



# Antifragile feedback control from biological to technical systems

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Nuremberg Institute of Technology, Germany



**ANTIFRAGILITY**  
gain from uncertainty

Applied Antifragility Group

**ohm**  
Technische  
Hochschule  
Nürnberg

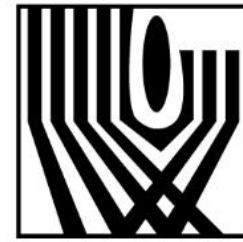
Cristian  
Axenie



Meisam  
Akbarzadeh

UNIVERSITY OF  
SURREY

Roman  
Bauer



**iimas**  
Oliver Lopez-  
Corona

Zürcher Hochschule  
für Angewandte Wissenschaften  
**zhaw** School of  
Engineering  
IDP Institut für Datenanalyse  
und Prozessdesign

UNITRENTO

MANCHESTER  
1824  
The University of Manchester

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CANCER CENTER

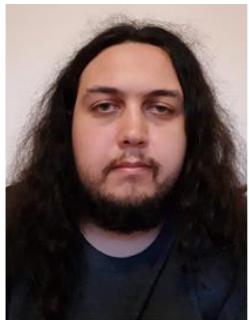
Michail  
Makridis

Matteo  
Saveriano

Alexandru  
Stancu

Jeffrey  
West





*Ertan Halilov*  
SHK



*Julian Main*  
SHK



*David Weiss*  
SHK



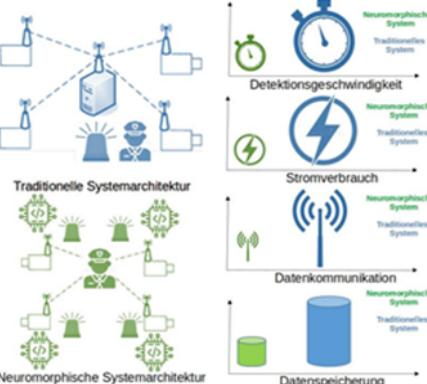
*Daniel Pommer*  
WissMit/  
Doktorand



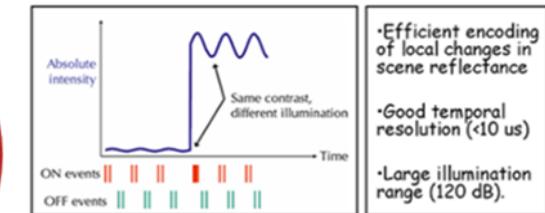
*Cristian Axenie*



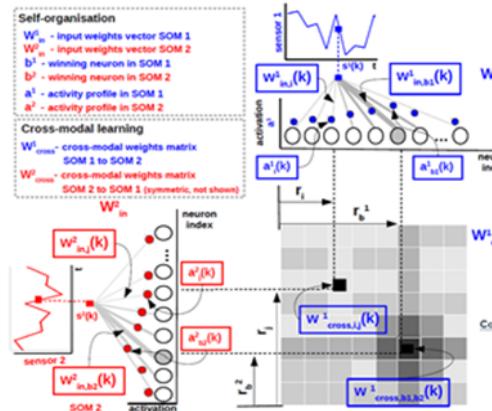
## Fast, Efficient, Robust and Green Inference



## Spike-based/Event-based data representation



## Learning algorithms



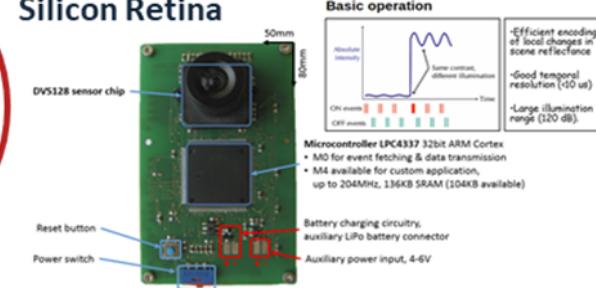
Algorithms for Spiking Neural Networks

## Neuromorphic Chips

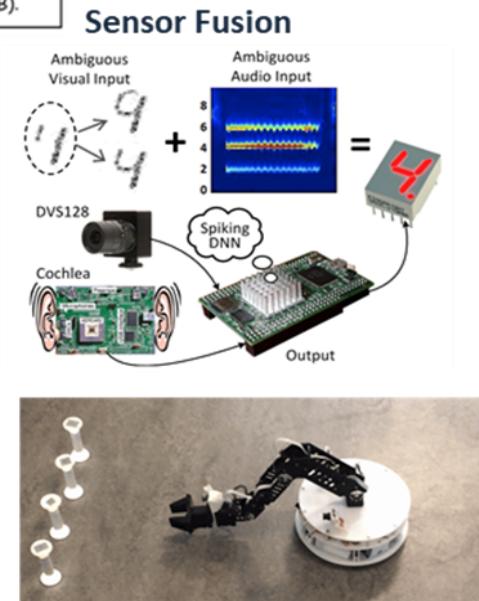


Neuromorphic Sensors and Computers

## Silicon Retina



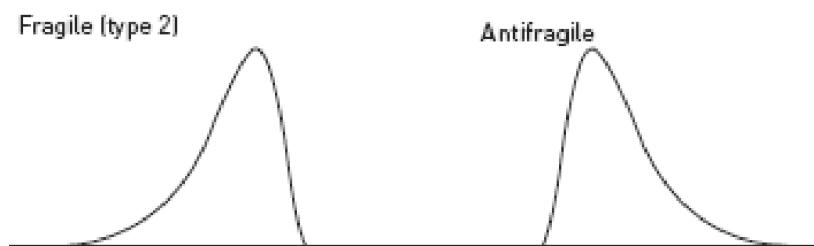
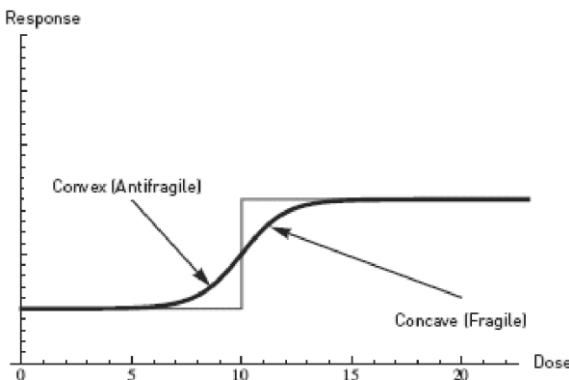
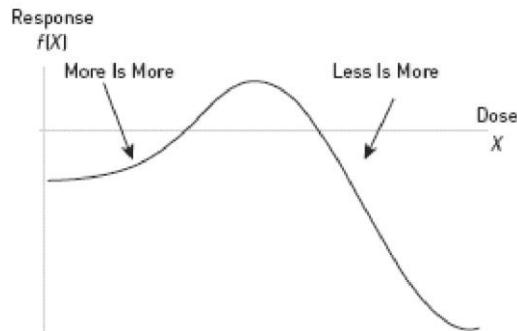
## Closed-loop Sensorimotor Control



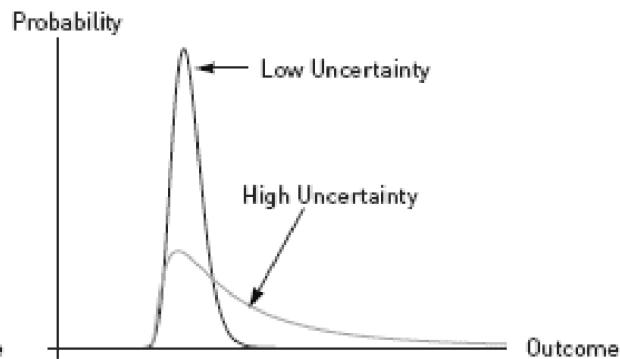
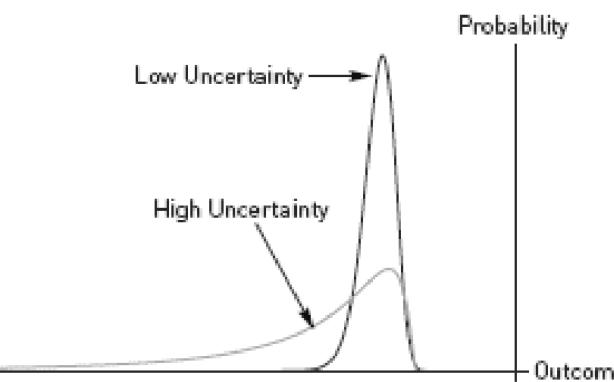
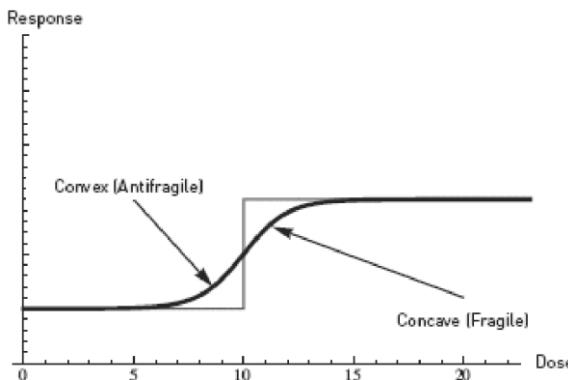
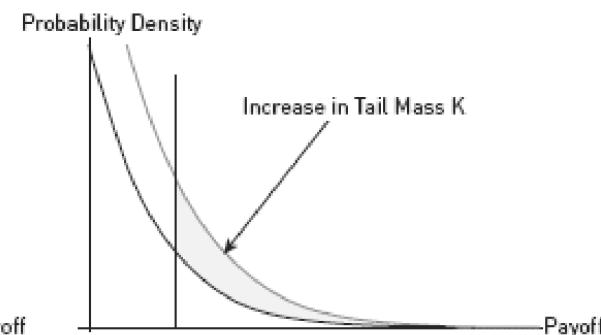
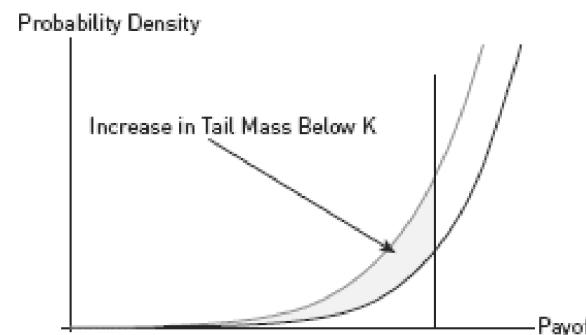
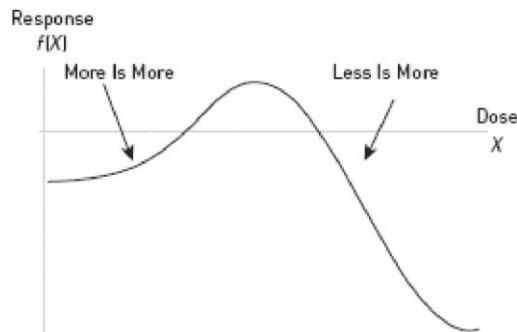
# Outline

- Antifragility
- Beyond probability distributions
- Antifragile feedback control
  - Mechanistic design
  - Learning from biology
  - Transferring to technology

