



# Transforming Data Across Environments Despite Structural Non-Identifiability

ACC 2019

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# Overview

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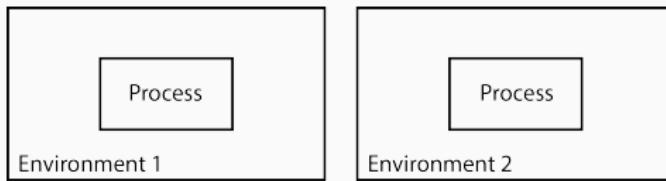
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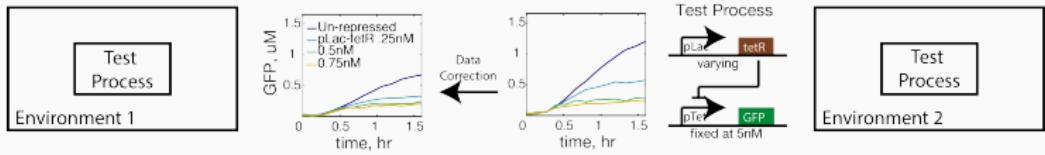
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- Examples of problems:
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  - characterize subsystem behavior → predict system behavior

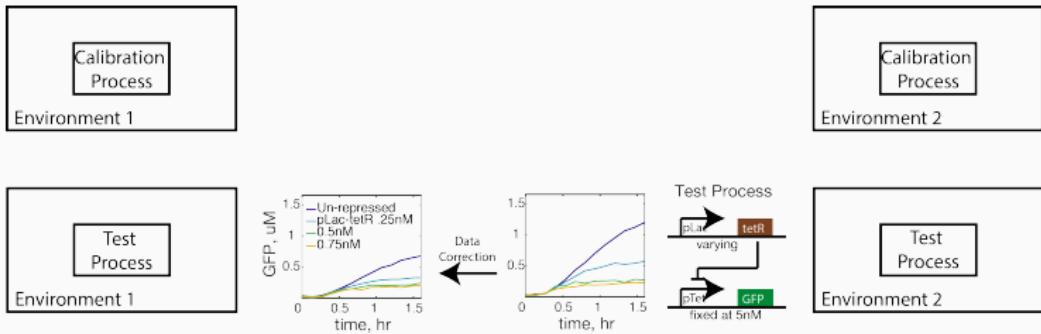
# Transforming Process Behavior Across Environments



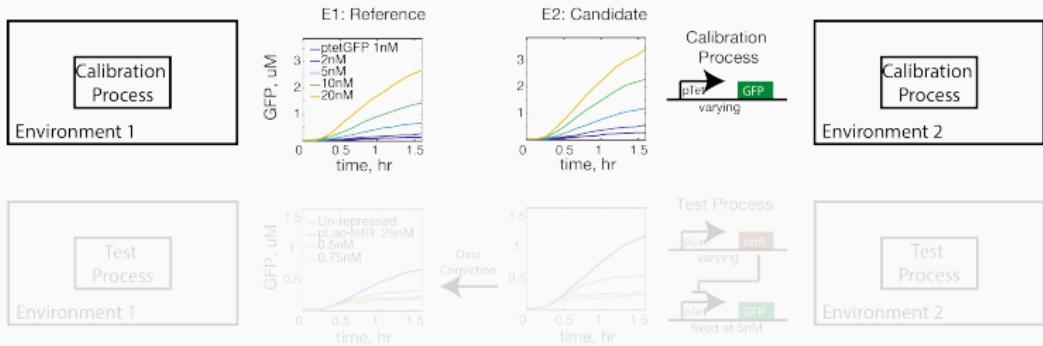
# The Data Correction Problem



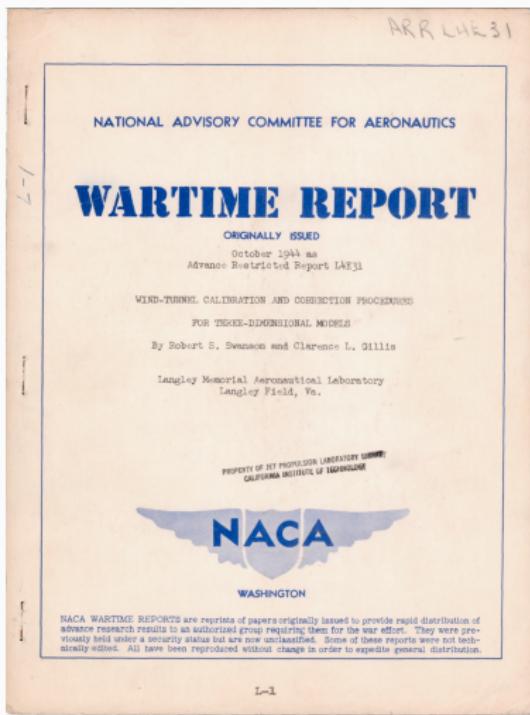
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# Solution: The Calibration-Correction Method

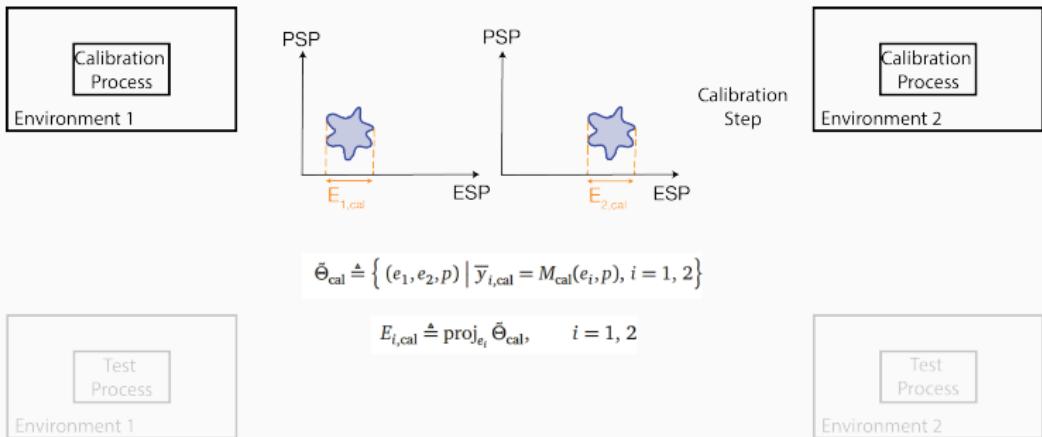


## Solution: The Calibration-Correction Method

The usual practice of predicting the flying qualities of airplanes from wind-tunnel tests of relatively small-scale models makes it imperative that the model test results be corrected to free-air conditions. In addition, the large number of wind tunnels in use makes it desirable that a more or less standard calibration and correction procedure be adopted in order to make data from different tunnels as nearly comparable as possible. Not much com-

