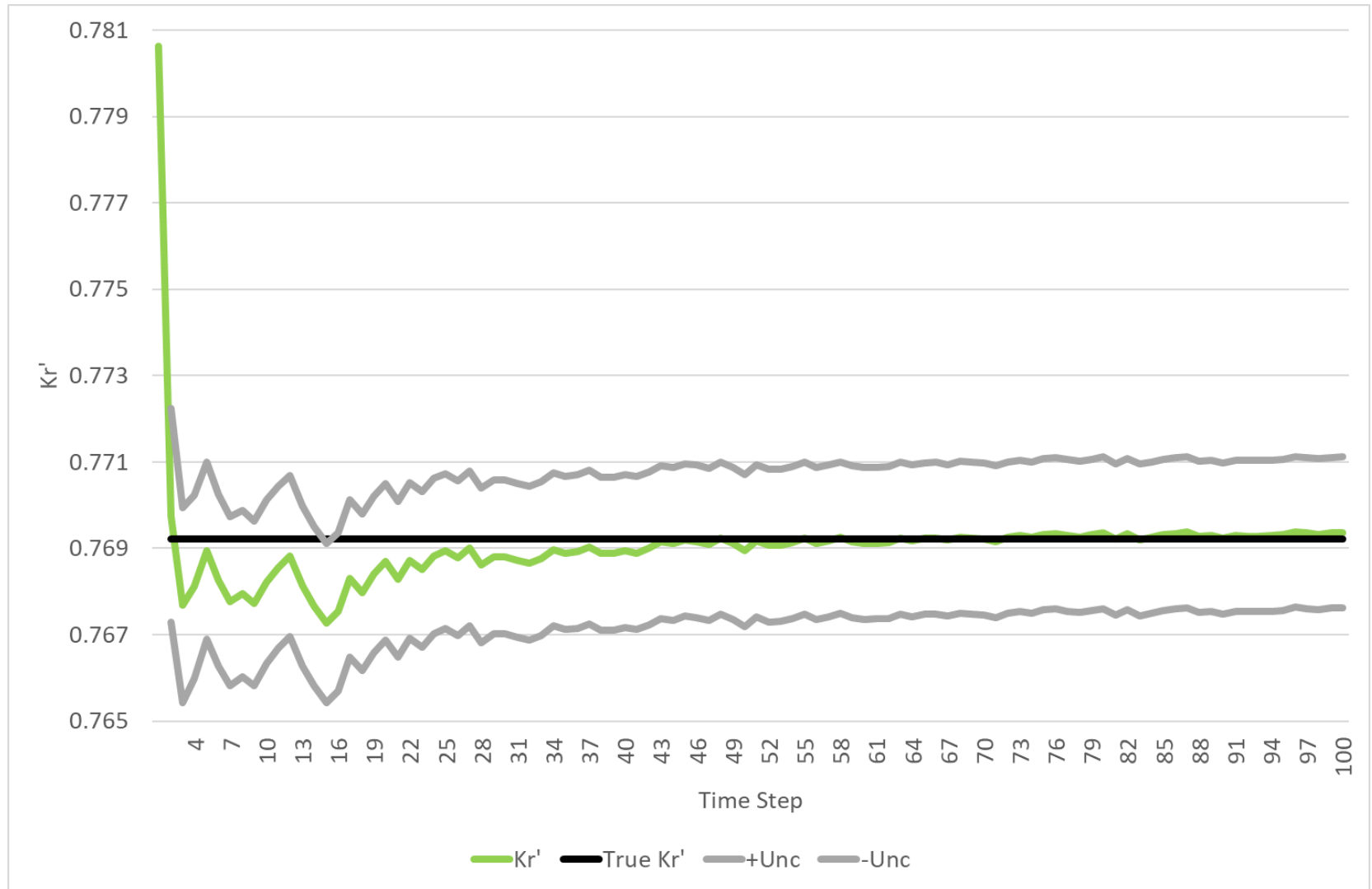
Hypothetical Example – Mass Flow Rate Uncertainty