# Antifragility

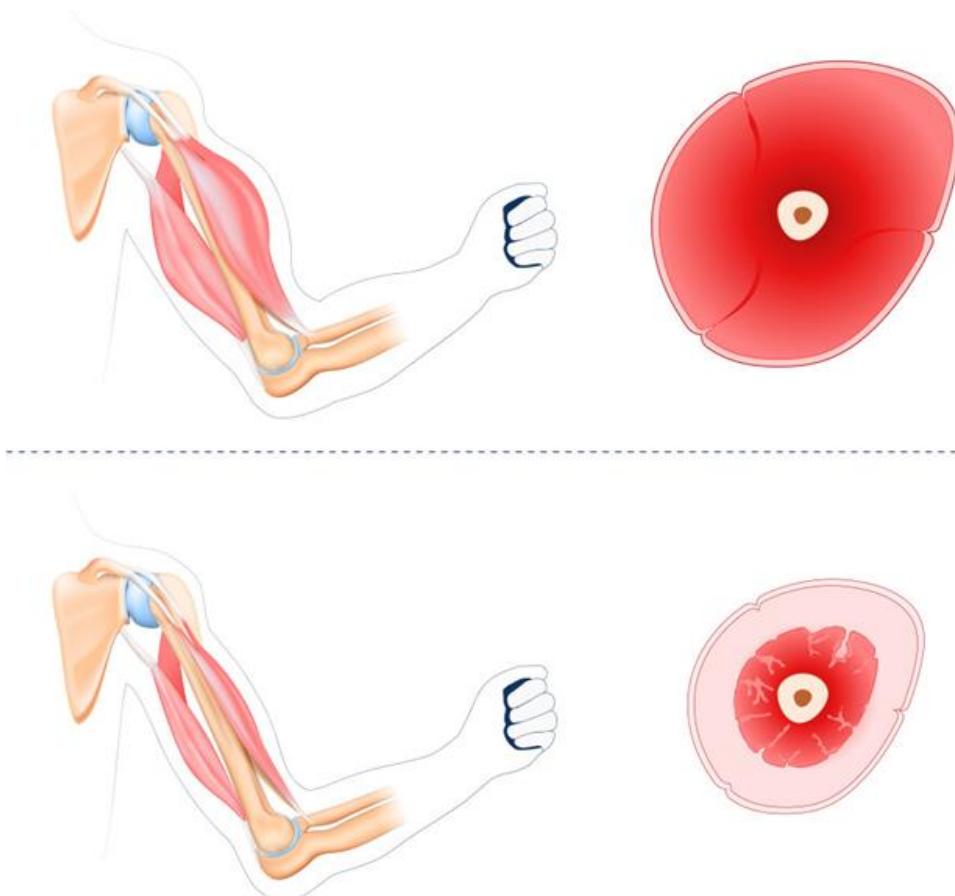


# Antifragility

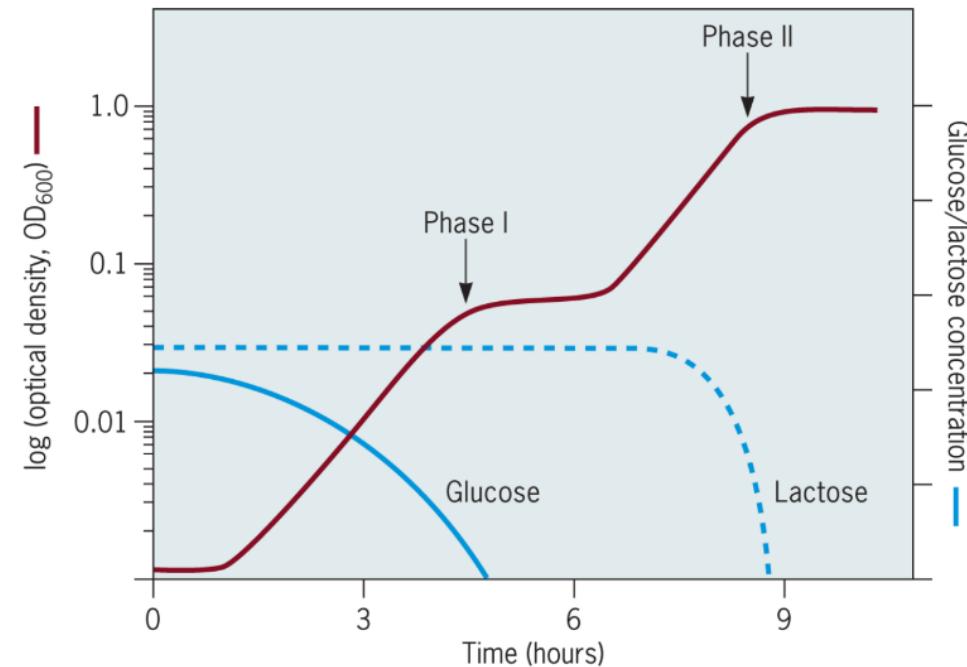


# Beyond probability distributions

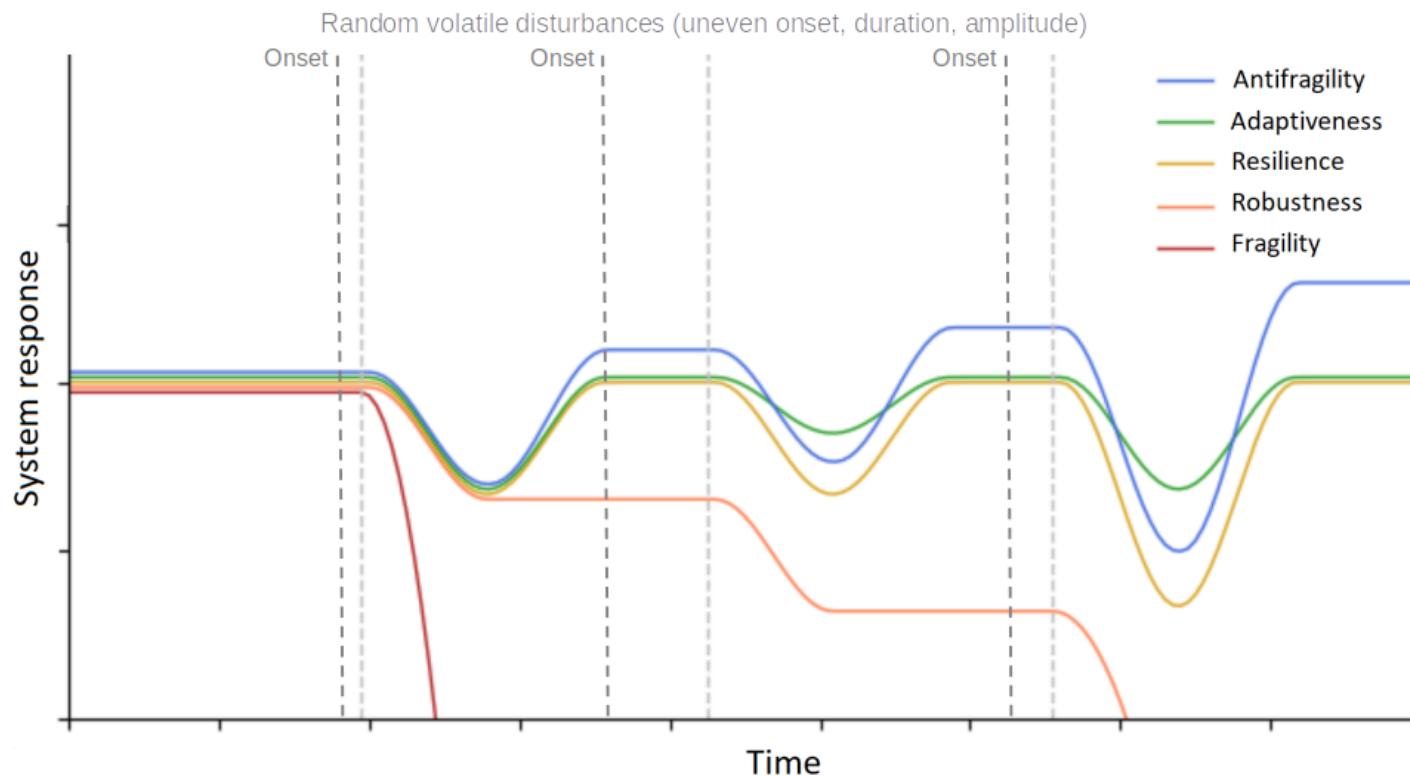
Muscle hypertrophy – atrophy



E-Coli lactose to glucose metabolic switch



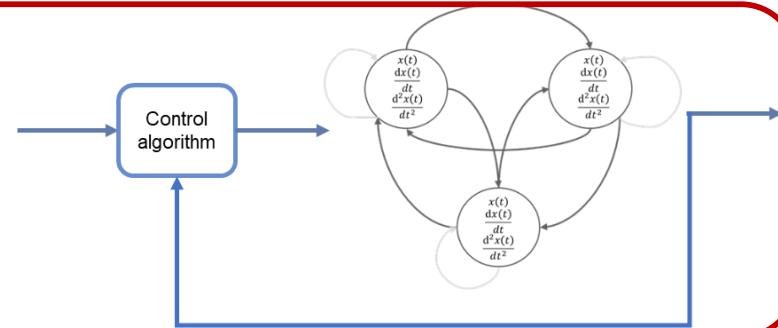
# Beyond probability distributions



# Beyond probability distributions

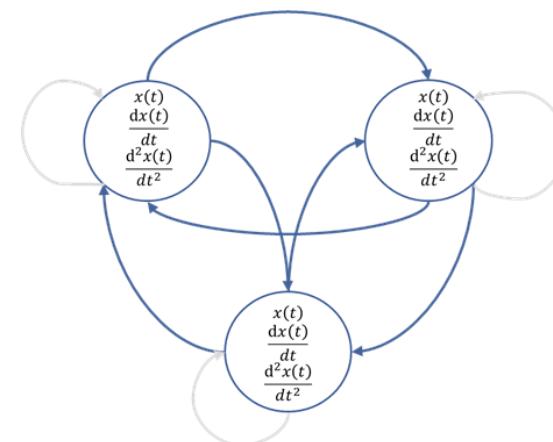
## Induced/interventional antifragility

Induced and interventional antifragile systems benefit from harm derived from input distribution unevenness based on emergent system dynamics in closed-loop with a controller driving the system towards prescribed dynamics in the presence of modulated or non-stationary disturbances, noise, and volatility.



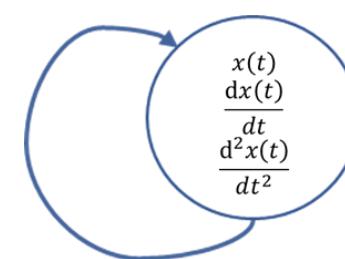
## Inherited/evolutionary antifragility

Inherited and evolutionary antifragile systems benefit from harm derived from input distribution unevenness, based on the emergent system dynamics and its interactions with the operating environment (i.e. disturbances, noise, modulated perturbations).

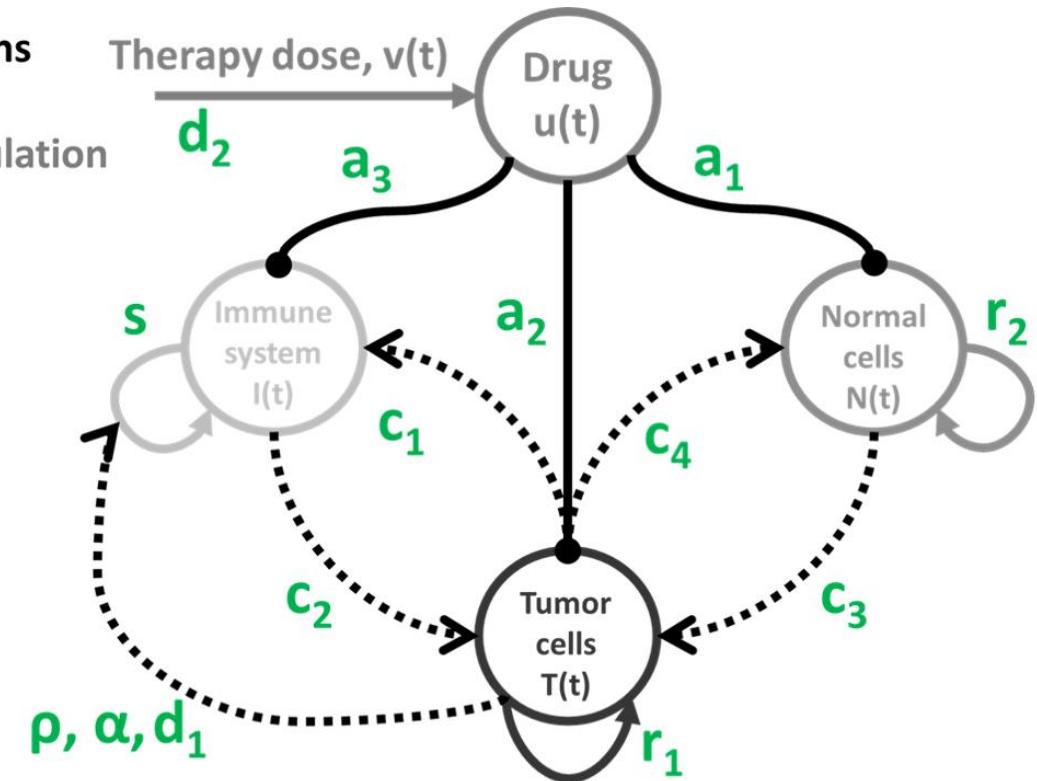
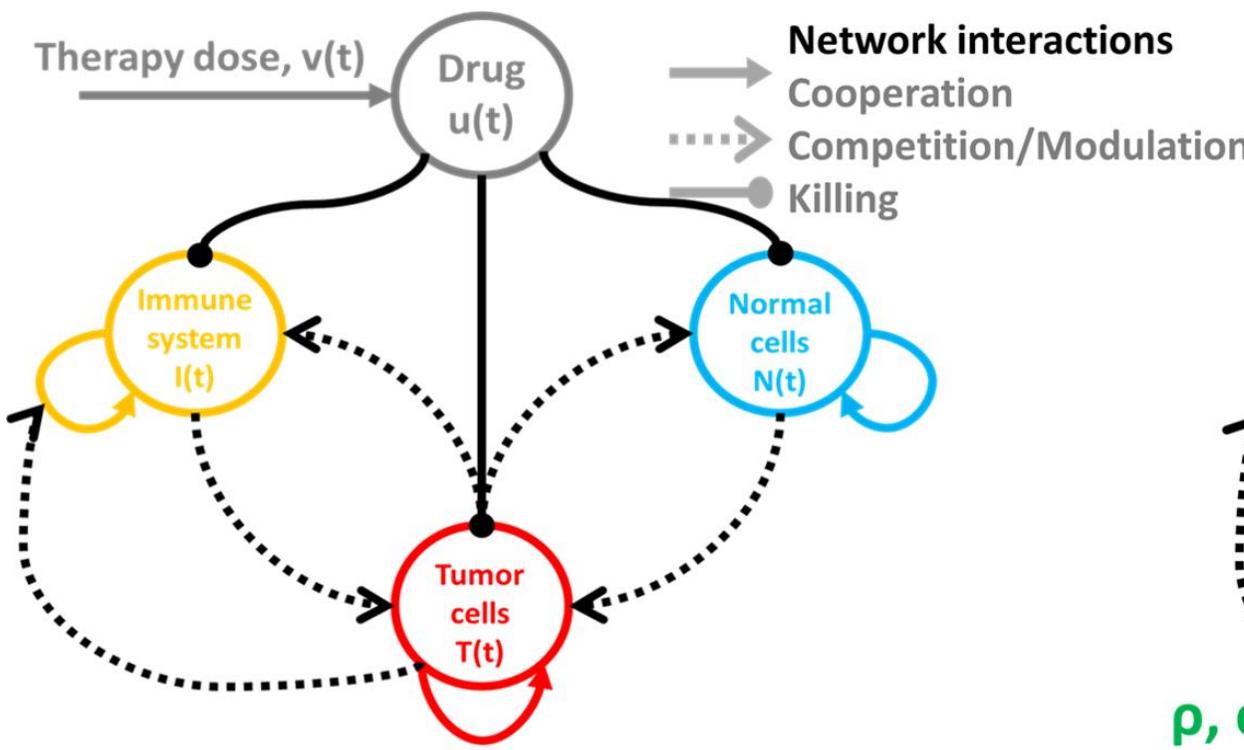