# The Calibration-Correction Method: Calibration Step

Estimate environment specific parameters

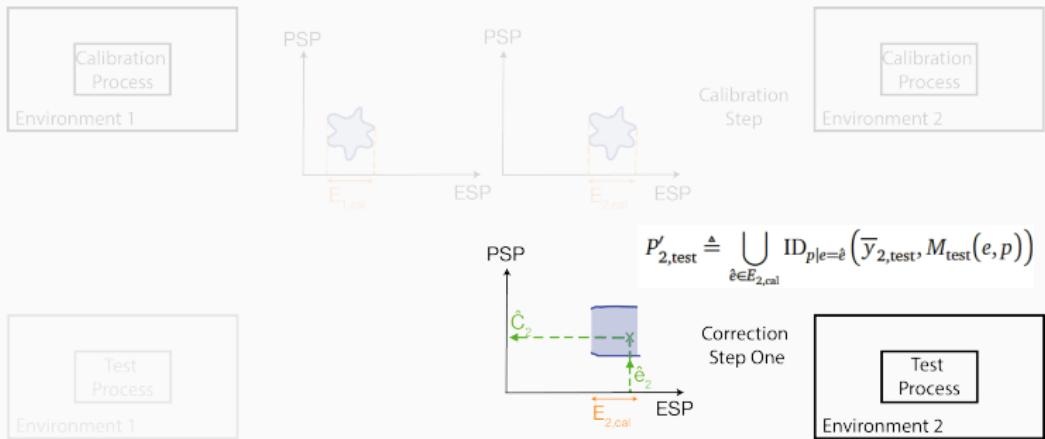


$$\tilde{\Theta}_{\text{cal}} \triangleq \left\{ (e_1, e_2, p) \mid \bar{y}_{i,\text{cal}} = M_{\text{cal}}(e_i, p), i = 1, 2 \right\}$$

$$E_{i,\text{cal}} \triangleq \text{proj}_{e_i} \tilde{\Theta}_{\text{cal}}, \quad i = 1, 2$$

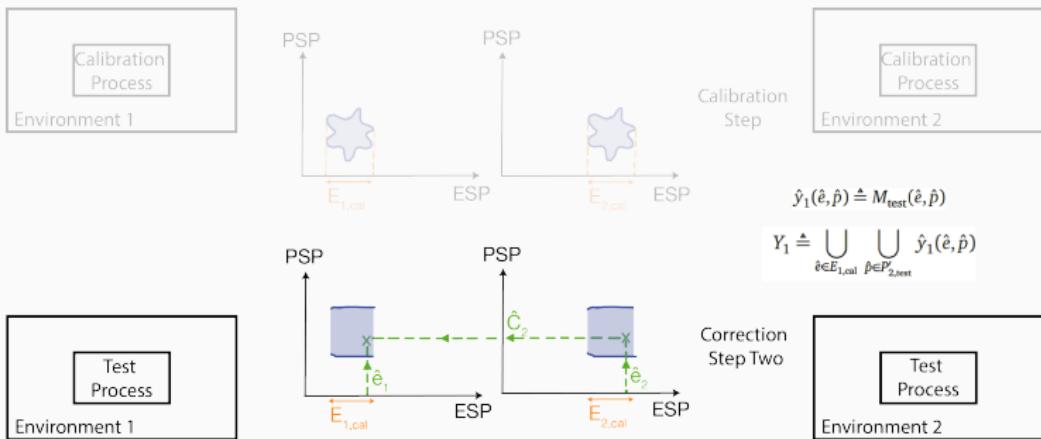
# The Calibration-Correction Method: Correction Step 1

Estimate test process parameters in Extract 2



# The Calibration-Correction Method: Correction Step 2

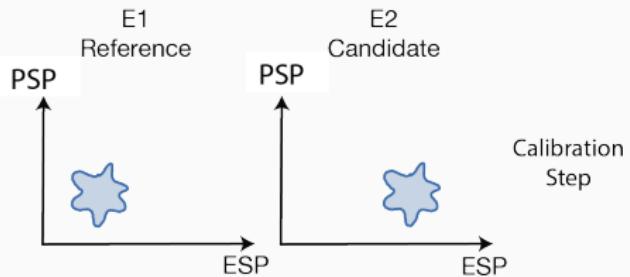
Predict test process behavior in Extract 1



# Parameter Consistency

## Theorem

Consider the data correction problem in the model universe. Furthermore, consider the calibration-correction method, and the sets  $\tilde{\Theta}_{\text{cal}}$ ,  $E_{1,\text{cal}}$ ,  $E_{2,\text{cal}}$  and  $C'_{2,\text{test}}$ . Define  $\Theta_{i,\text{test}} \triangleq \text{ID}_\theta(\bar{y}_{i,\text{test}}, \bar{M}_{\text{test}}(\theta))$  for  $i = 1, 2$ . Then, the conditions,



$$\tilde{\Theta}_{\text{cal}} \neq \emptyset$$

$$E_{2,\text{cal}} \subseteq \text{proj}_e \Theta_{2,\text{test}}$$

$$E_{1,\text{cal}} \times C'_{2,\text{test}} \subseteq \Theta_{1,\text{test}},$$

are necessary and sufficient for the calibration-correction method to solve the data correction problem.