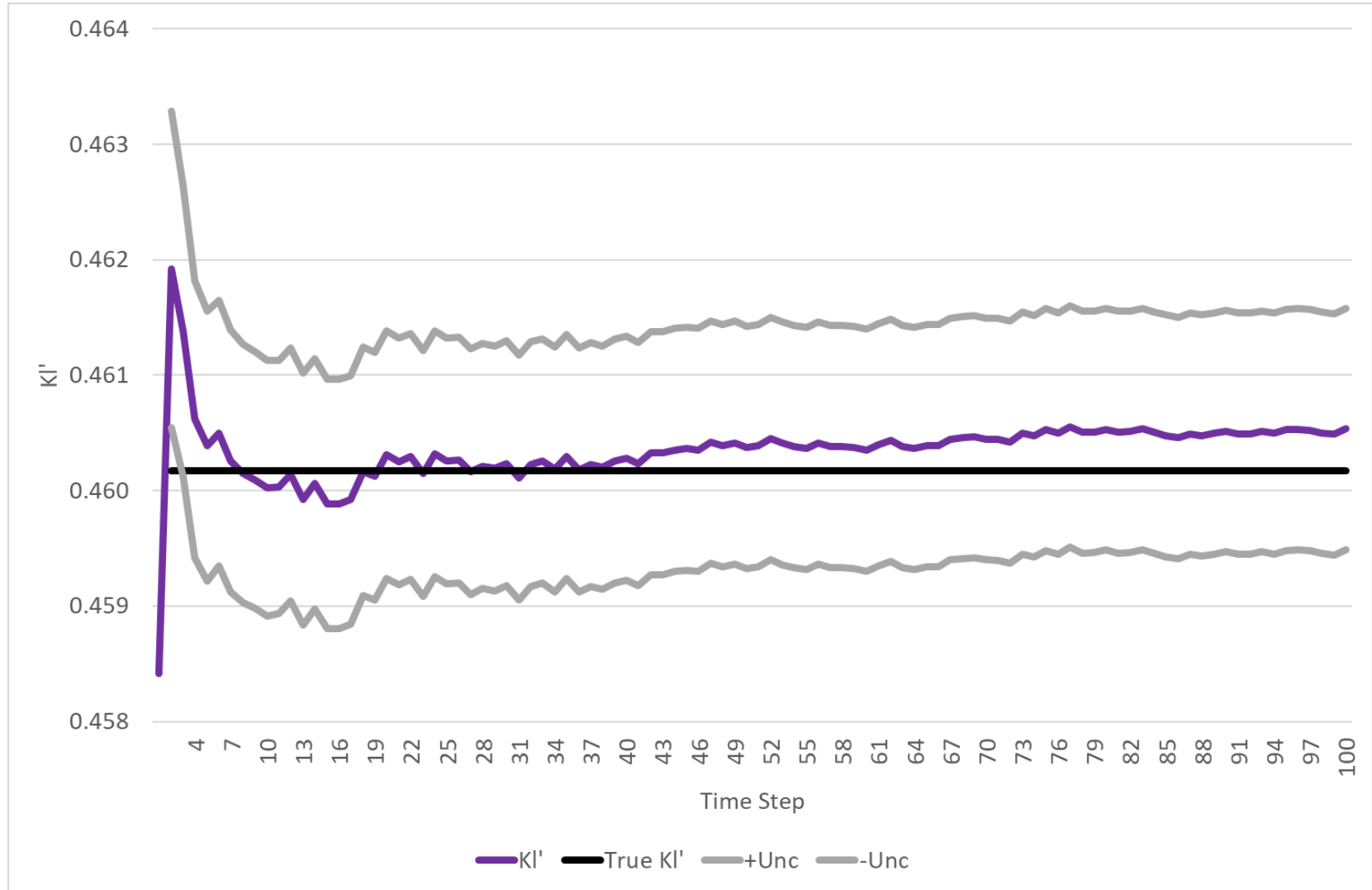
Hypothetical Example – Cd' Evolution



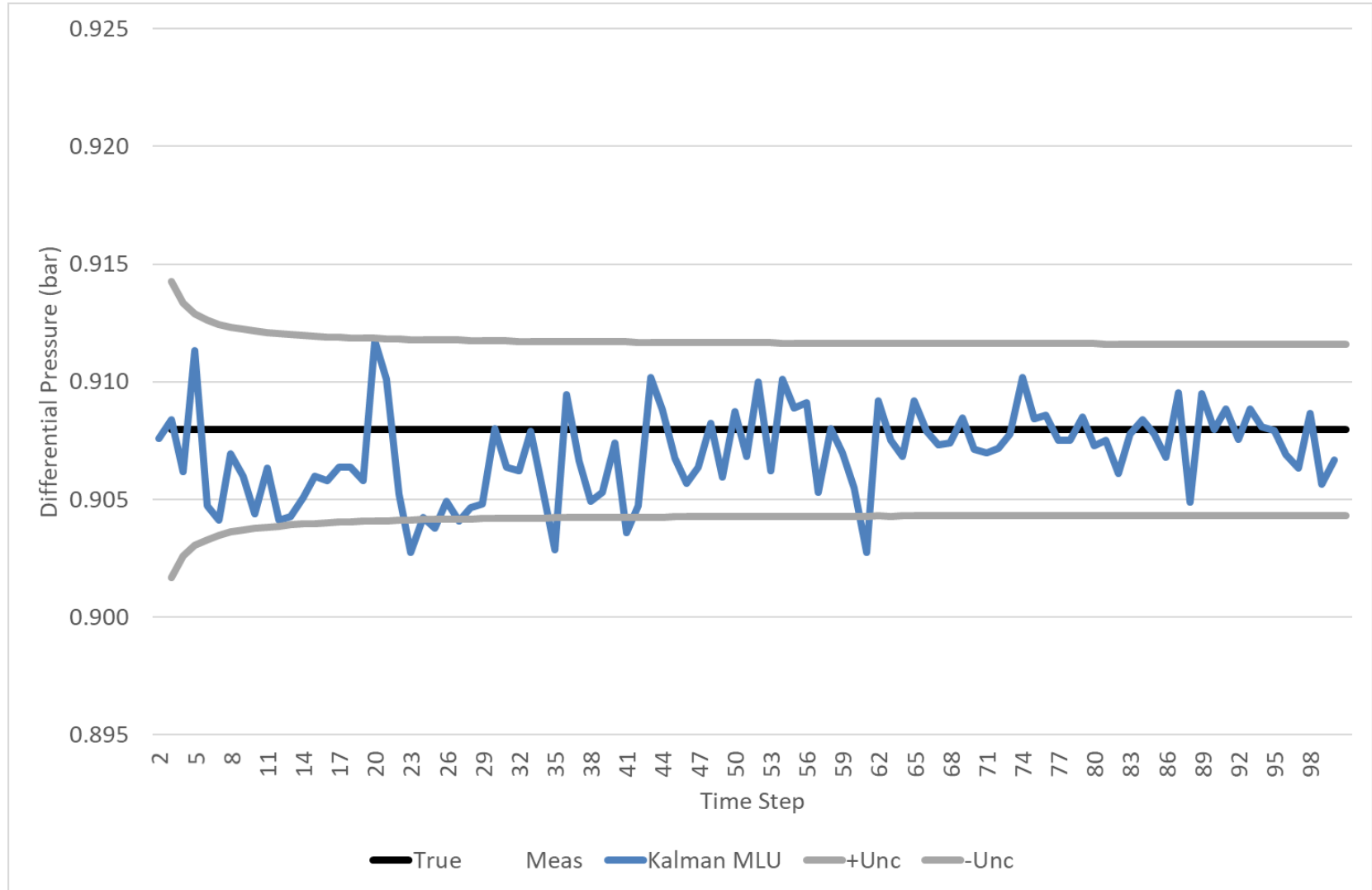