## Intrinsic/ecological antifragility

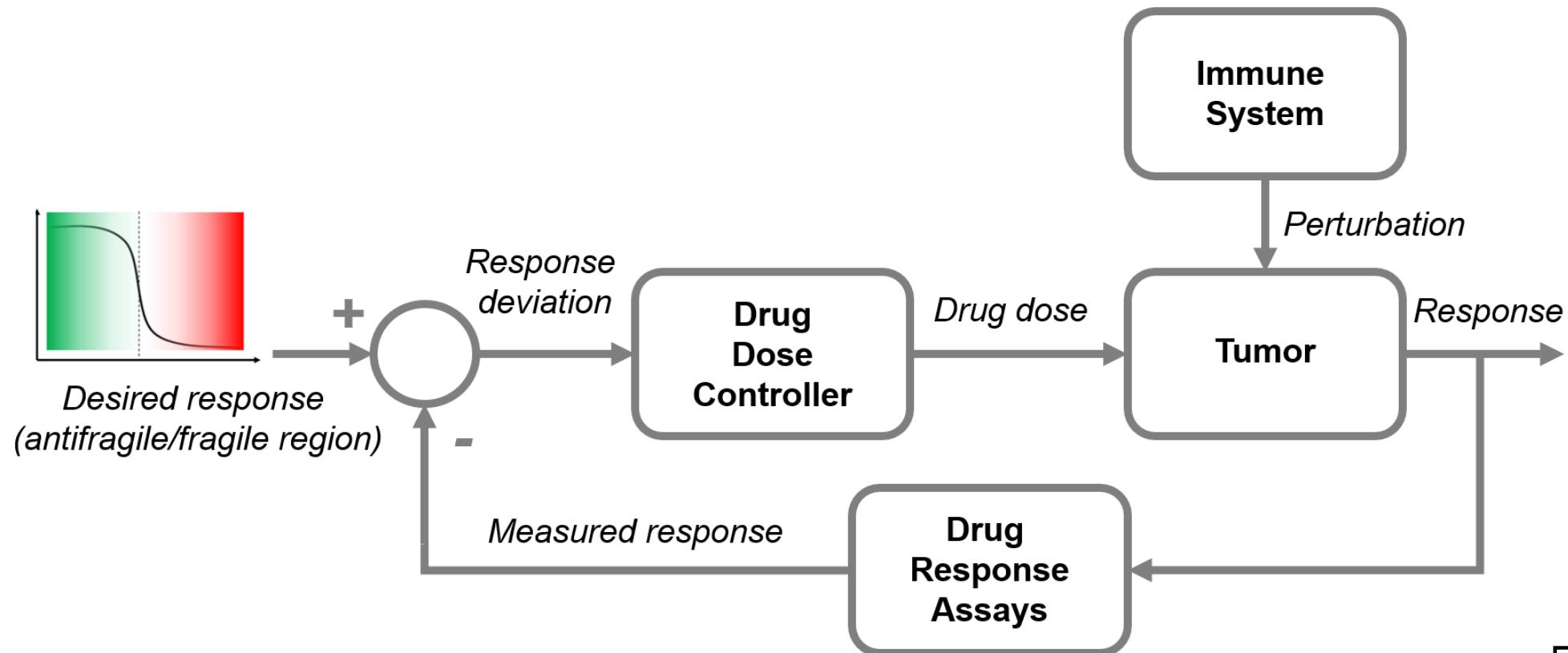
Intrinsic and ecological antifragile systems benefit from harm derived from internal dynamics distribution unevenness, based on the convexity of the response function of the system without external input and solely based on the internal components' heterogeneity and resilience.



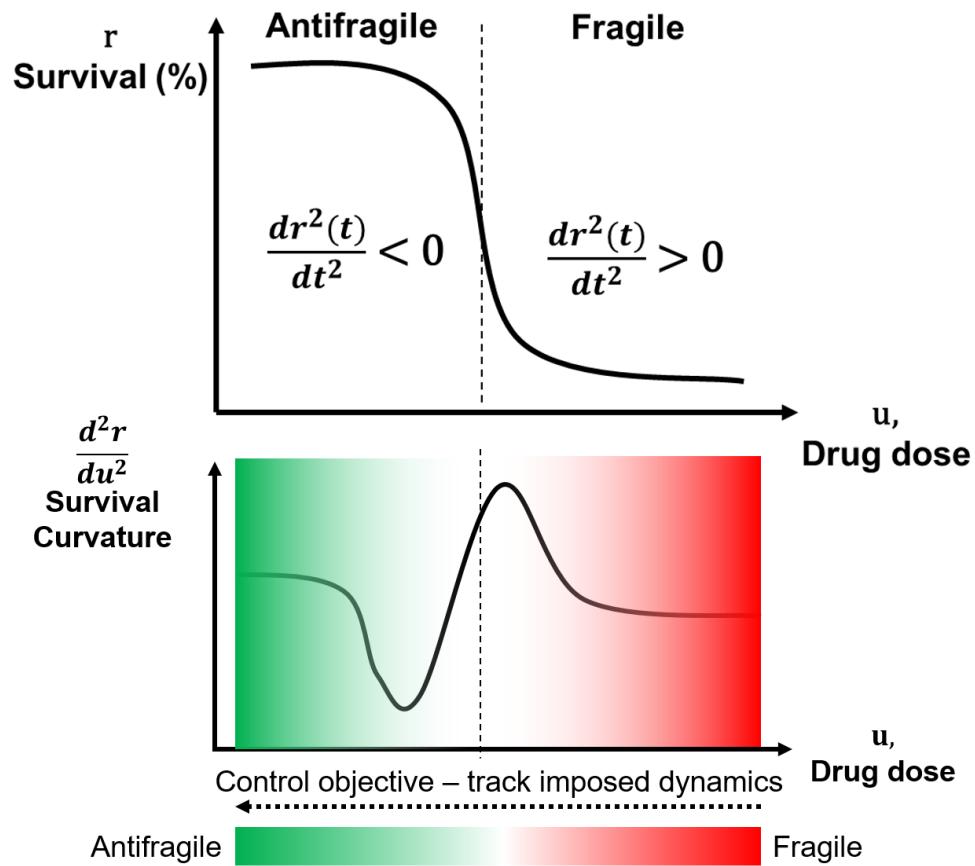
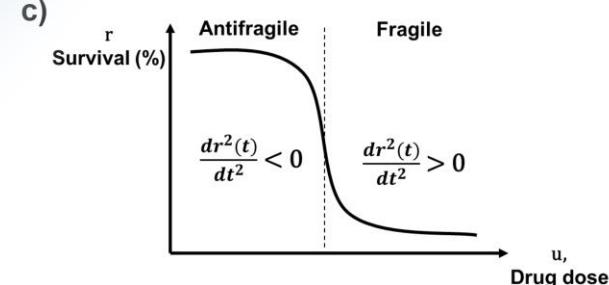
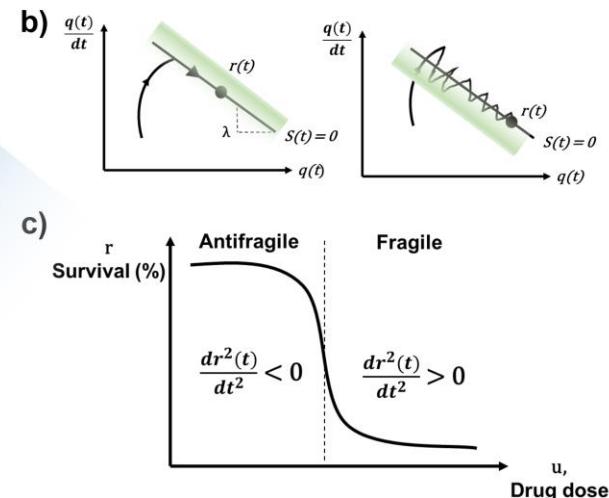
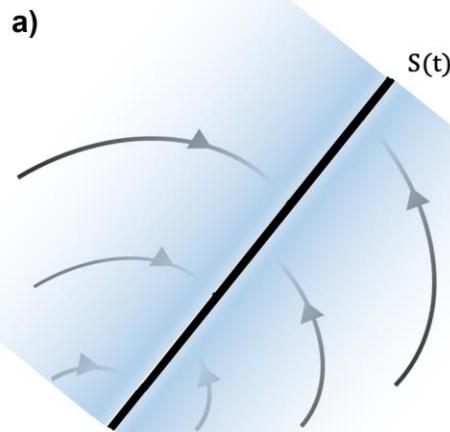
# Antifragile feedback control: mechanistic design