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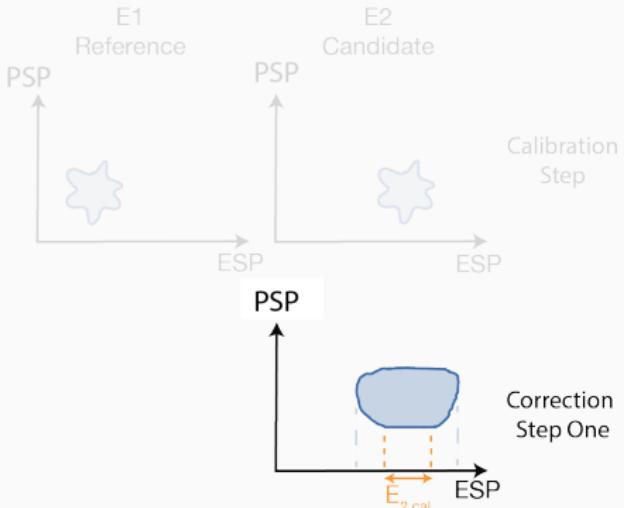
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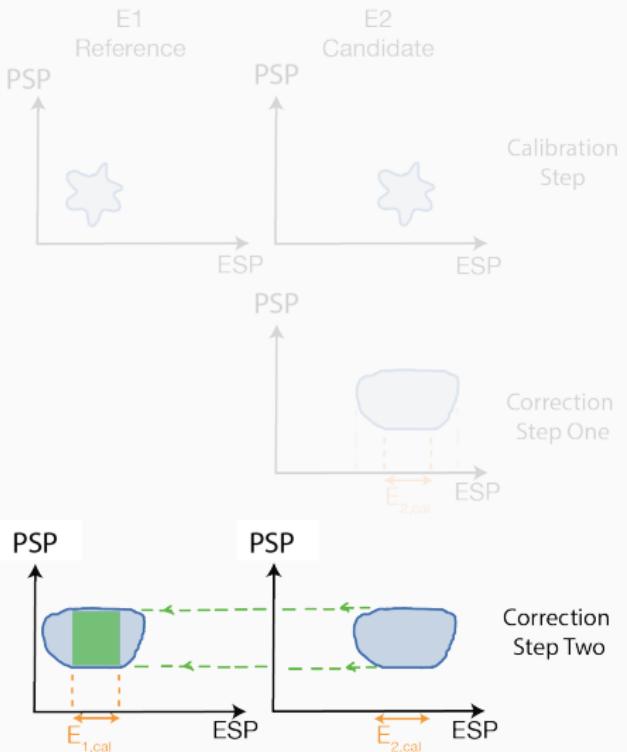
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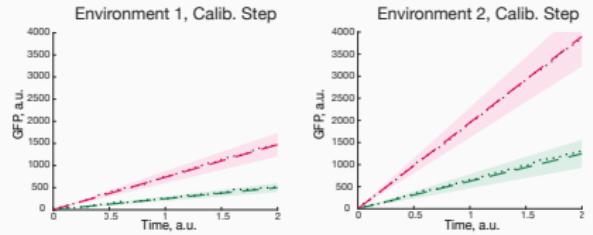
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# Consider the Test = Calibration Process Case

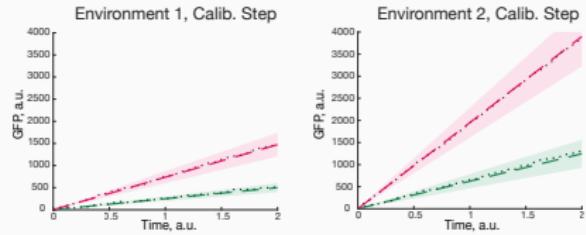


"Test = Calibration" Case



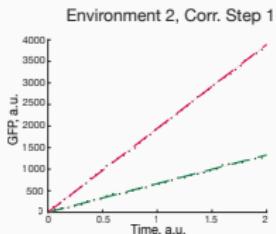
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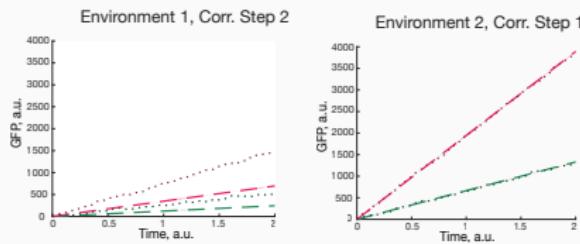
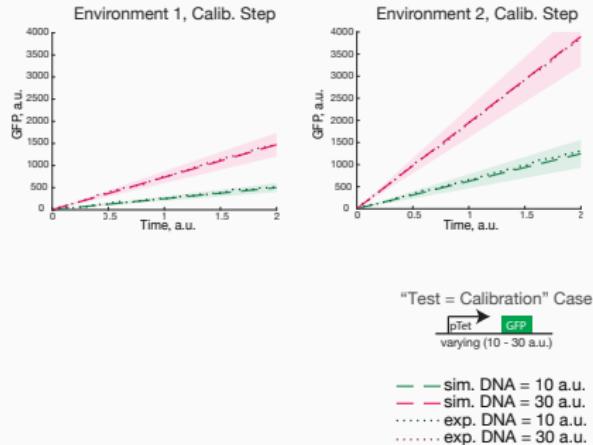


"Test = Calibration" Case  
pTet → GFP  
varying (10 - 30 a.u.)

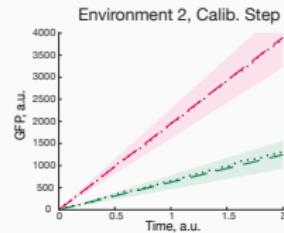
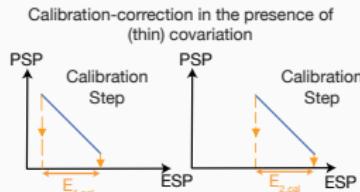
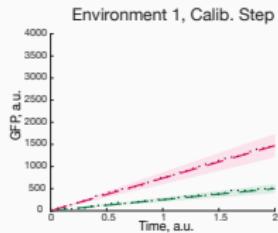
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# Correcting the Calibration Data Fails

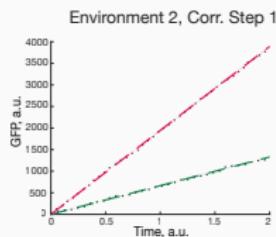
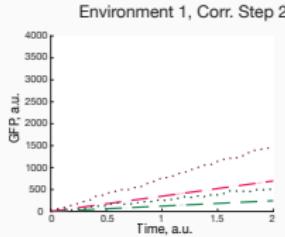


# Parameter Covariation Could be a Problem

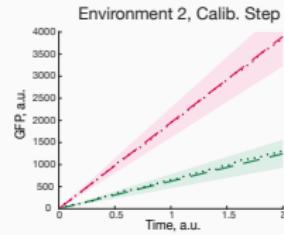
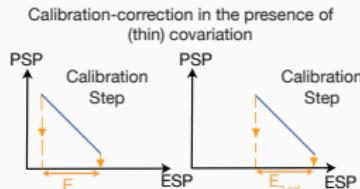
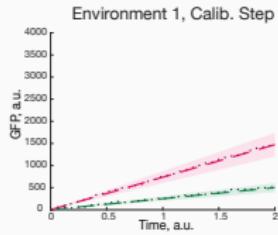


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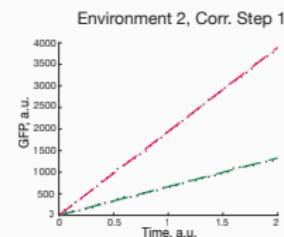
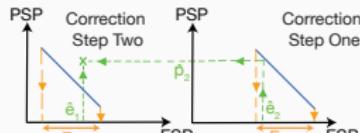
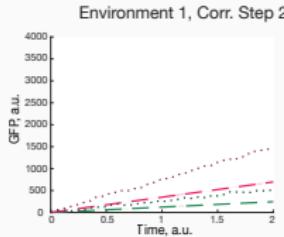