Hypothetical Example – Kr' Evolution



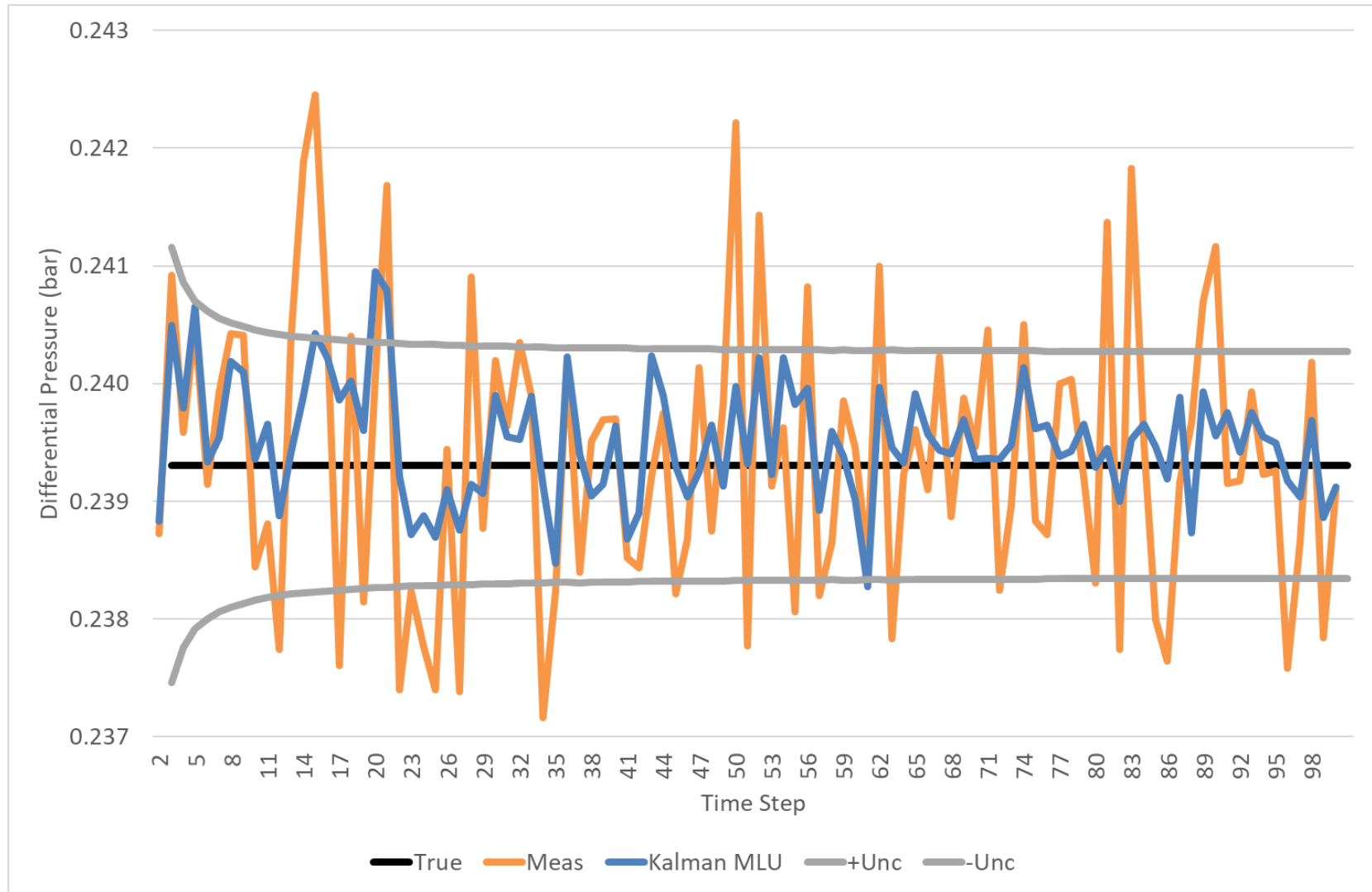
Hypothetical Example – K_{PPL} Evolution