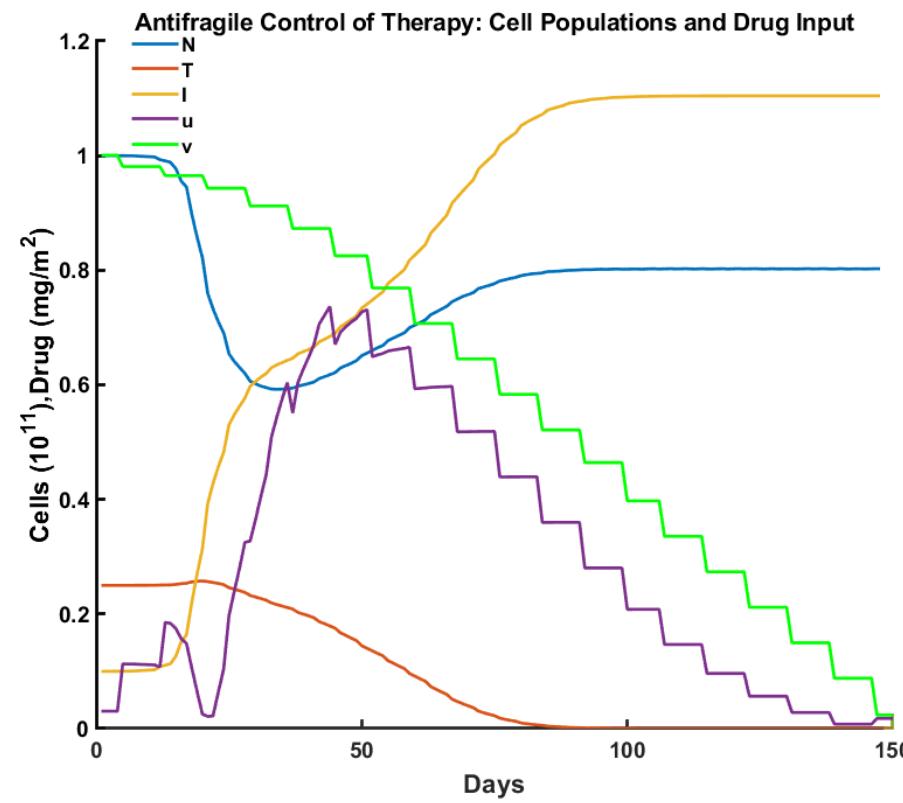
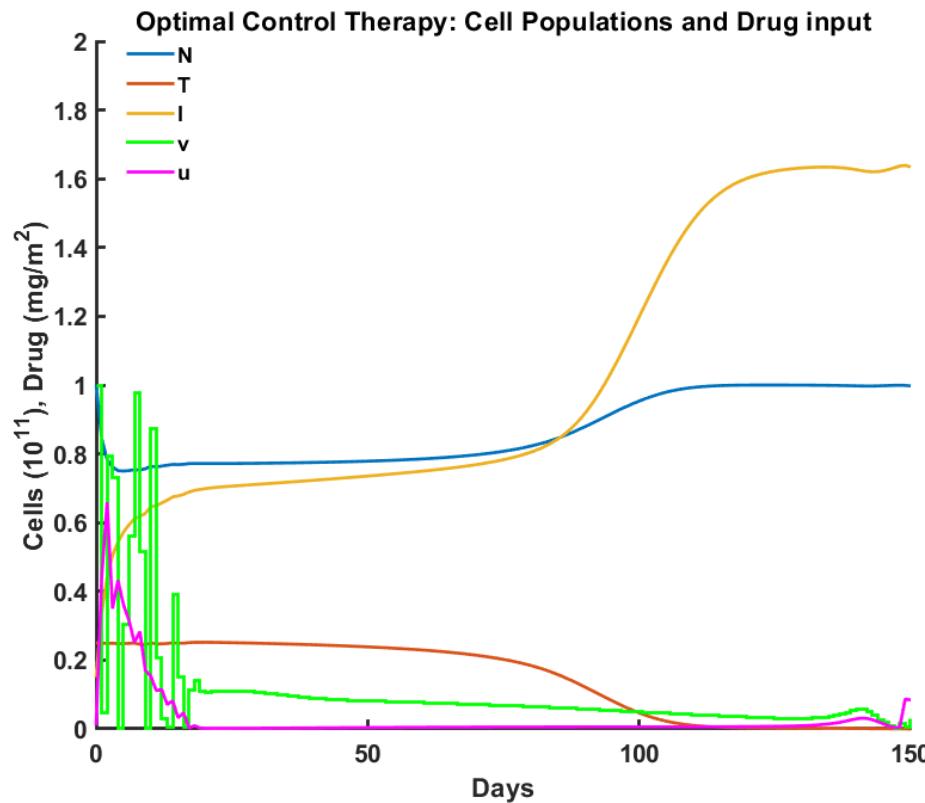
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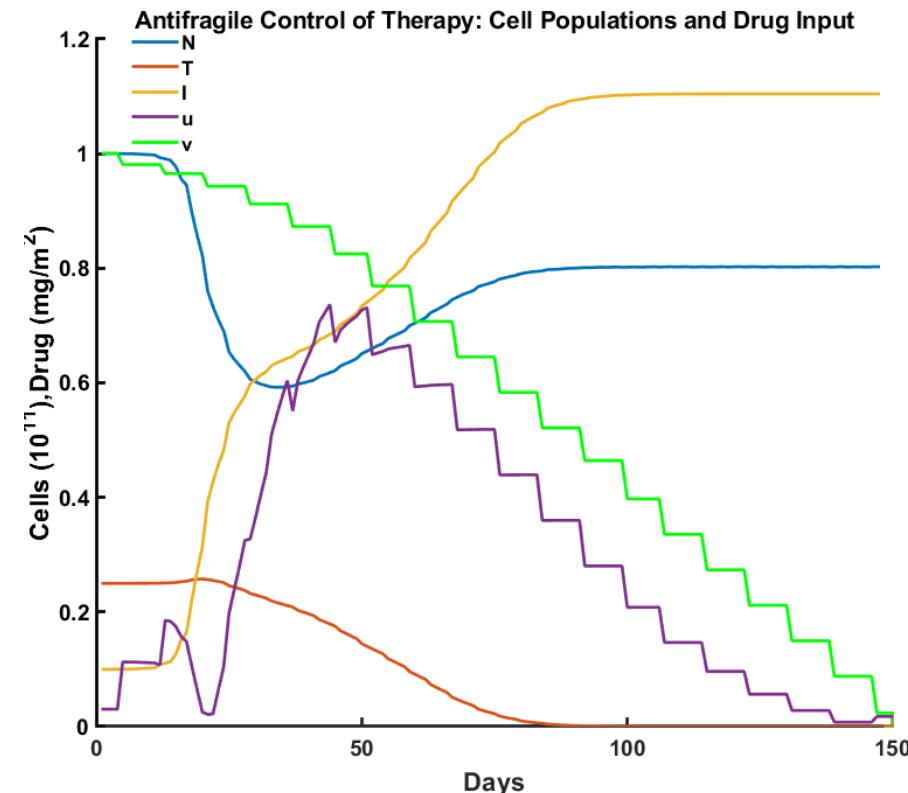
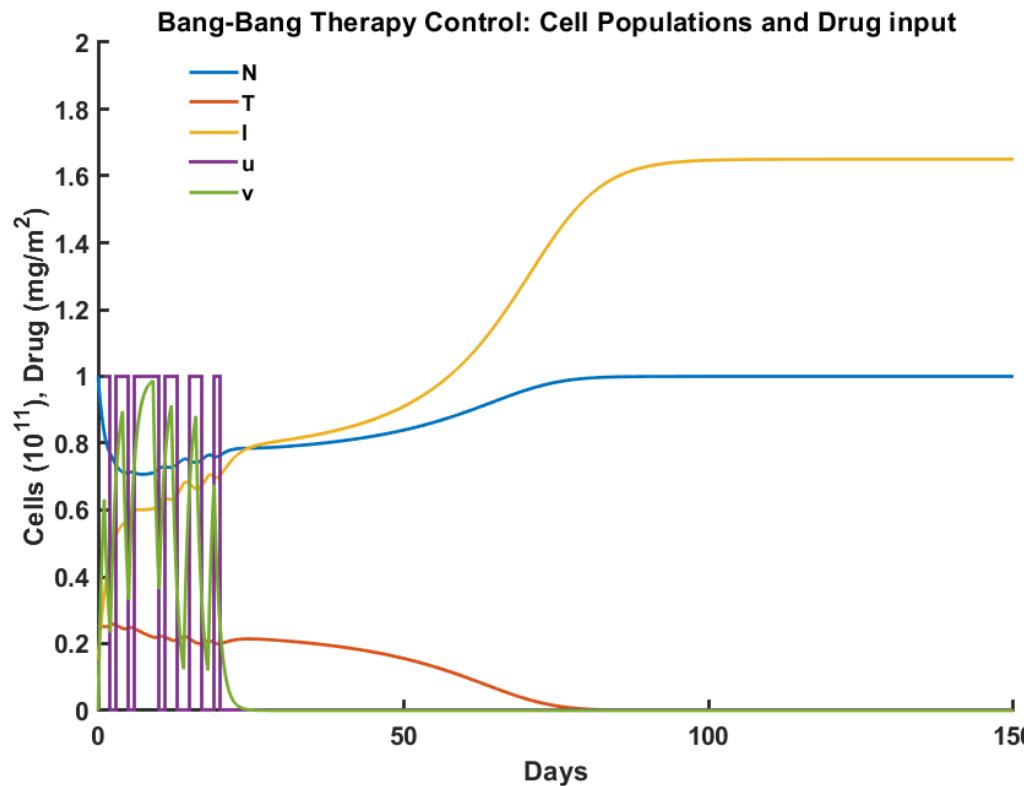
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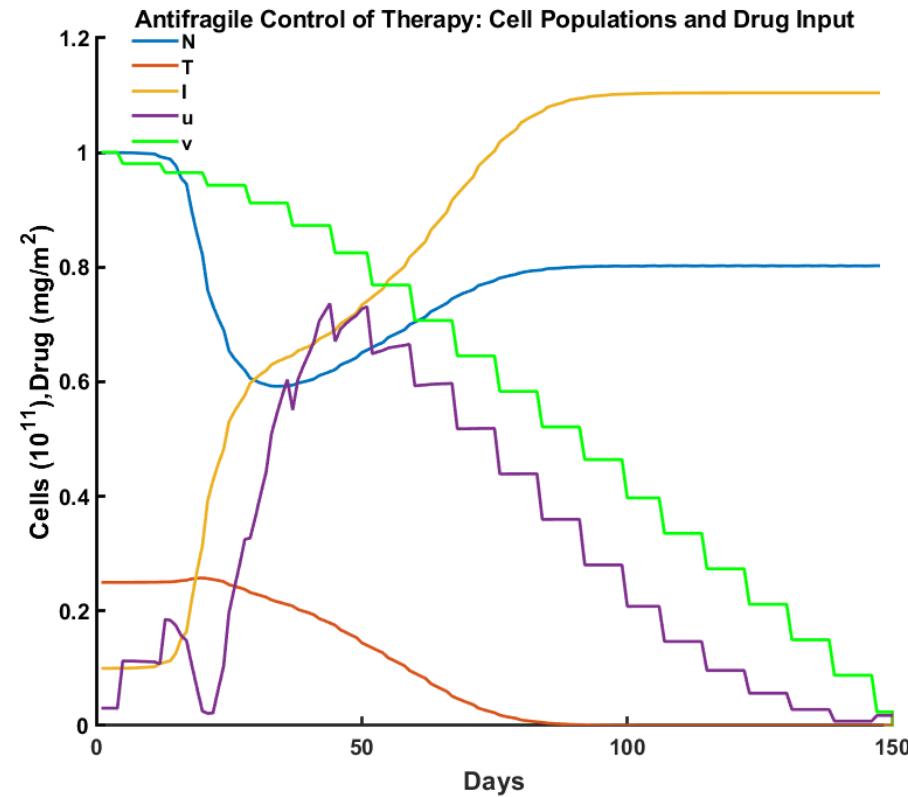
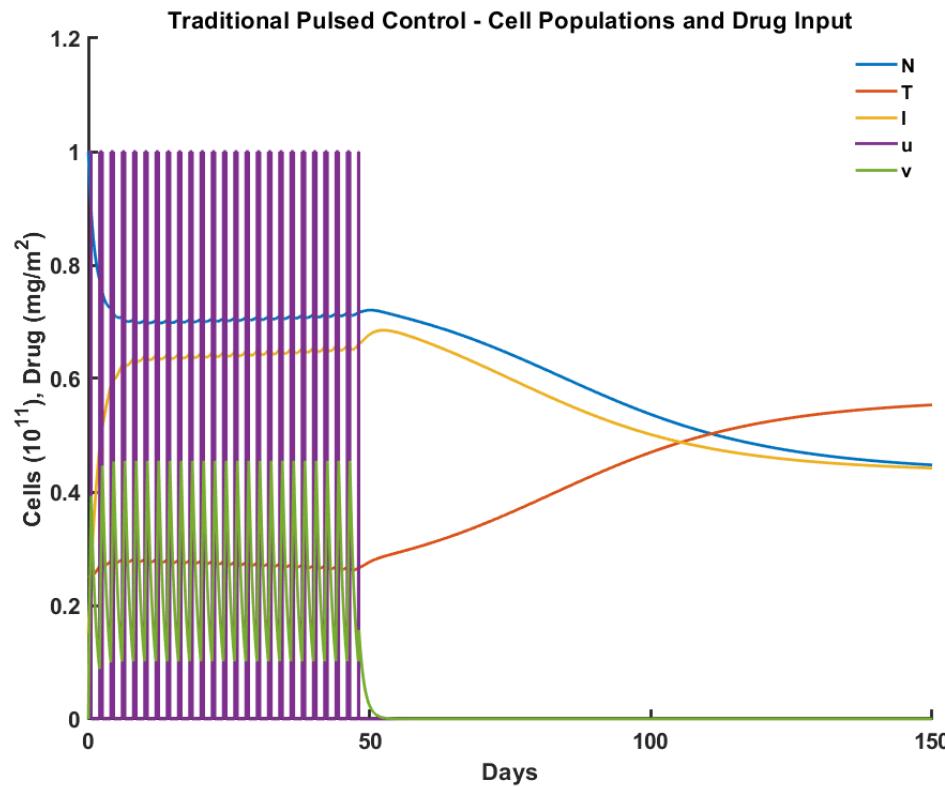
# Antifragile feedback control: mechanistic design