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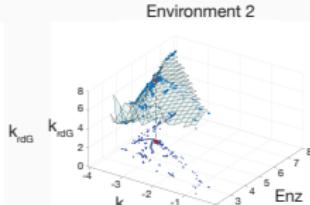
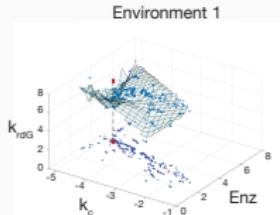
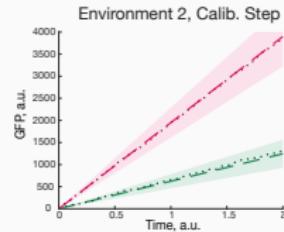
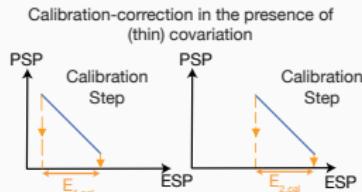
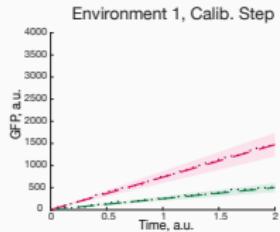


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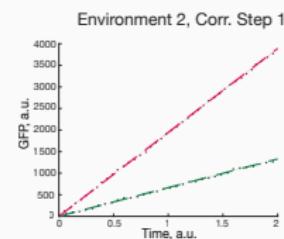
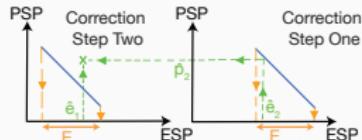
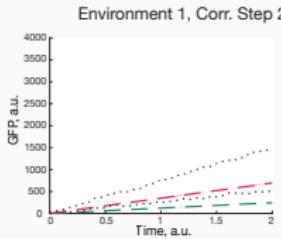


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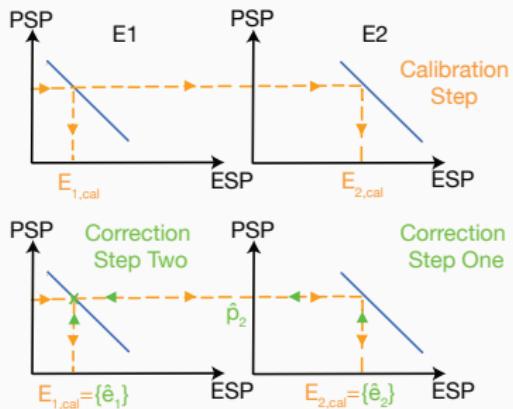
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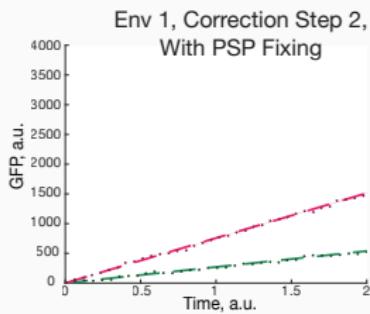
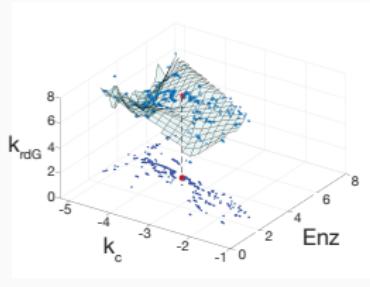
# PSP Fixing Helps

Calibration Correction with PSP fixing

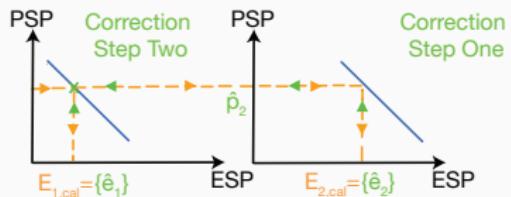
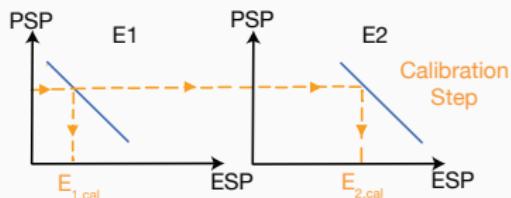


# PSP Fixing Helps

Env. 1, With PSP Fixing



Calibration Correction with PSP fixing



# Formally...

## Proposition

Consider the 'Test = Calib' case of the DCP. Assume the conditions

$$\begin{aligned}\tilde{\Theta}_{\text{cal}} &\neq \emptyset, \\ C'_{2,\text{cal}} &\subseteq \text{proj}_c \Theta_{1,\text{cal}}, \\ E_{i,\text{cal}} &\subseteq \text{proj}_e \Theta_{i,\text{cal}}, \quad i = 1, 2,\end{aligned}$$

hold, but the set

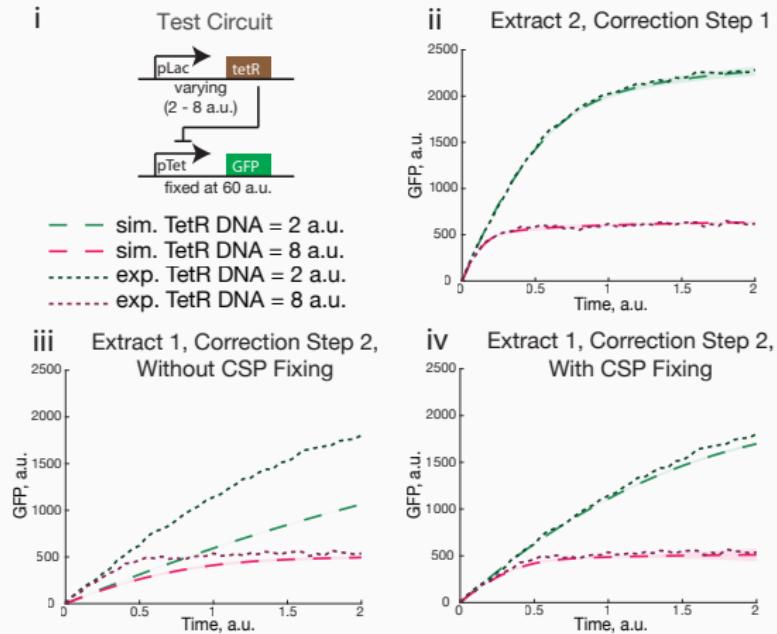
$$\Theta'_{1,\text{cal}} \triangleq \Theta_{1,\text{cal}} \cap (E_{1,\text{cal}} \times \text{proj}_c \Theta_{2,\text{cal}})$$

has covariation with respect to the  $(e, c)$  partition. Then, the calibration-correction method fails to solve this problem.