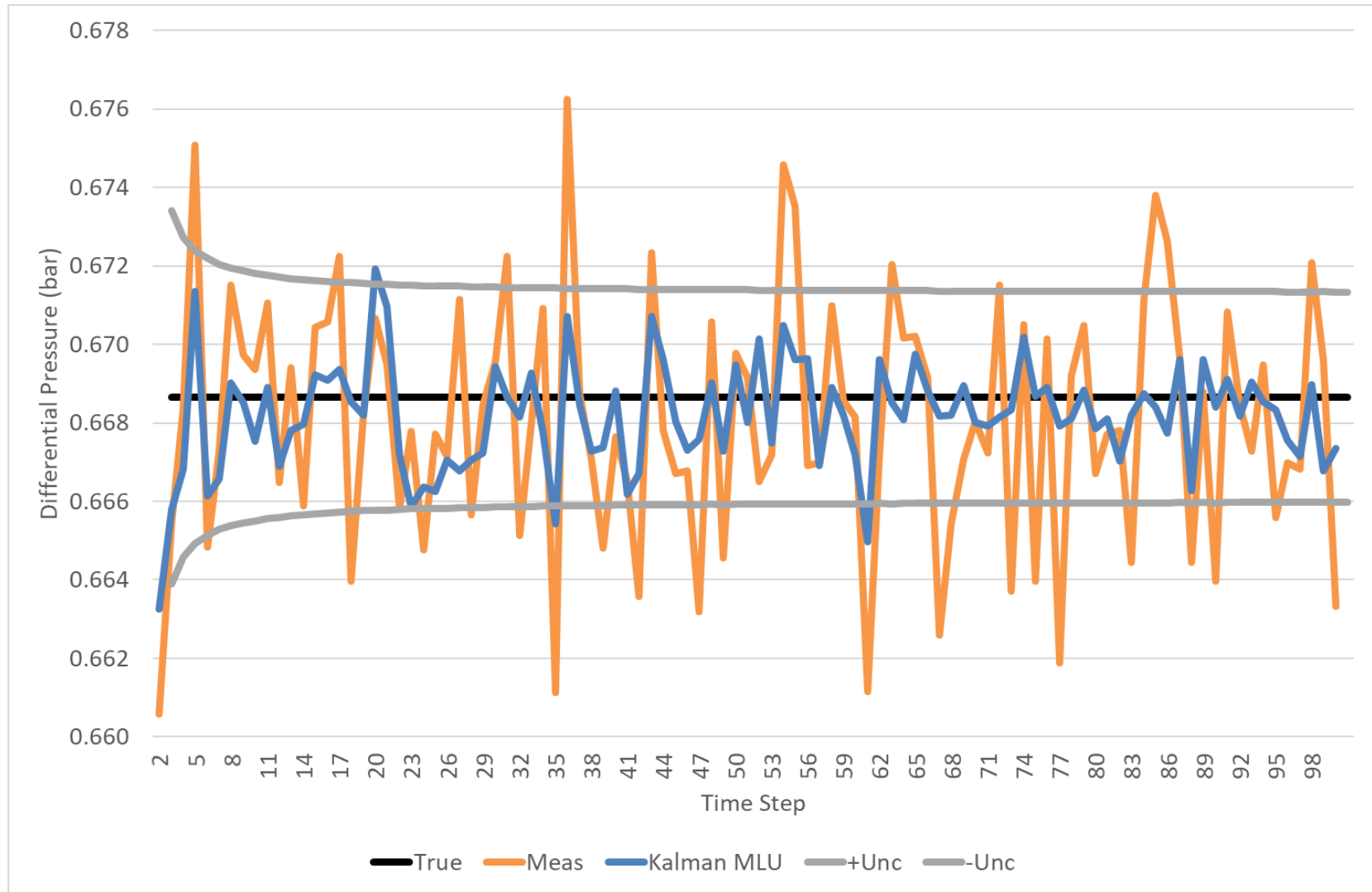
Hypothetical Example – ΔP_t Evolution

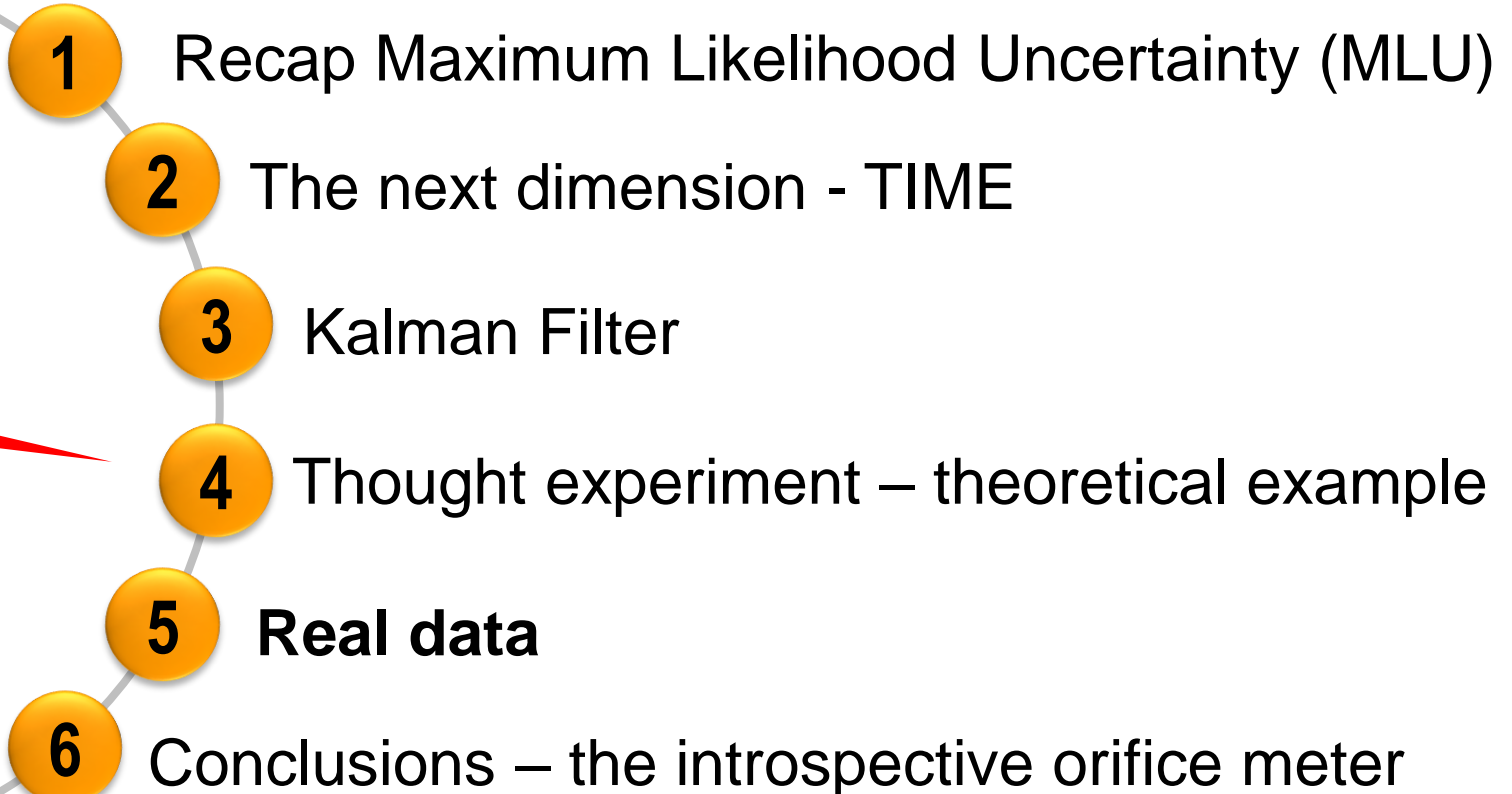


Hypothetical Example – ΔP_r Evolution



Hypothetical Example – ΔP_{pL} Evolution

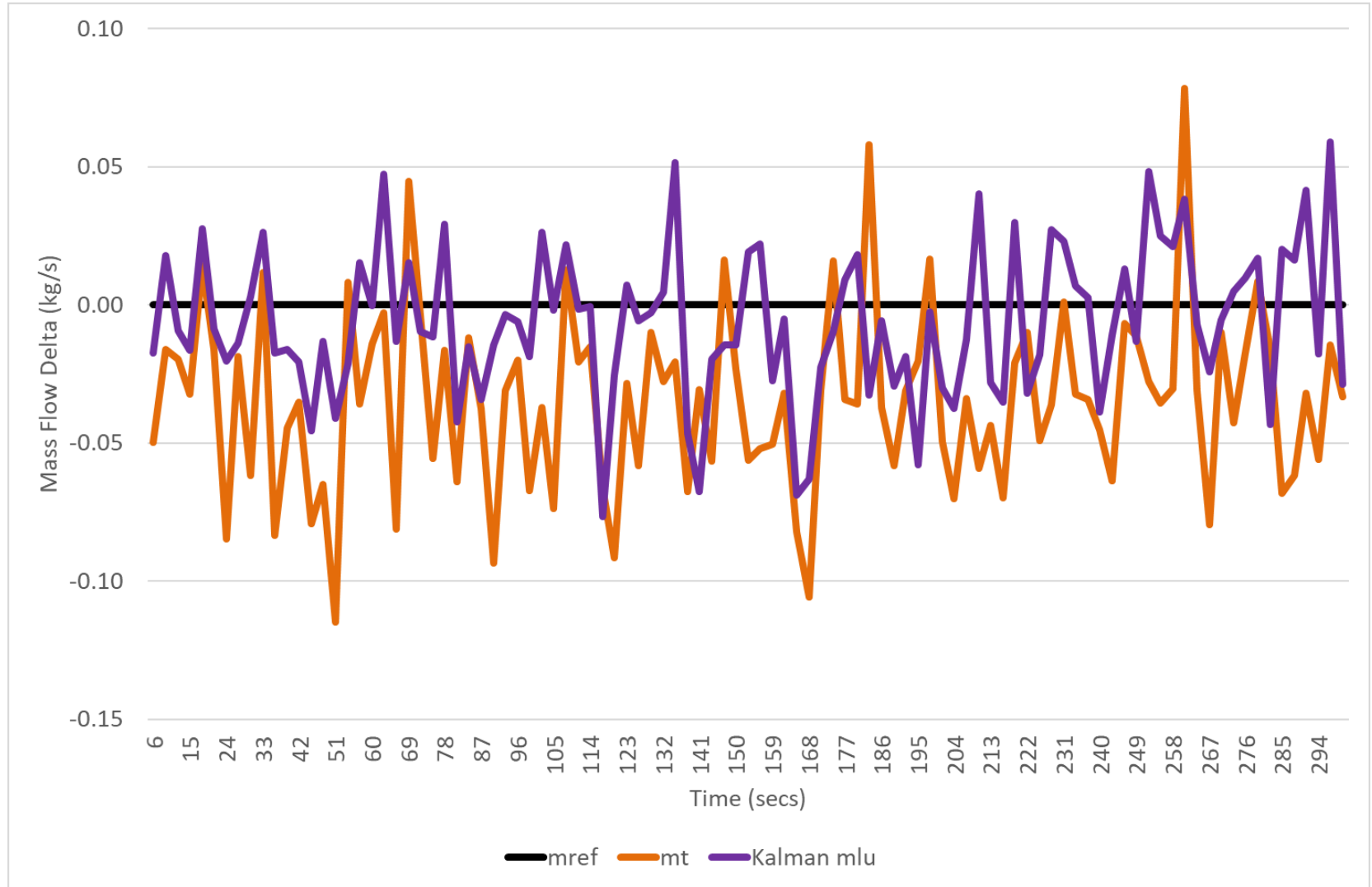


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- 