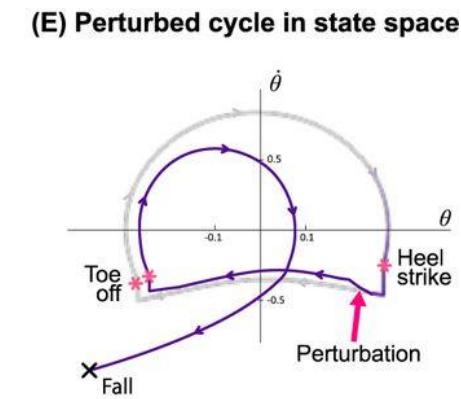
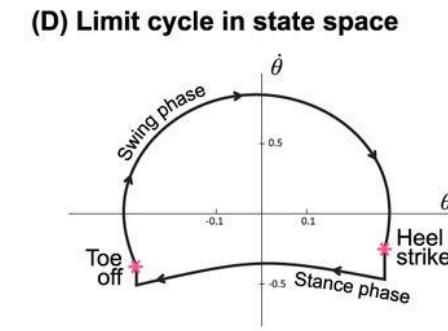
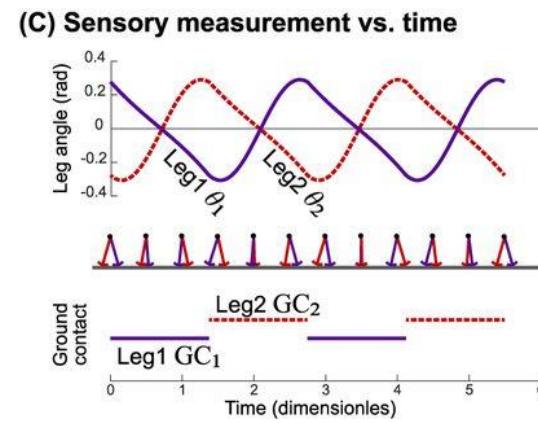
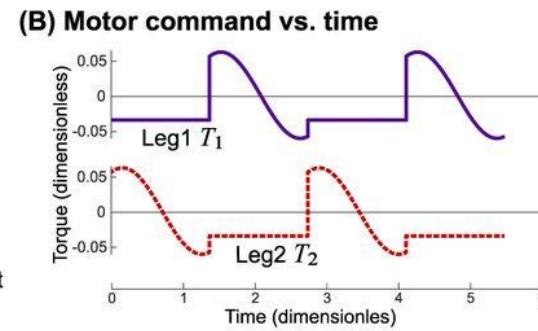
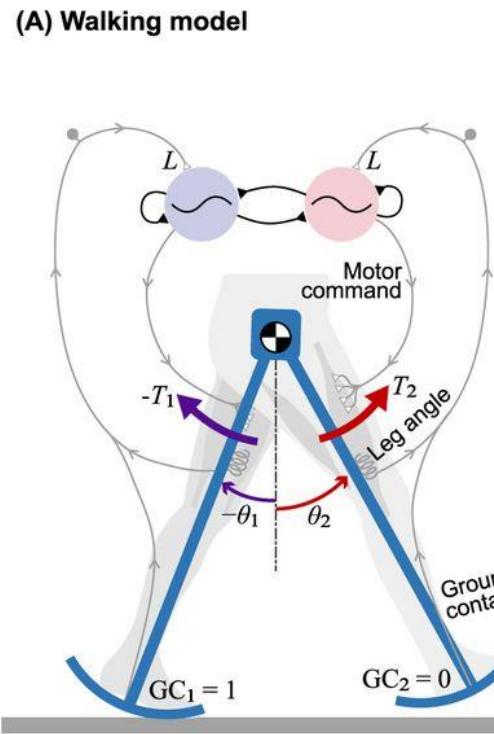
# Antifragile feedback control: mechanistic design



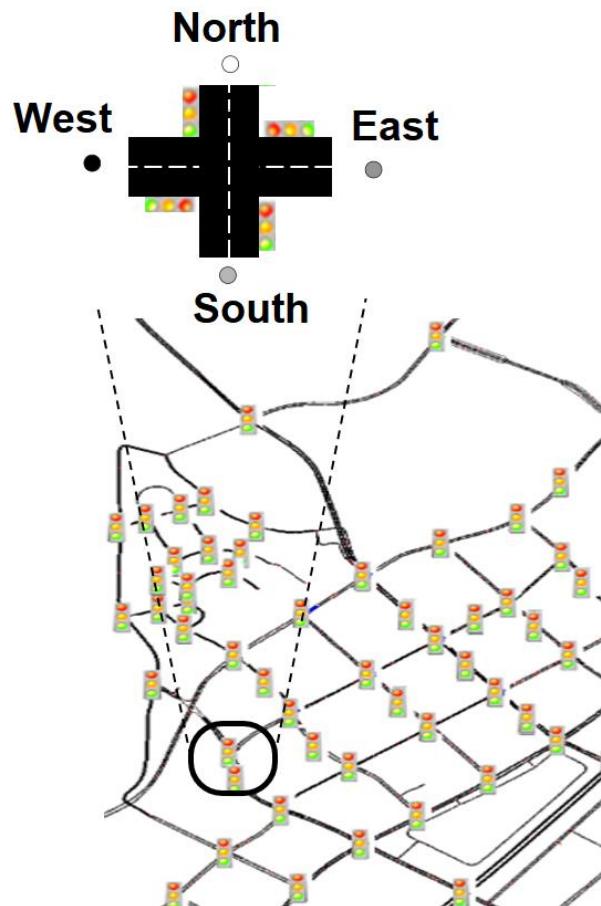
# Antifragile feedback control: mechanistic design



# Antifragile feedback control: learning from biology



# Antifragile feedback control: learning from biology



Single cross dynamics



# Antifragile feedback control: learning from biology

## Oscillator-based modelling and nonlinear control

$$\frac{d\theta_i(t)}{dt} = \omega_i(t) + k_i(t) \sum_{j=1}^N A_{ij} \sin(\theta_j(t) - \theta_i(t)) + F_i \sin(\theta^*(t) - \theta_i(t)) + u_i(t)$$

Internal properties      Network coupling      External coupling      Control law

with

$$u_i(t) = \epsilon_1 \int_0^t \hat{s}_i(\tau) d\tau$$

$$\frac{d\hat{s}_i(t)}{dt} = \epsilon_2 \left( \sum_{i,j} (\hat{s}_j(t) - \hat{s}_i(t)) + s_i(t) \right)$$

$$\frac{ds_i(t)}{dt} = \epsilon_3 \sum_j (s_j(t) - \frac{d\hat{s}_i(t)}{dt}) - \text{sign}(\hat{s}_i(t)) \frac{d^2\theta_i(t)}{dt^2}$$

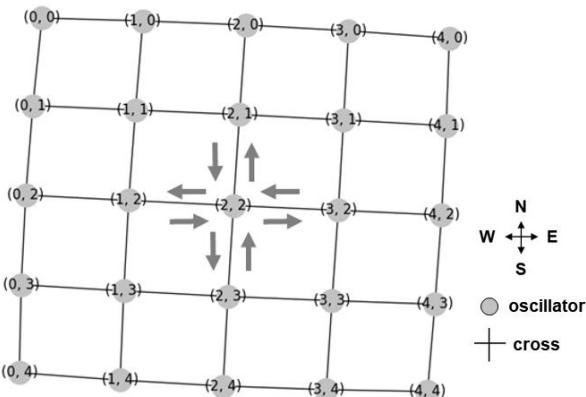
$$0 < \epsilon_1 < \epsilon_2 < \epsilon_3 < 1$$

where:

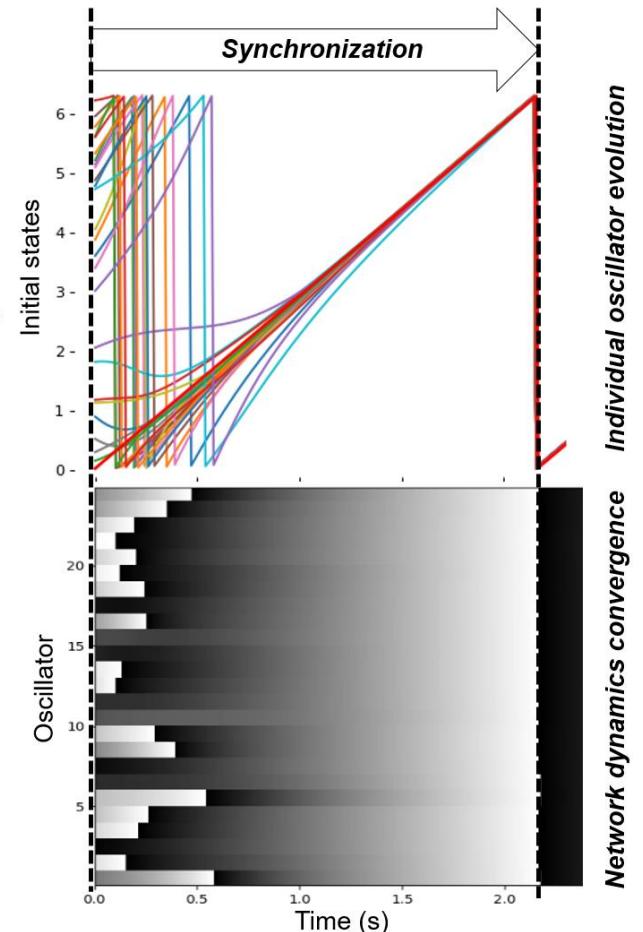
$s_i(t)$  - the surplus energy of traffic light  $i$  oscillator

$\hat{s}_i(t)$  - the estimated surplus energy of traffic light  $i$  oscillator

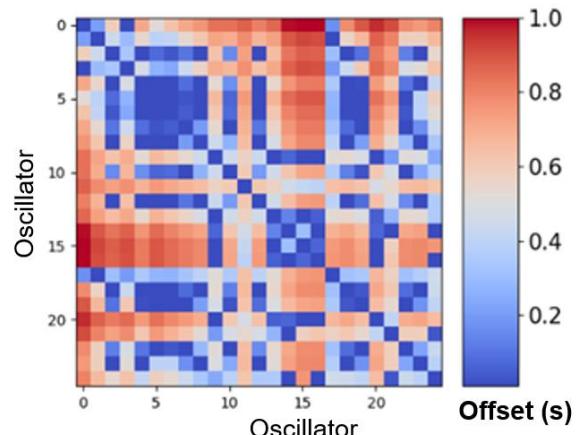
a. Example road network topology



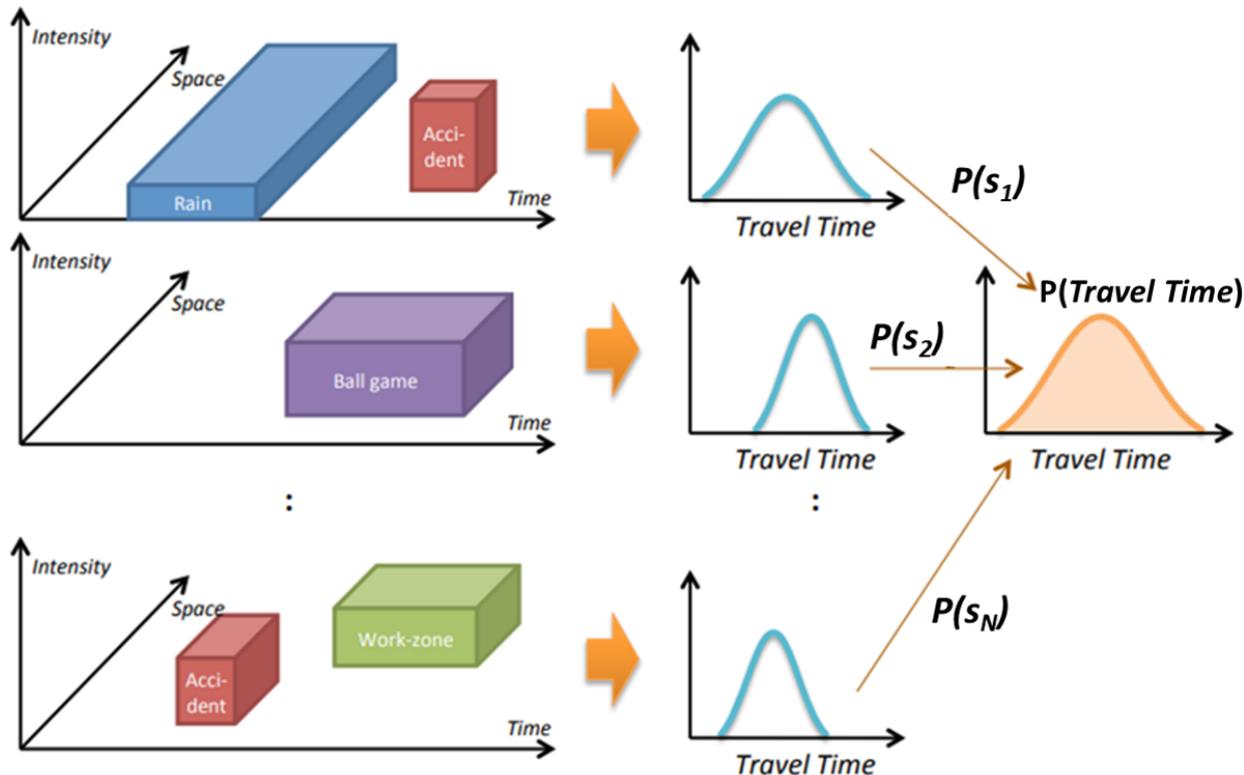
b. Dynamics of the oscillator network



c. Time to synchronization (offset)