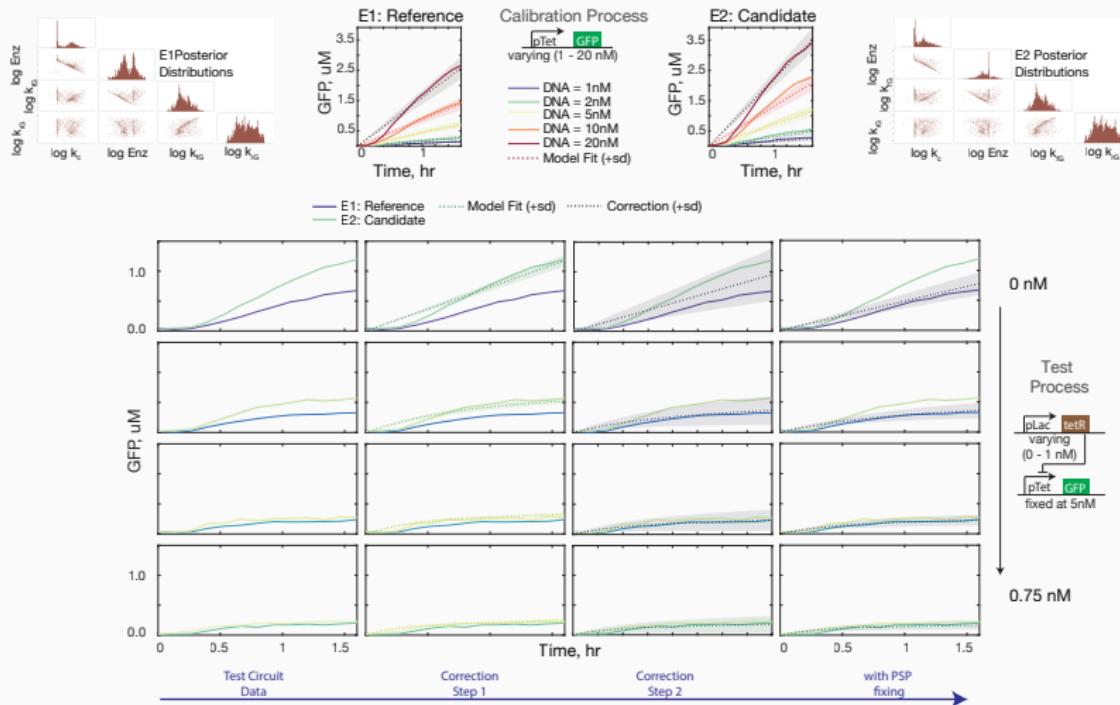
## Proposition

Consider the sets  $\Theta_{i,\text{cal}} \triangleq \text{ID}_\theta(\bar{y}_{i,\text{cal}}, M_{\text{cal}}(\theta))$  for  $i = 1, 2$ , and the partition  $\theta = (e, c)$ . Assume that the  $\Theta_{i,\text{cal}}$  have thin covariation in their  $c$  coordinates with respect to their  $e$  coordinates. Then, the calibration-correction method with CSP fixing is able to solve the DCP for the 'Test = Calib' case.

# PSP Fixing Helps for New Test Data Too



# PSP Fixing Helps with Experimental Data



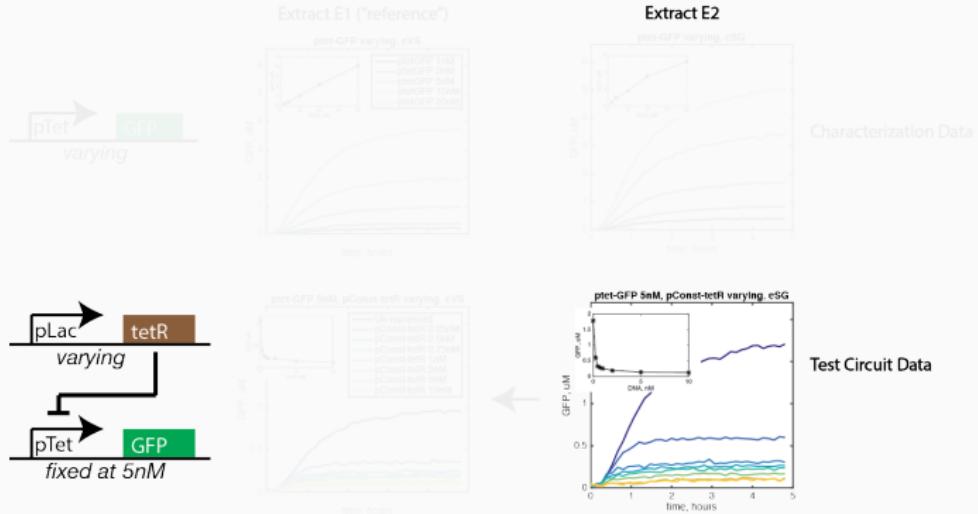
# Acknowledgements

Thanks to Richard Murray, Sam Clamons, Anandh Swaminathan, William Poole, Wolfgang Halter, Andras Gyorgy, and the Murray lab.

Research supported by:

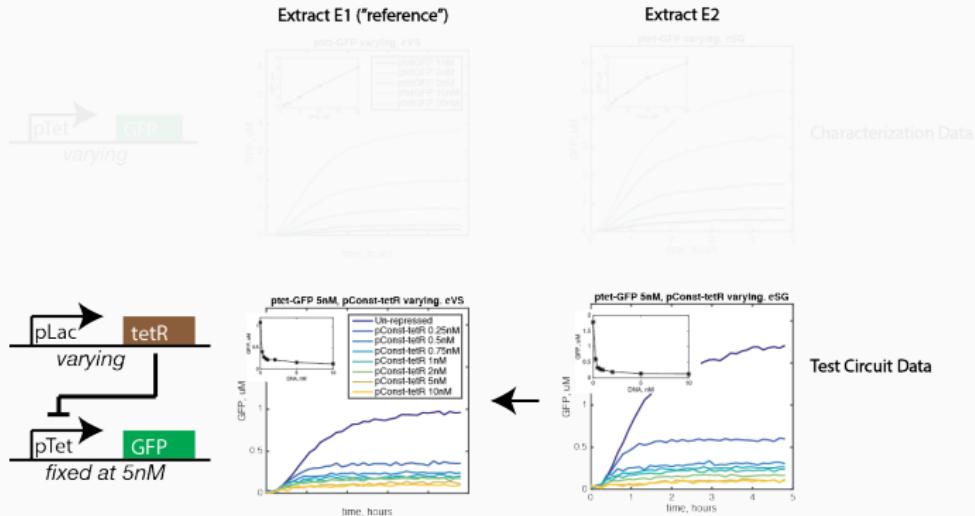


# The Data Correction Problem



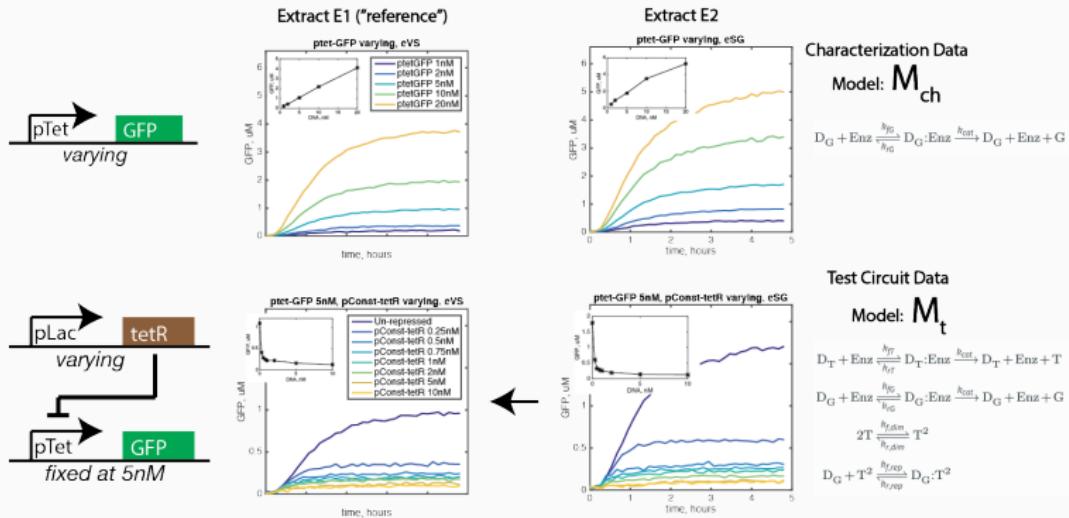
**Figure 1:** Given: Experimental data ( $\bar{y}_{2,t}$ ) for a circuit in an extract, E2. The subscript t stands for test circuit.

# The Data Correction Problem



**Figure 2:** Goal: Transform it into the behavior ( $\bar{y}_{1,t}$ ) of that experiment in a different extract, E1, which we call the 'reference' extract.

# The Data Correction Problem



**Figure 3:** Available tools: Choice of a set of characterization experiments ( $(\bar{y}_{1,ch}, \bar{y}_{2,ch})$ ) and corresponding models for both the characterization ( $M_{ch}$ ) and test ( $M_t$ ) circuits.

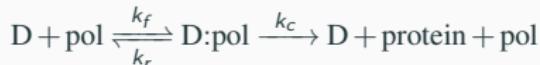
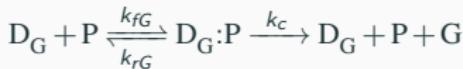
# The Data Correction Problem (DCP)

- Have
  - Data for test circuit in E2, the candidate extract,  $\bar{y}_{2,t}$
  - Ability to choose characterization experiments, and collect data  $\bar{y}_{1,ch}, \bar{y}_{2,ch}$
  - Ability to choose models,  $M_{ch}$  and  $M_t$
- Find a method that gives  $\hat{y}_{1,t}$ , such that  $\hat{y}_{1,t} = \bar{y}_{1,t}$

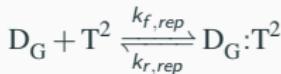
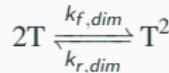
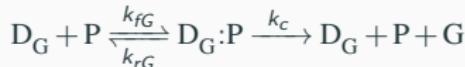
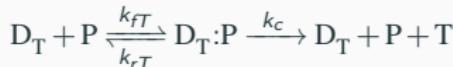
# Some Notation

- *Extracts vs Circuits*
- Model parameter vector ( $\theta$ ) coordinates partitioned into Extract Specific Parameters ( $e$ ) and Circuit Specific Parameters ( $c$ ), so that  $\theta = (e, c)$

# ESPs vs CSPs



- Environment specific parameters, e:  $[P]_{\text{total}}$ ,  $k_c$
- Process specific parameters, c: everything else



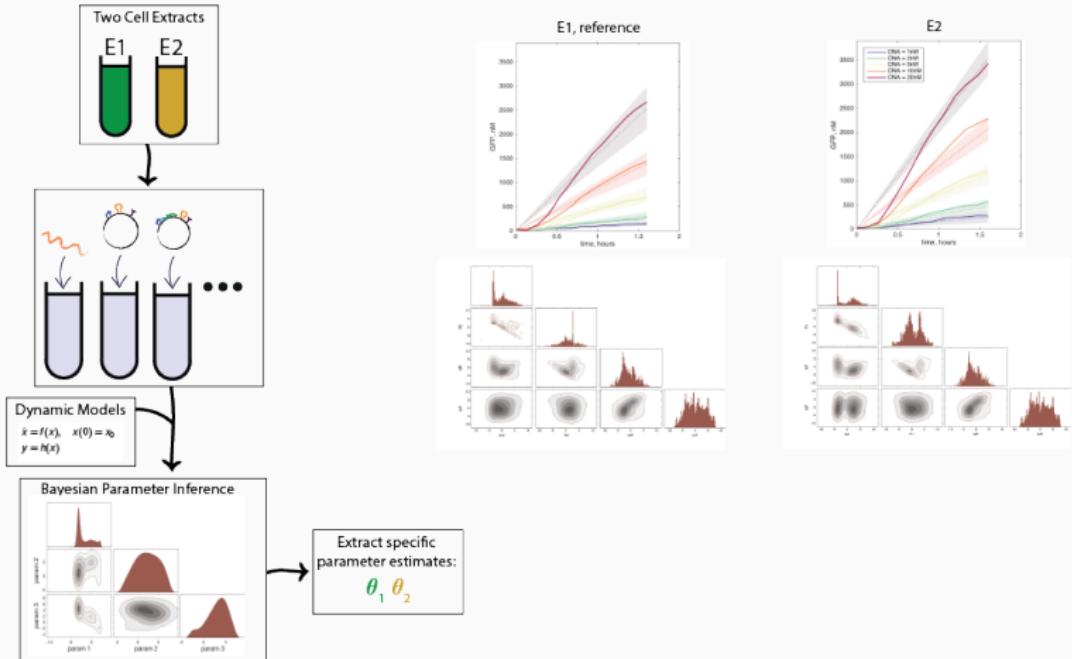
## Proposed Solution

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# The Calibration-Correction Method - Calibration Step