1 Recap Maximum Likelihood Uncertainty (MLU)
 - 2 The next dimension - TIME
 - 3 Kalman Filter
 - 4 Thought experiment – theoretical example
 - 5 **Real data**
 - 6 Conclusions – the introspective orifice meter

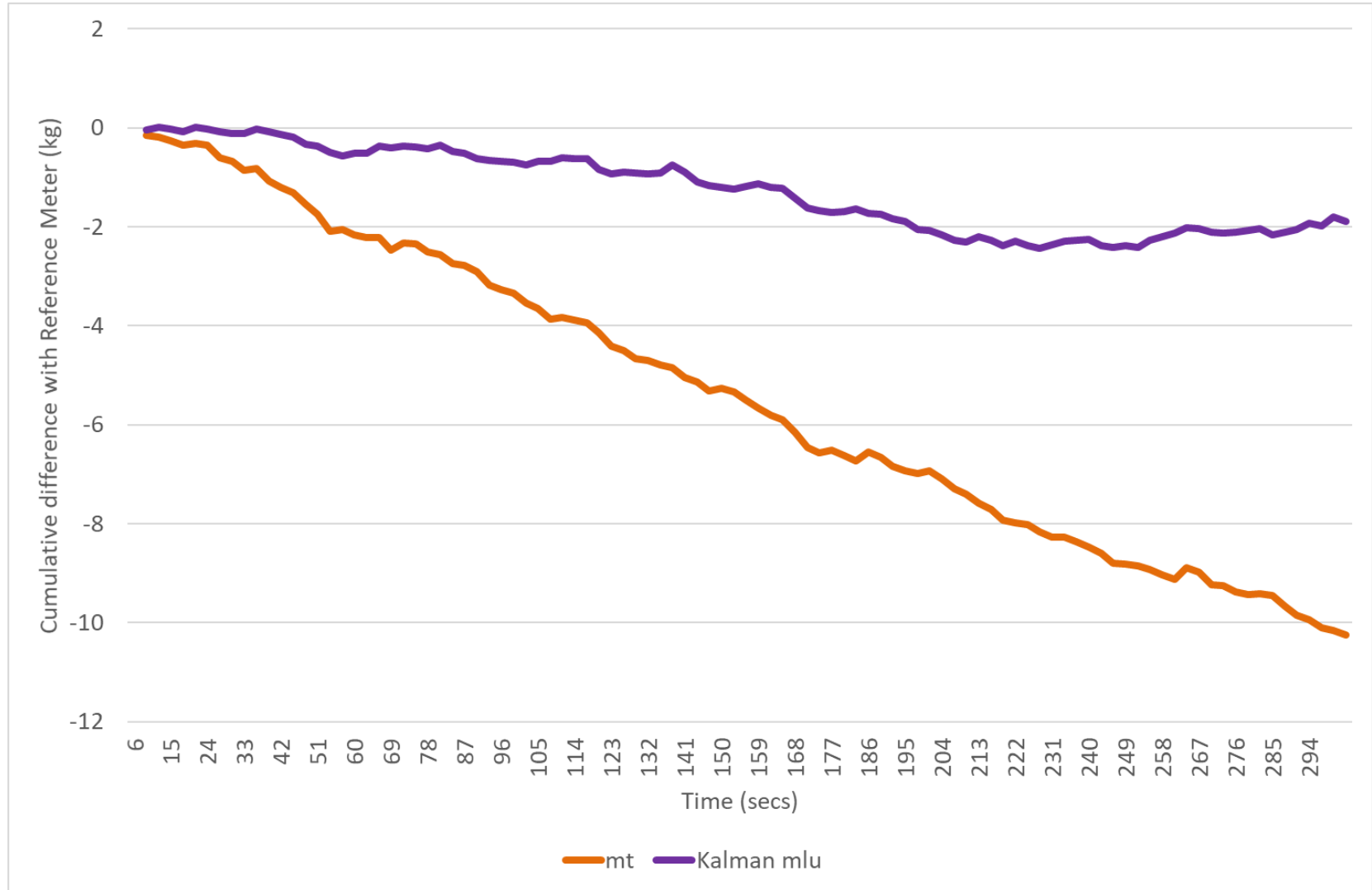
Real Data – 6", 0.6 Beta Orifice DP Meter