# Antifragile feedback control: transfer to technology



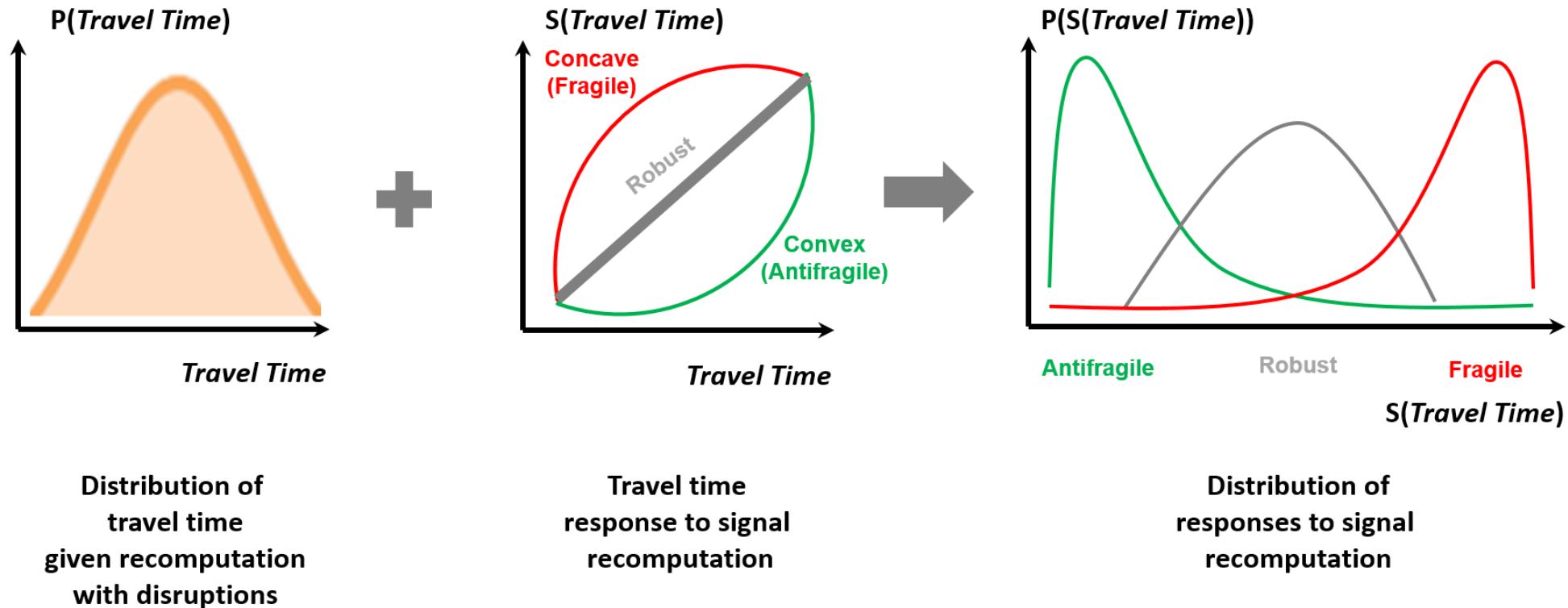
Uncertainty sources  
(space-time-intensity  
traffic disruptions)

Signal  
recomputation  
alters travel time distribution

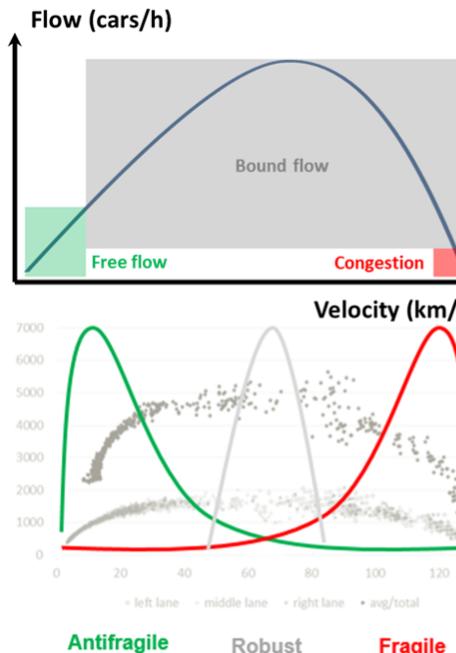
Travel time  
responses to signal  
recomputation



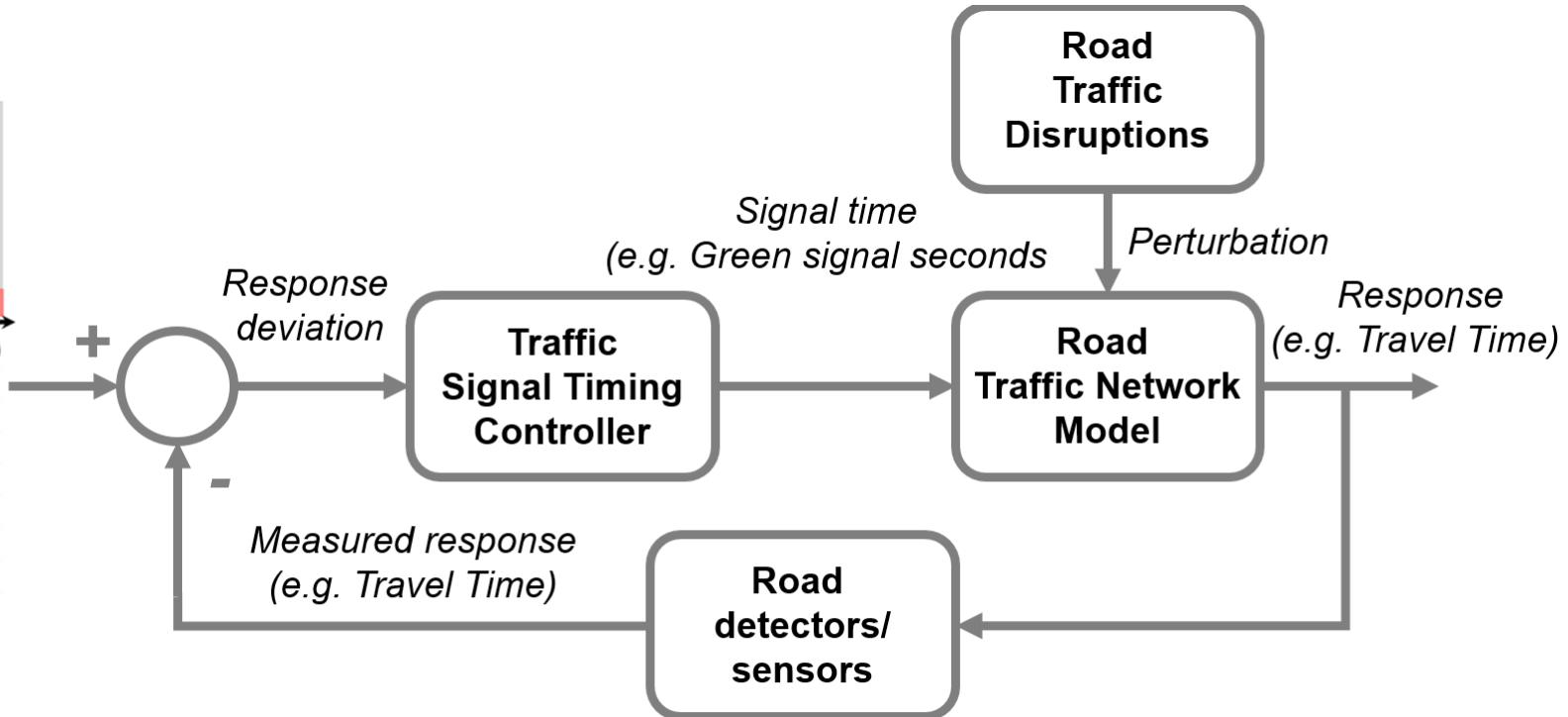
# Antifragile feedback control: transfer to technology



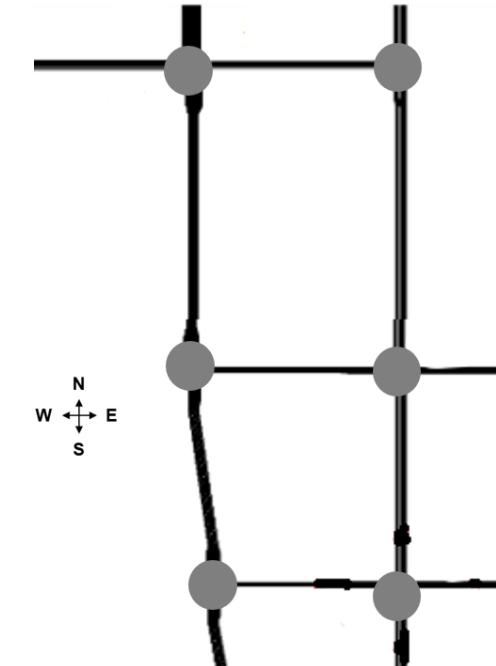
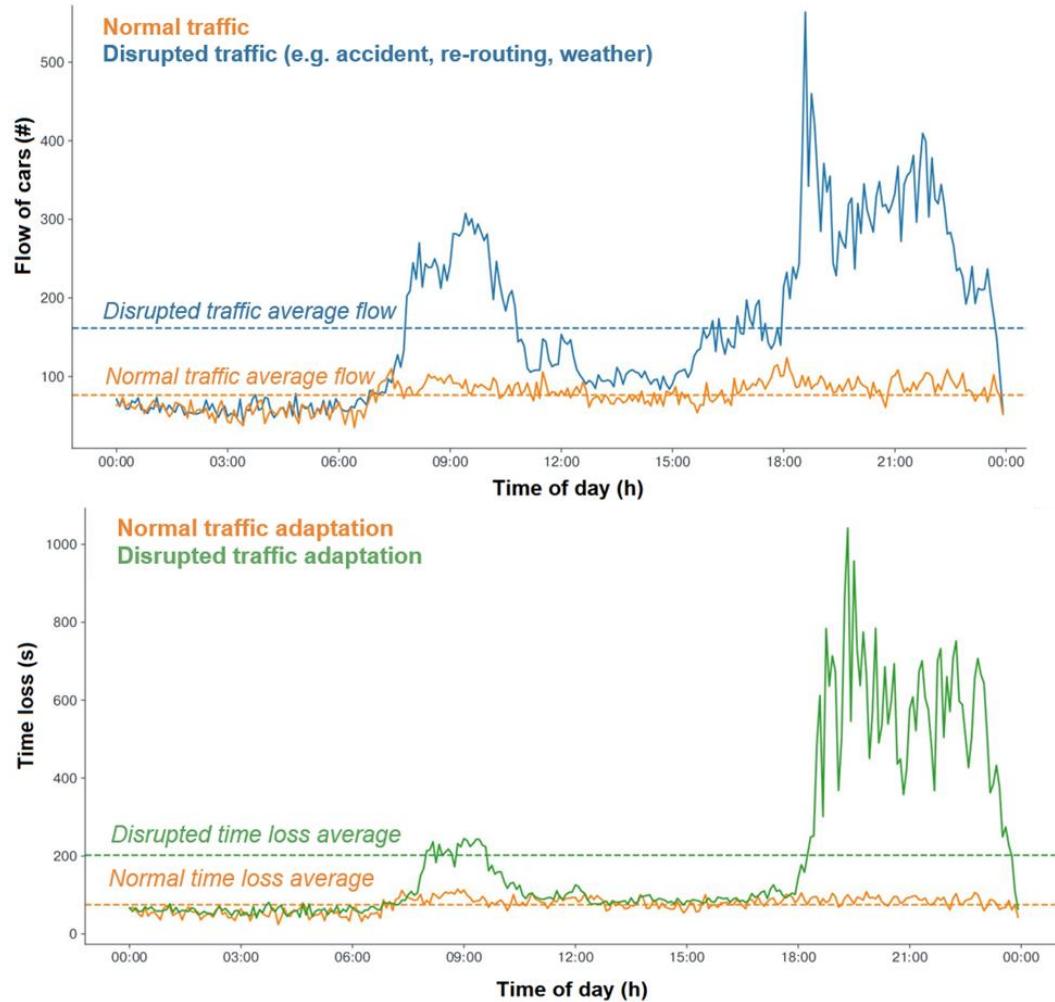
# Antifragile feedback control: transfer to technology