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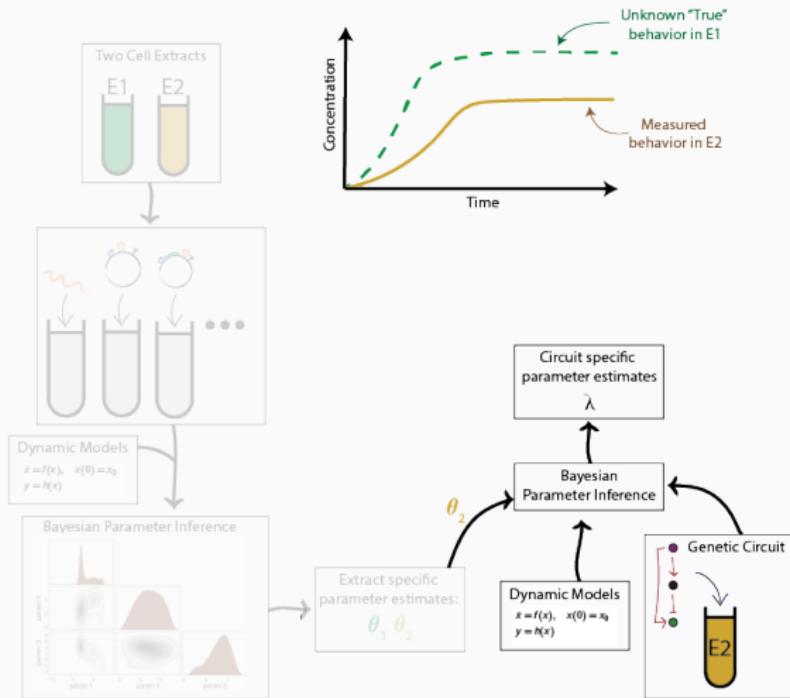
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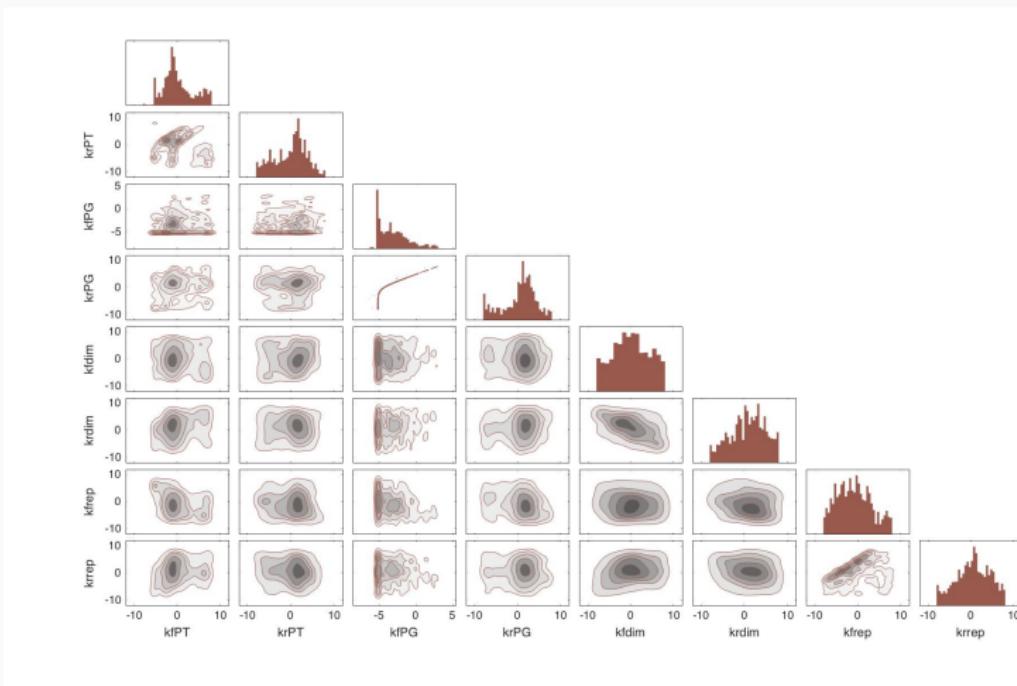
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  - All such points:

$$E_{i,ch} \triangleq \text{proj}_e \text{ID}_\theta(\bar{y}_{i,ch}, M_{ch}) \quad \forall i \in \{1, 2\}. \quad (1)$$

# The Calibration-Correction Method - Correction Step1



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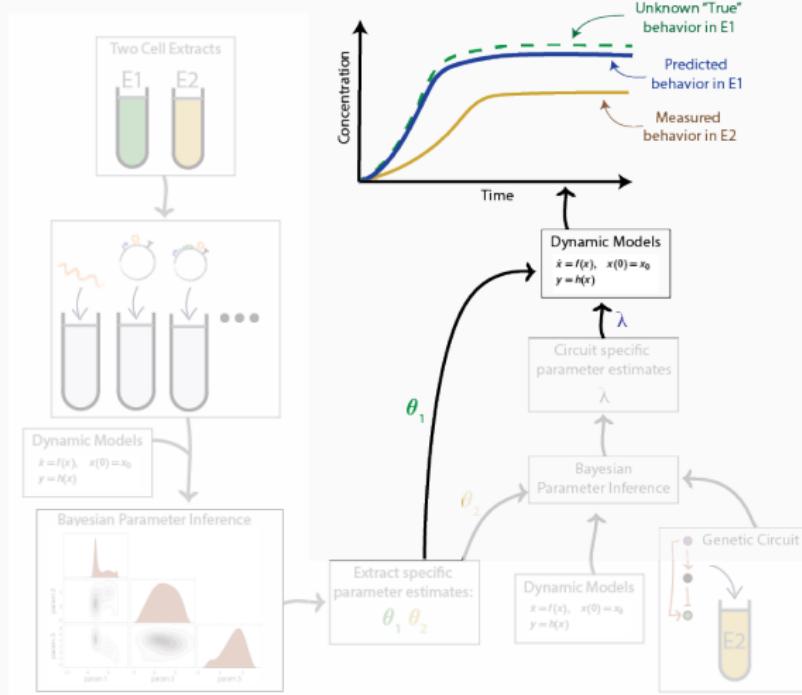
- **Correction Step One:** Test data, Extract 2. ( $\bar{y}_{2,t}$ ) + extract parameters fixed at  $\hat{e}_{2,ch} \rightarrow$  identify test parameters  $\hat{c}_{2,t}$