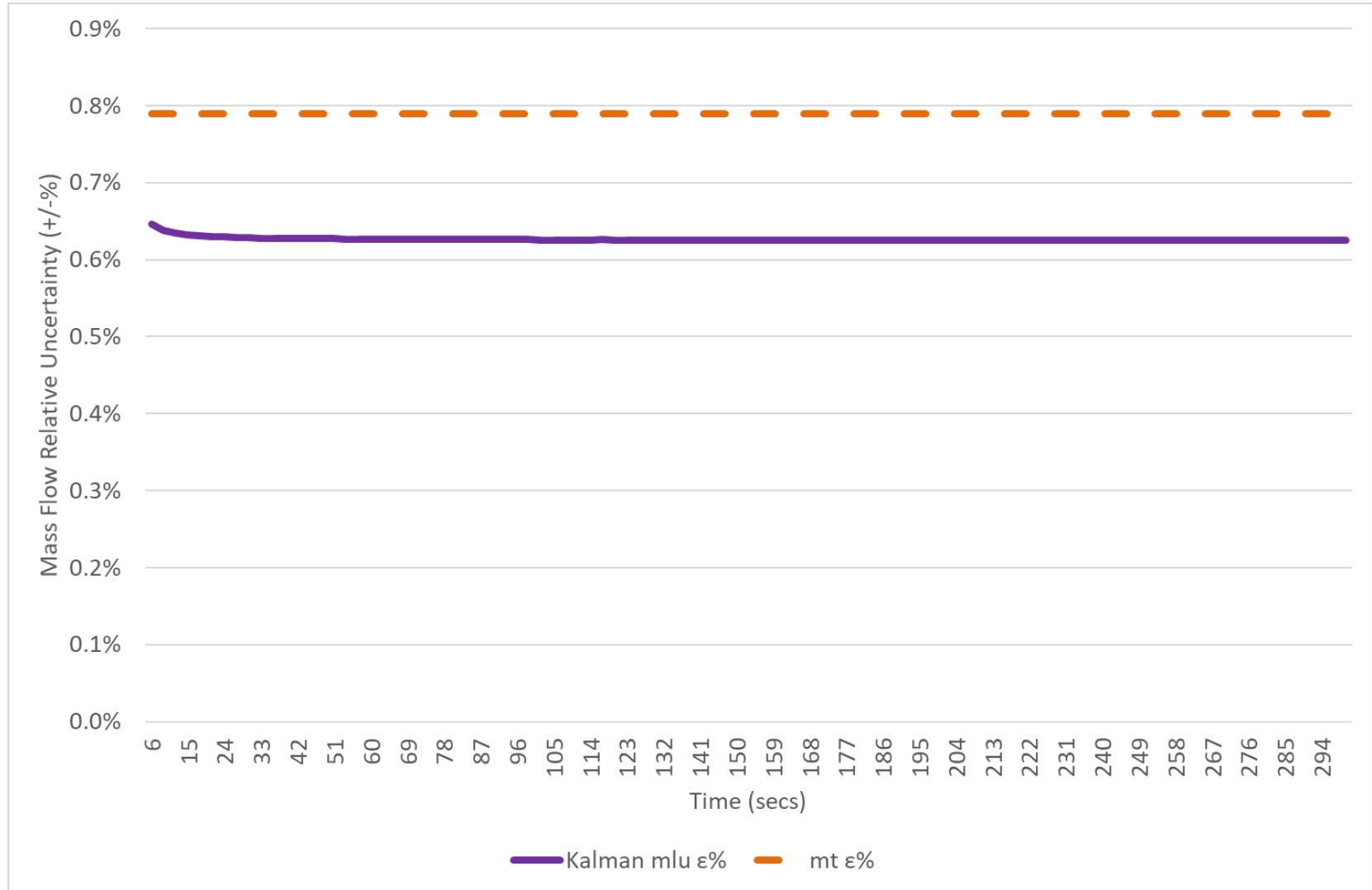
Real Data – Mass Flow Rate



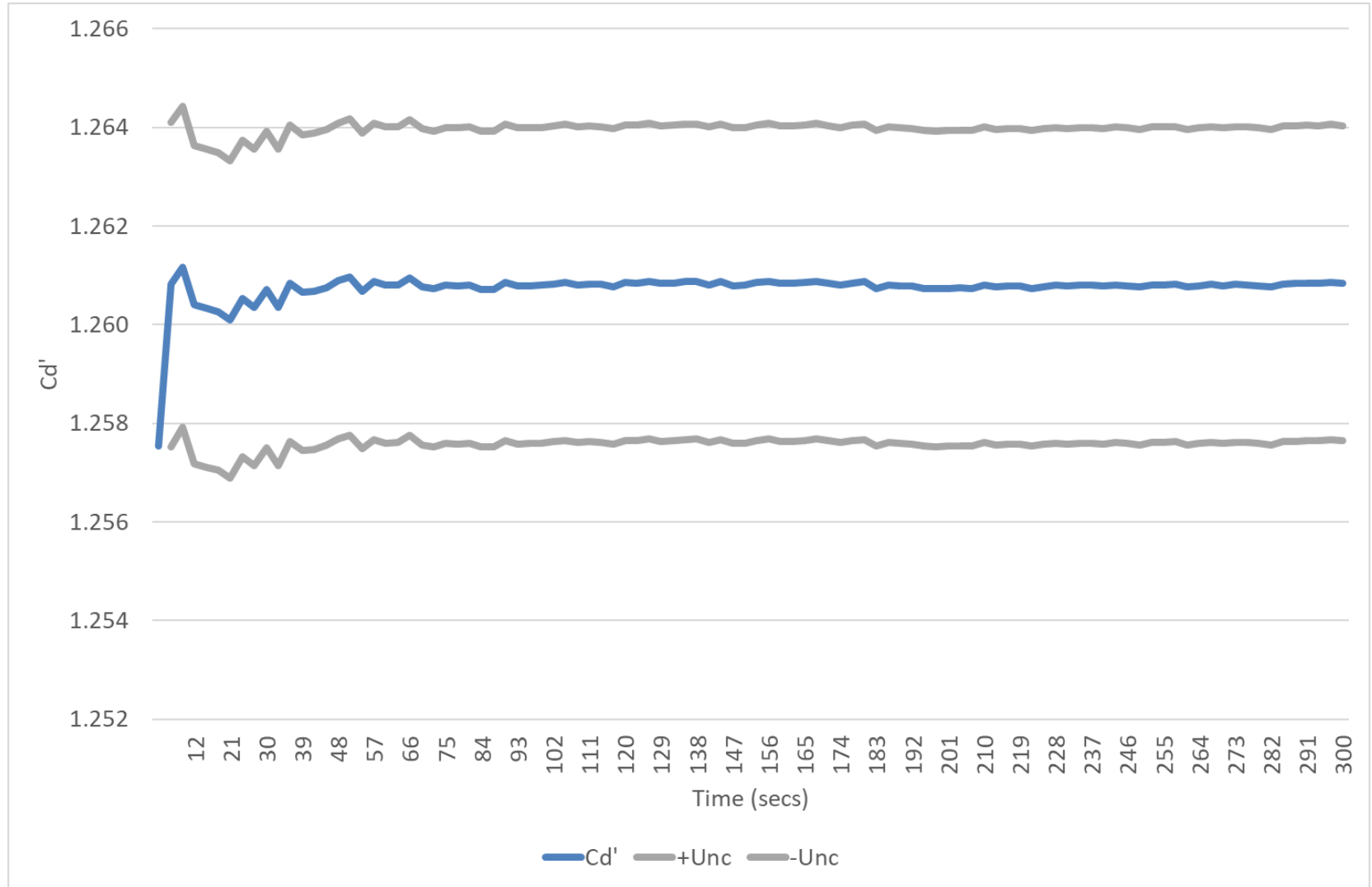
Real Data – Cumulative Mass Difference



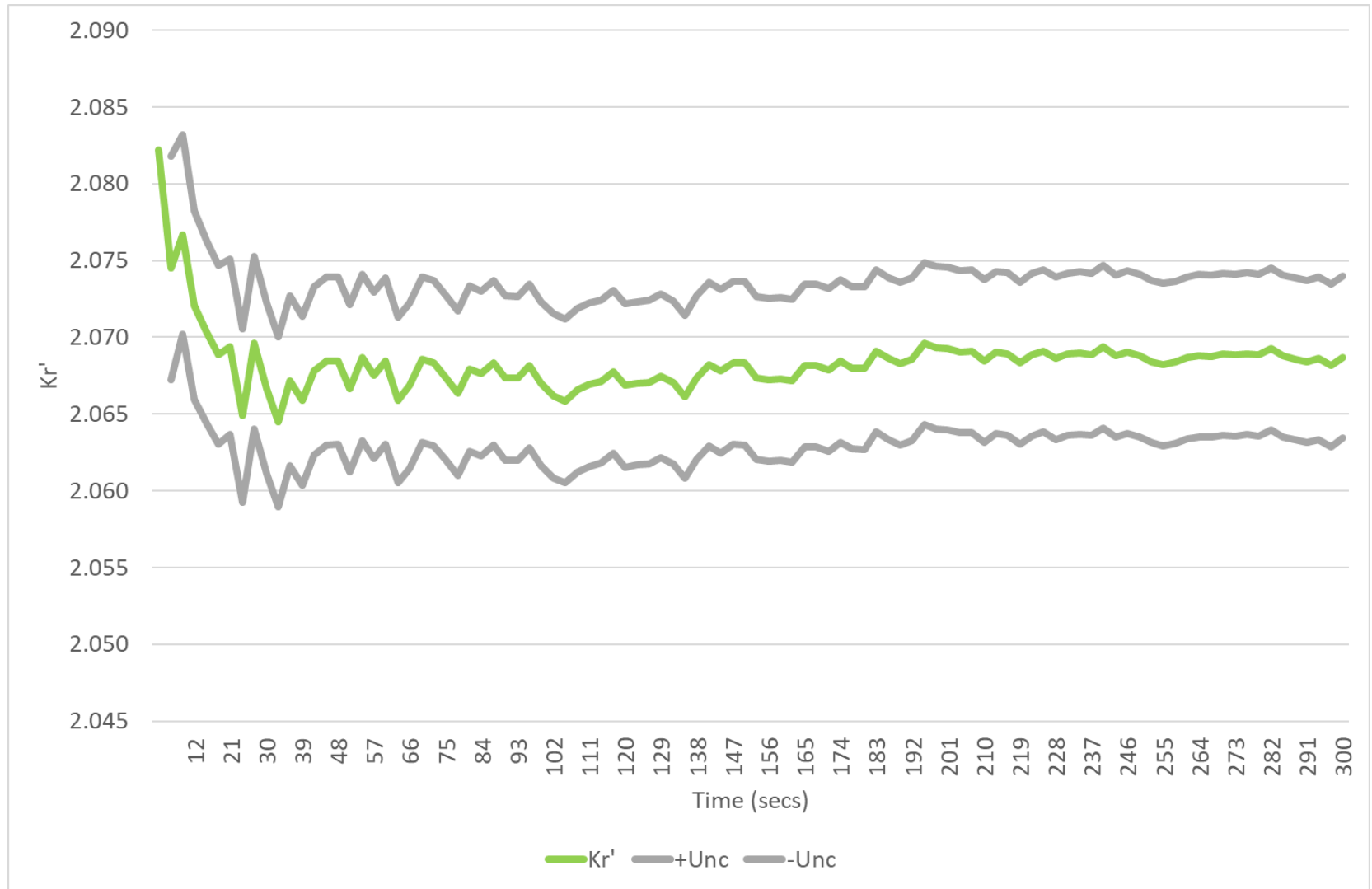
Real Data – Mass Flow Rate Uncertainty



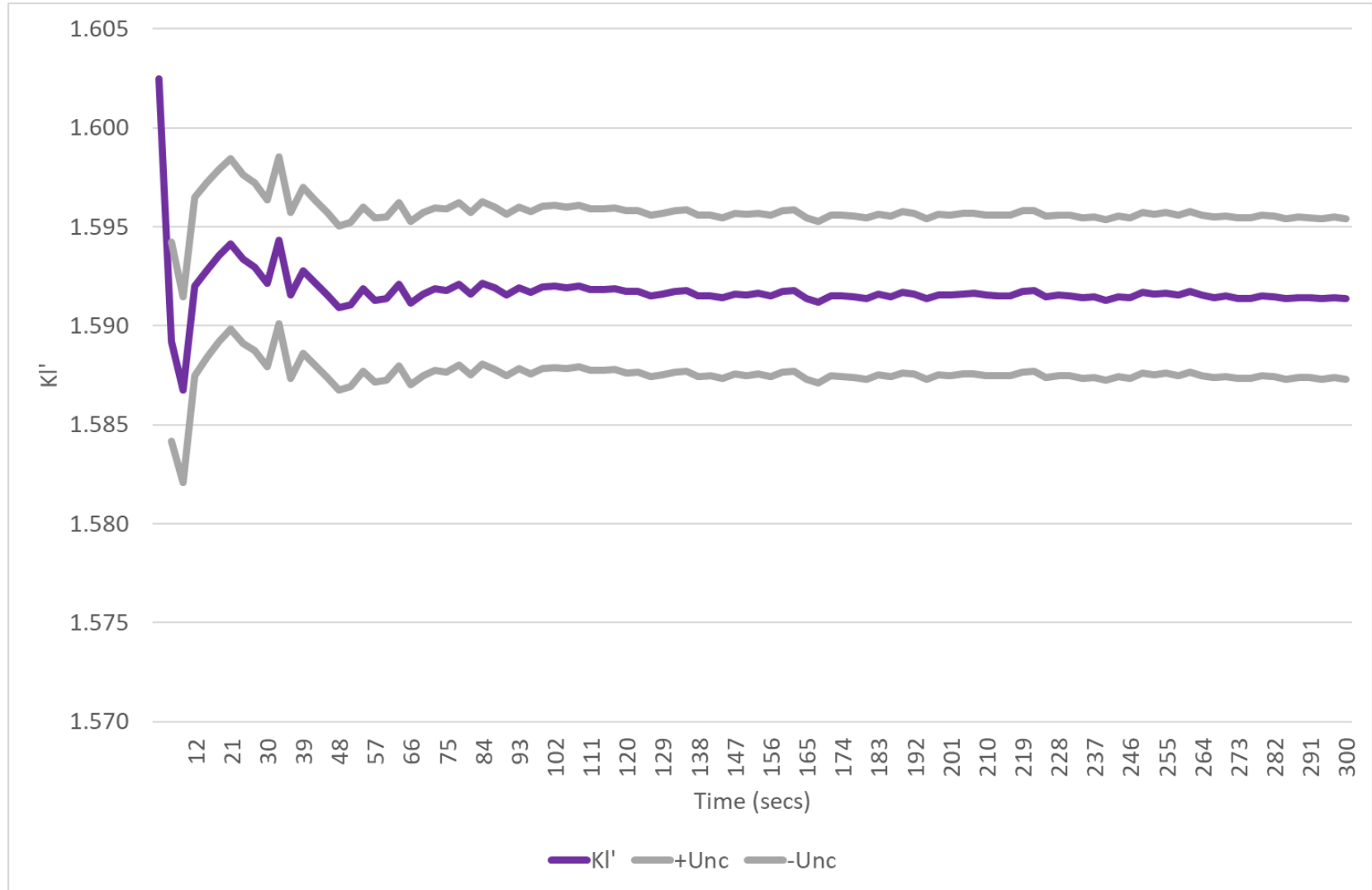
Hypothetical Example – Cd' Evolution

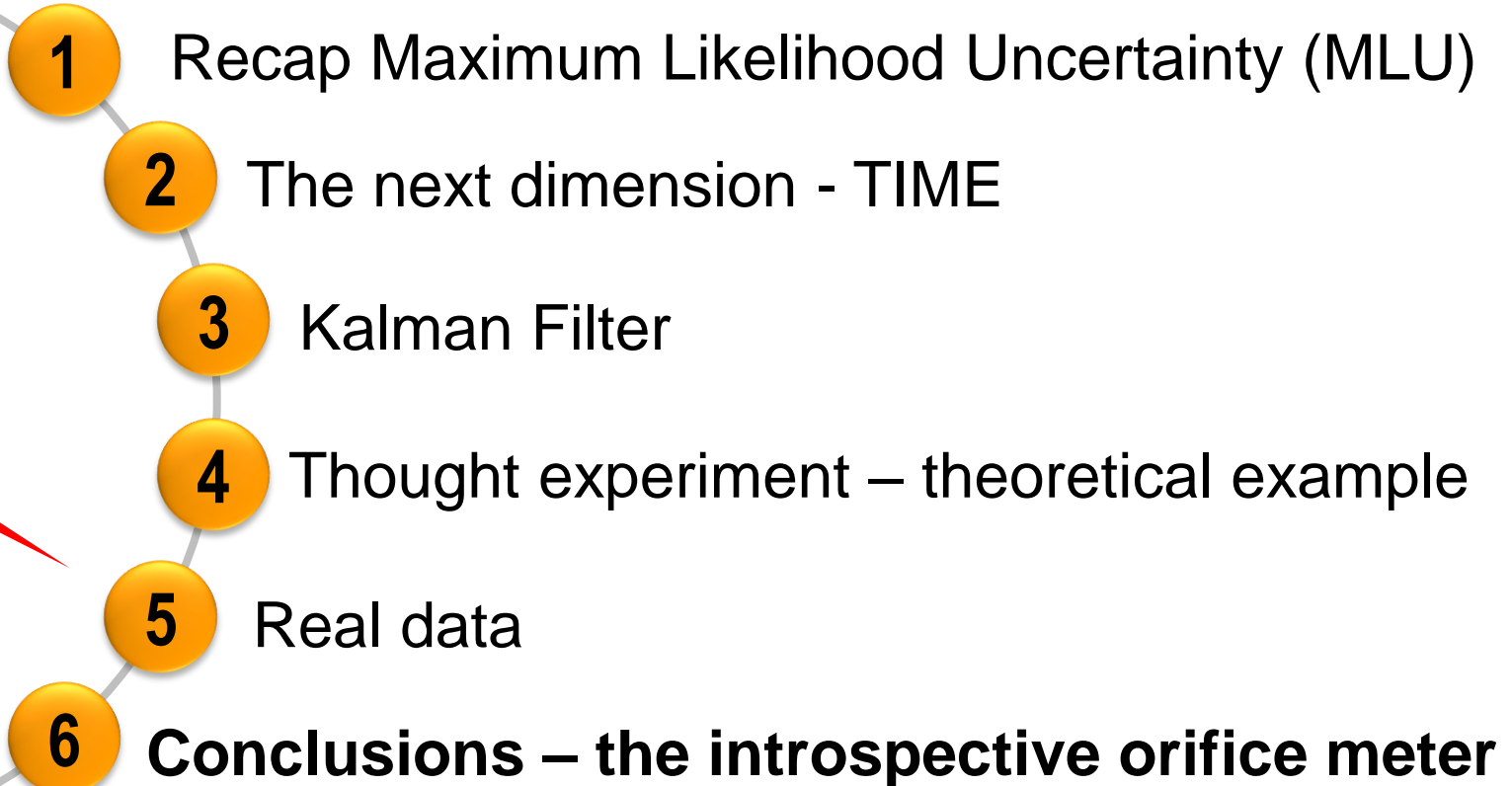


Hypothetical Example – Kr' Evolution



Hypothetical Example – K_{PPL} Evolution



- 
- 1 Recap Maximum Likelihood Uncertainty (MLU)
 - 2 The next dimension - TIME
 - 3 Kalman Filter
 - 4 Thought experiment – theoretical example
 - 5 Real data
 - 6 **Conclusions – the introspective orifice meter**

The Introspective Orifice Meter

DP meter, well established



Data: DP signal variation in time

Kalman MLU



Utilises the time dimension

Demonstrated using theoretical and real data



The introspective orifice meter

What Consider the two patterns HTH and HTT

Which of the following is true:

A

The average number of tosses until HTH is **larger** than the average number of tosses until HTT

B

The average number of tosses until HTH is the **same** as the average number of tosses until HTT

C

The average number of tosses until HTH is **smaller** than the average number of tosses until HTT

Heads and tails sequences

H T H

H T T H T H

H T T

H T H T T

To raise new questions, new possibilities, to regard old problems from a new angle, requires creative imagination and marks real advance in science.