*Stabilization into the  
antifragile region of MFD*



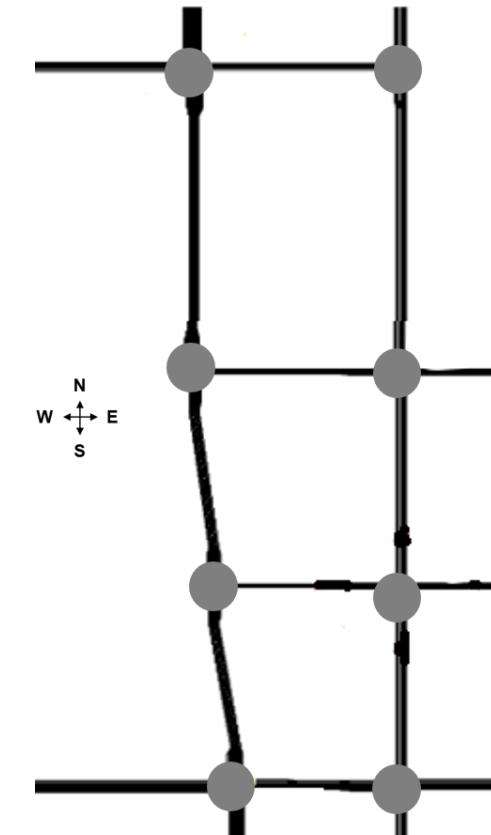
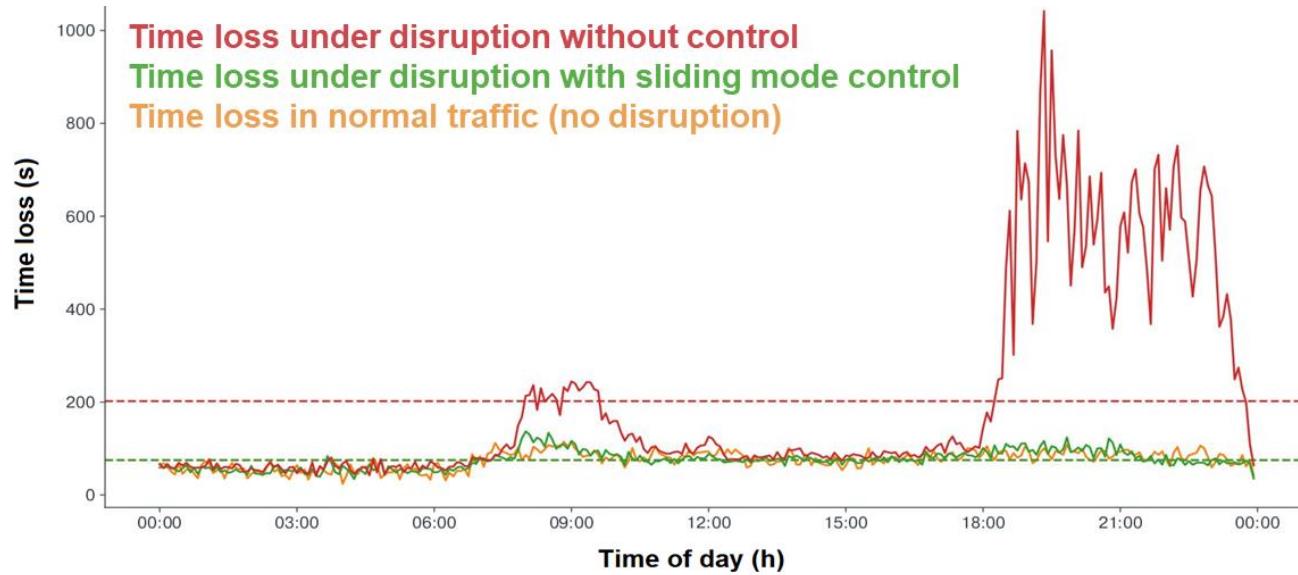
# Antifragile feedback control: transfer to technology



**Crosses**  
1 oscillator per direction  
5 crosses with 4 directions  
3 crosses with 3 directions



# Antifragile feedback control: transfer to technology



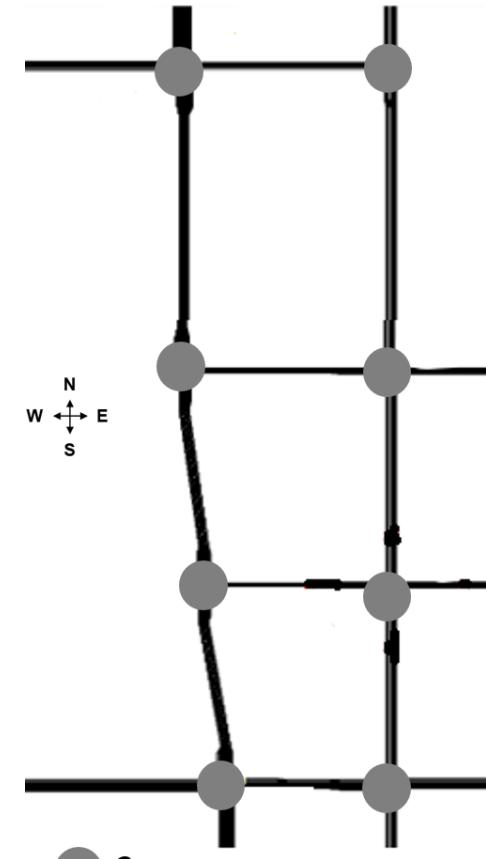
**Crosses**  
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5 crosses with 4 directions  
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# Antifragile feedback control: transfer to technology



Model	Single cross Region (8 crosses)	
MILP	0.0510	0.3930
OSCILLATOR	0.0568	0.4544
Robust OBELISC ODE	0.0489	0.4534
<b>Robust OBELISC NEF</b>	<b>0.0071</b>	<b>0.0426</b>

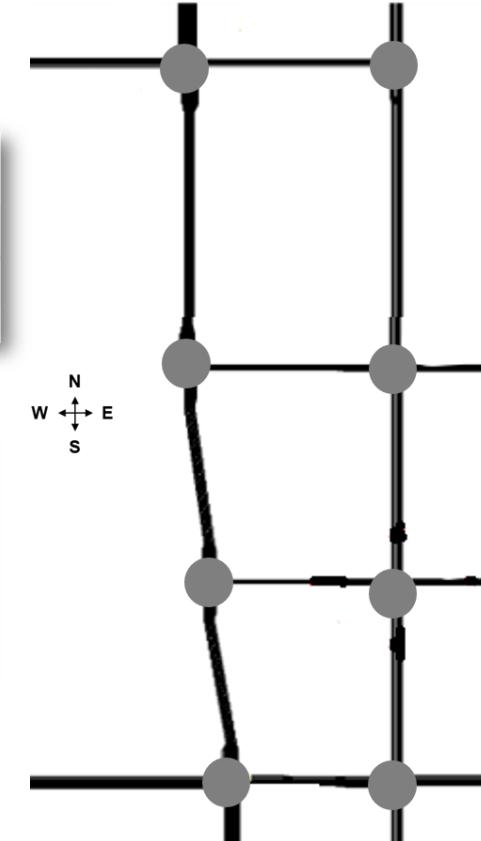
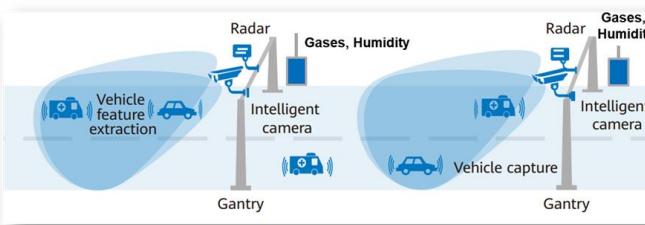
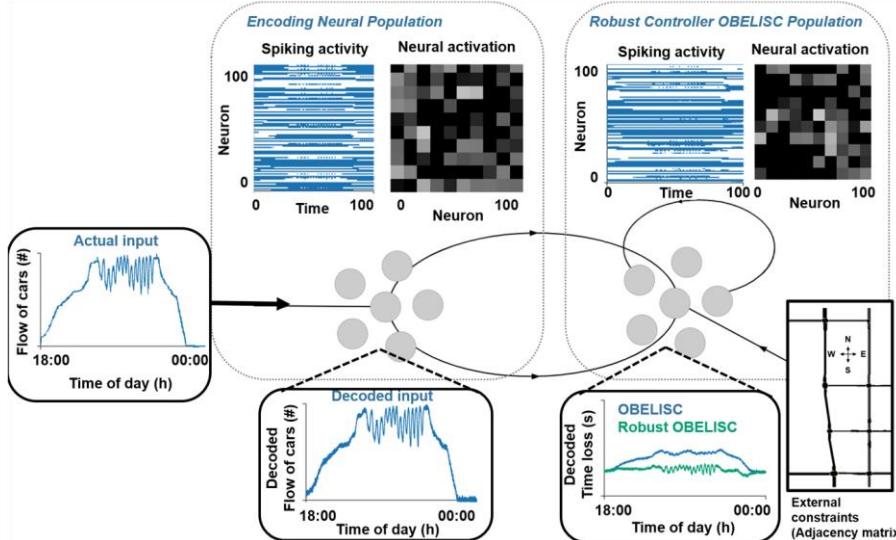