# The Calibration-Correction Method - Correction Step 1

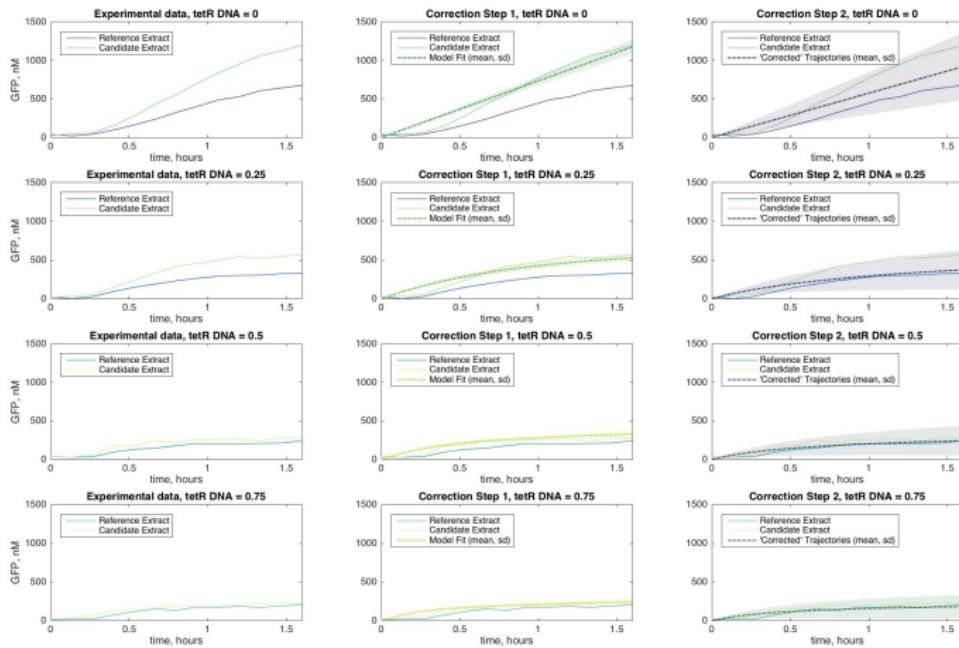
- **Correction Step One:** Test data, Extract 2.  $(\bar{y}_{2,t})$  + extract parameters fixed at  $\hat{e}_{2,ch} \rightarrow$  identify test parameters  $\hat{c}_{2,t}$ 
  - All such points:

$$C'_{c,t} \triangleq \bigcup_{\hat{e} \in E_{2,ch}} \text{ID}_{c|e=\hat{e}} \left( \bar{y}_{2,t}, M_t(e, c) \right)$$

# The Calibration-Correction Method - Correction Step 2



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- **Correction Step Two:** Predict  $\hat{y}_{1,t} = M_t(\hat{e}_{1,ch}, \hat{c}_{2,t})$ 
  - All such trajectories:

$$Y_1 \triangleq \bigcup_{\hat{e} \in E_{1,ch}} \bigcup_{\hat{c} \in C'_{2,t}} \tilde{y}_1(\hat{e}, \hat{c}), \quad (2)$$

where,

$$\tilde{y}_1(\hat{e}, \hat{c}) = M_t(\hat{e}, \hat{c}). \quad (3)$$

## Failure Conditions for this Method

- A parameter identification step is attempted when no parameter exists such that the model fits the data
- Correction step two is able to produce a trajectory not equal to the true trajectory

# Results

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$$E_{2,t} \triangleq \text{proj}_e \text{ID}_{\theta}(\bar{y}_{2,t}, \bar{M}_t).$$

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$$E_{1,ch} \subseteq E_{1,t}$$

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Proof Ideas: If  $E_{2,ch} \subseteq E_{2,t}$  does not hold, parameter estimation of the  $C_{2,t}$  fails in first correction step. If  $E_{1,ch} \subseteq E_{1,t}$  does not hold, then the prediction of a correct trajectory  $\hat{y}_{1,t}$  is impossible for any  $c$ .

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Proof Idea: In the second correction step, the prediction  $\hat{y}_{1,t}$  is generated by plugging in an arbitrary point from  $C'_{2,t}$  into what would have been  $C'_{1,t}$ .

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Proof idea: In the second correction step, we pick arbitrary points from the sets  $E_{1,ch}$  and  $C'_{2,t}$  as a proxy for a point in  $\Theta_{1,t}$ .

## Supplement

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## Definition (Parameter Identification)

Let the set  $\mathcal{M}_Y$  be the set of all pairs  $(\bar{y}, M(\theta, u, x_0))$  for which there exists a parameter  $\hat{\theta} \in \Omega$  such that  $\bar{y} = M(\hat{\theta}, u, x_0)$ . Also, let  $\mathcal{P}(\Omega)$  be the power set of  $\Omega$ . We define the *parameter identification of  $\theta$*  as an operator  $ID_\theta : \mathcal{M}_Y \rightarrow \mathcal{P}(\Omega)$ , with  $ID_\theta(\bar{y}, M) = \{\hat{\theta} \in \Omega \mid \bar{y} = M(\hat{\theta}, u, x_0)\}$

## Definition

$M(\theta_A)$  and  $M(\theta_B)$  *output-indistinguishable* if,

$$\theta_A, \theta_B \in \Omega, \\ y(t, \theta_A, u, x_0) = y(t, \theta_B, u, x_0) \quad \forall t \geq 0, \forall u \in \mathcal{U}, \forall x_0 \in \chi. \quad (4)$$

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The  $i^{th}$  coordinate of  $\theta_A$ , denoted  $\theta_{A,i}$ , is *structurally globally identifiable* (SGI) if, for almost any  $\theta_A \in \Omega$ , (4) has a unique solution for  $\theta_{B,i}$ .

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## Definition

$M(\theta)$  SGI if all its parameters  $\theta_i$  are SGI.

## Proof: Necessity of $E_{2,ch} \subseteq E_{2,t}$

- AFSOC  $\exists \tilde{e} \in E_{2,ch}$  s. t.  $\tilde{e} \notin E_{2,t}$ .

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- This implies  $\nexists \tilde{c}$  s.t.  $M_t(\tilde{e}, \tilde{c}) = \bar{y}_{2,t}$ .
- $\implies$  parameter ID fails at correction step one (first failure condition).

## Proof: Necessity of $E_{1,ch} \times C'_{2,t} \subseteq \Theta_{1,t}$

- Split into three sub conditions

$$E_{1,ch} \subseteq E_{1,t} \triangleq \text{proj}_e \text{ID}_{\theta}(\bar{y}_{1,t}, \bar{M}_t),$$

$$C'_{2,t} \subseteq C'_{1,t},$$

$$E_{1,ch} \times C'_{2,t} \subseteq \Theta_{1,t},$$

## Sufficiency is straightforward

The argument is a simple exercise in checking that the parameters that get picked are in sets such that the correct trajectories get generated.