**Crosses**  
1 oscillator per direction  
5 crosses with 4 directions  
3 crosses with 3 directions



# Antifragile feedback control: transfer to technology

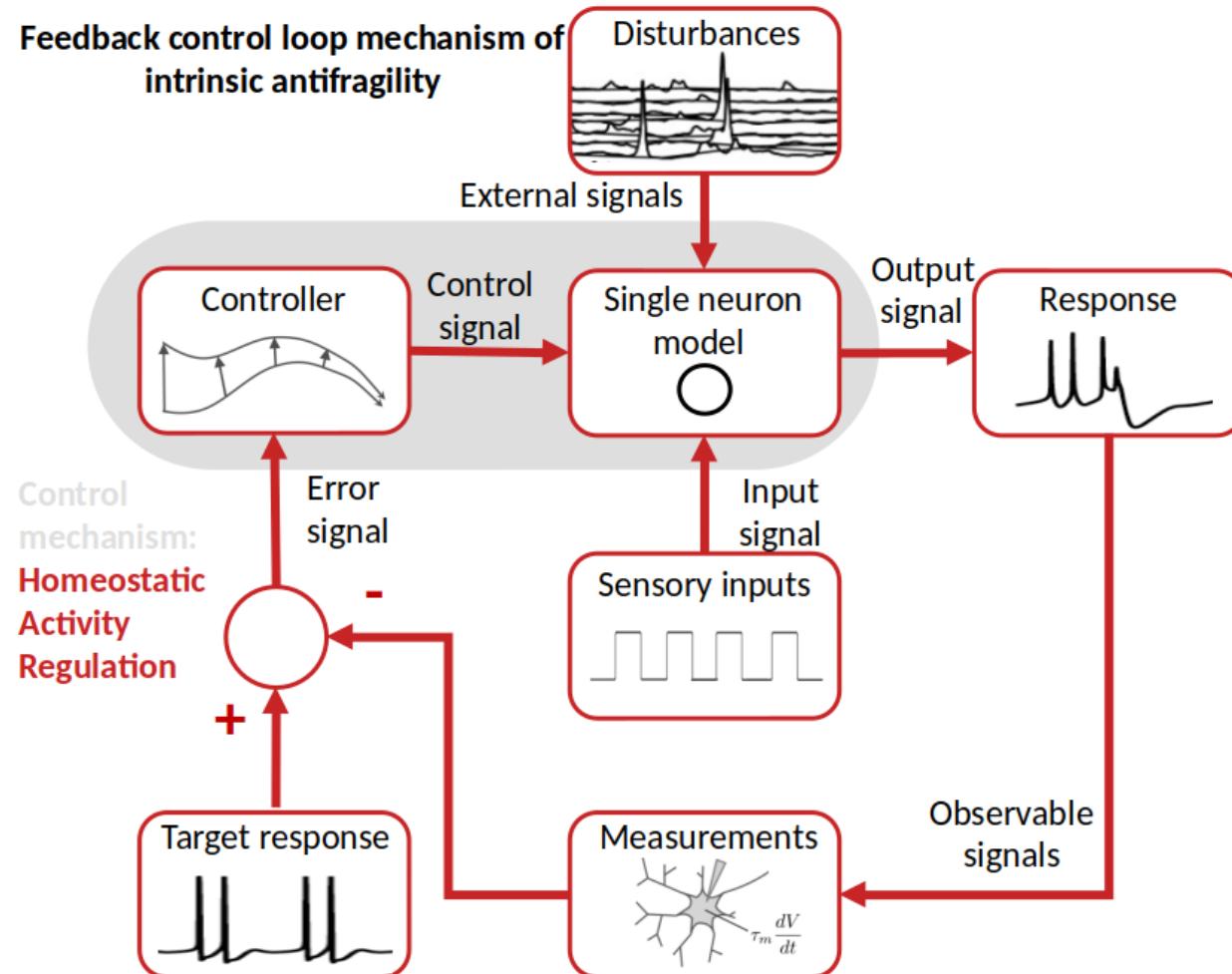
## Efficient neural learning



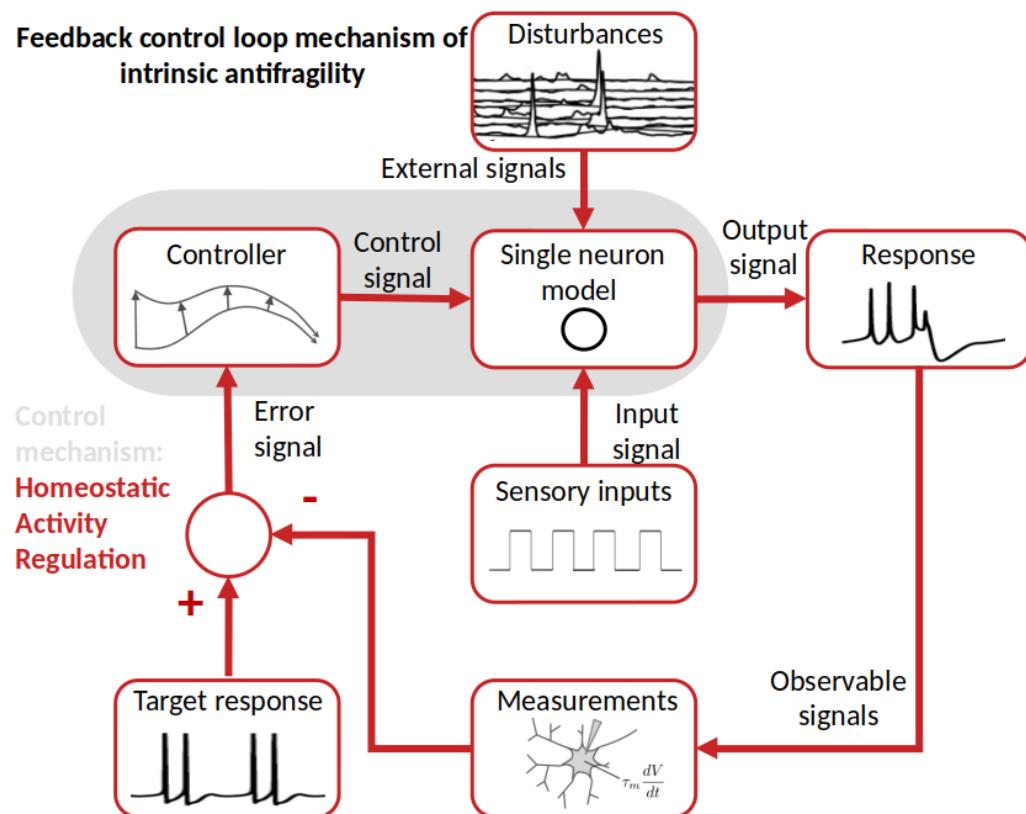
**Crosses**  
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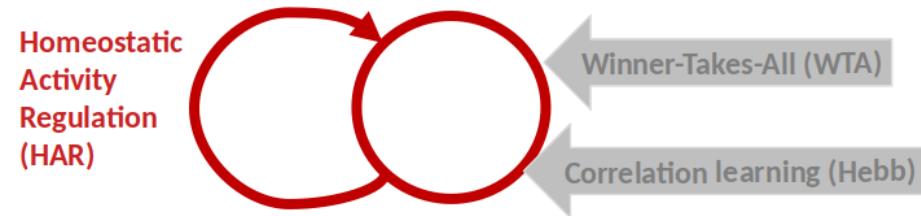
# Antifragile feedback control: learning from biology



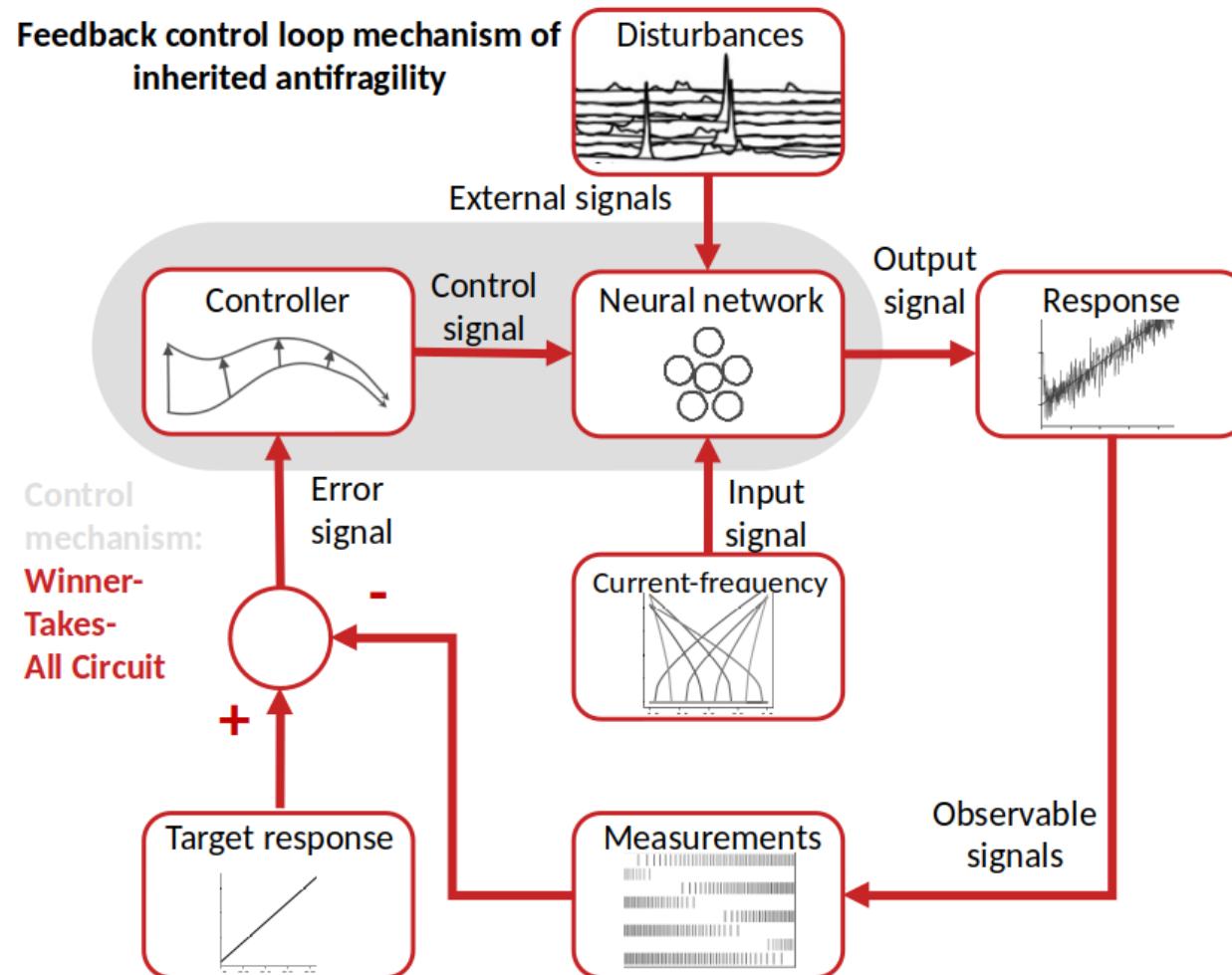
# Antifragile feedback control: learning from biology



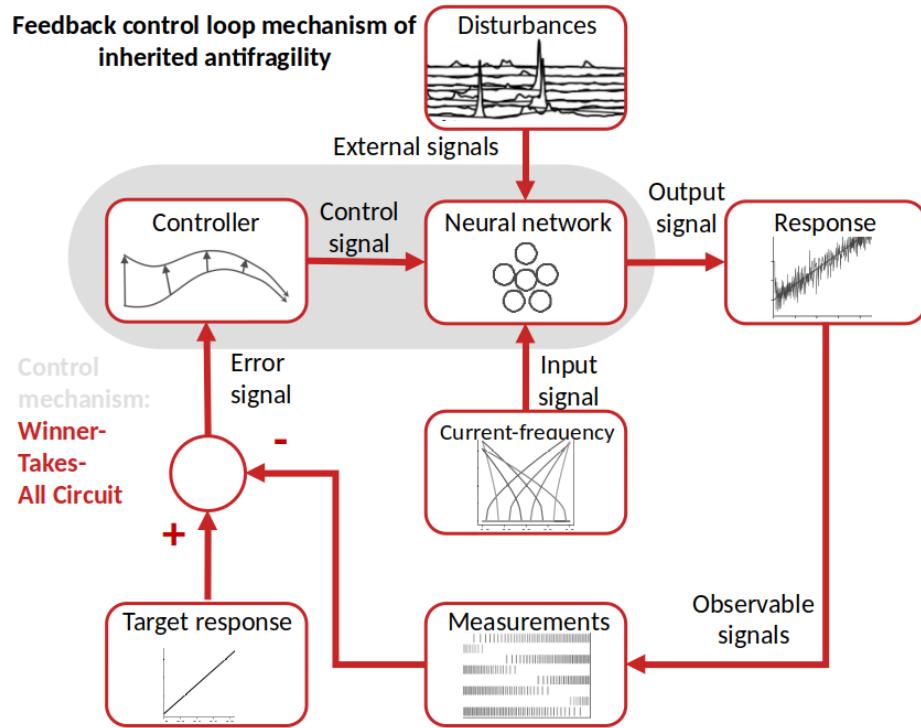
**Controller mechanism of intrinsic antifragility**



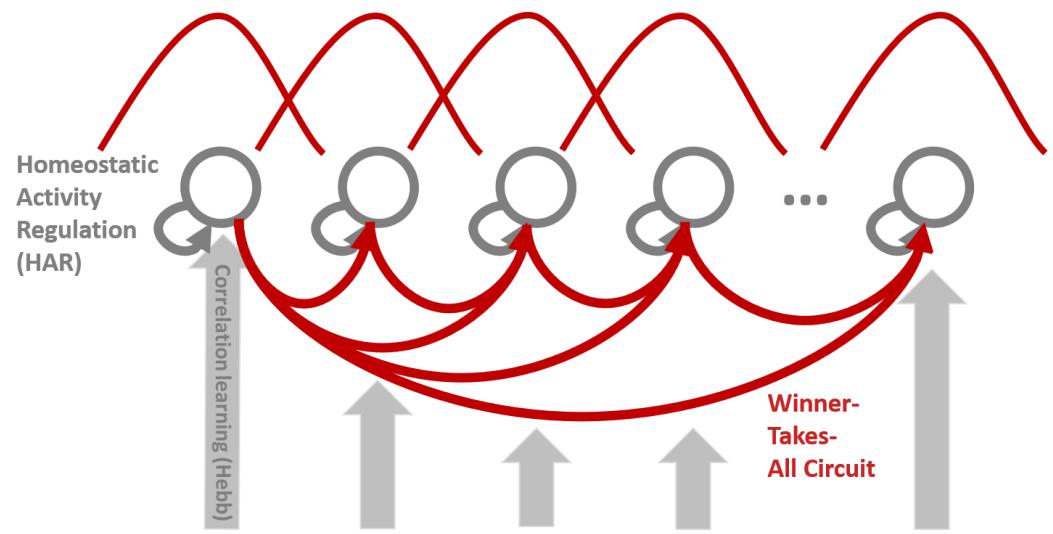
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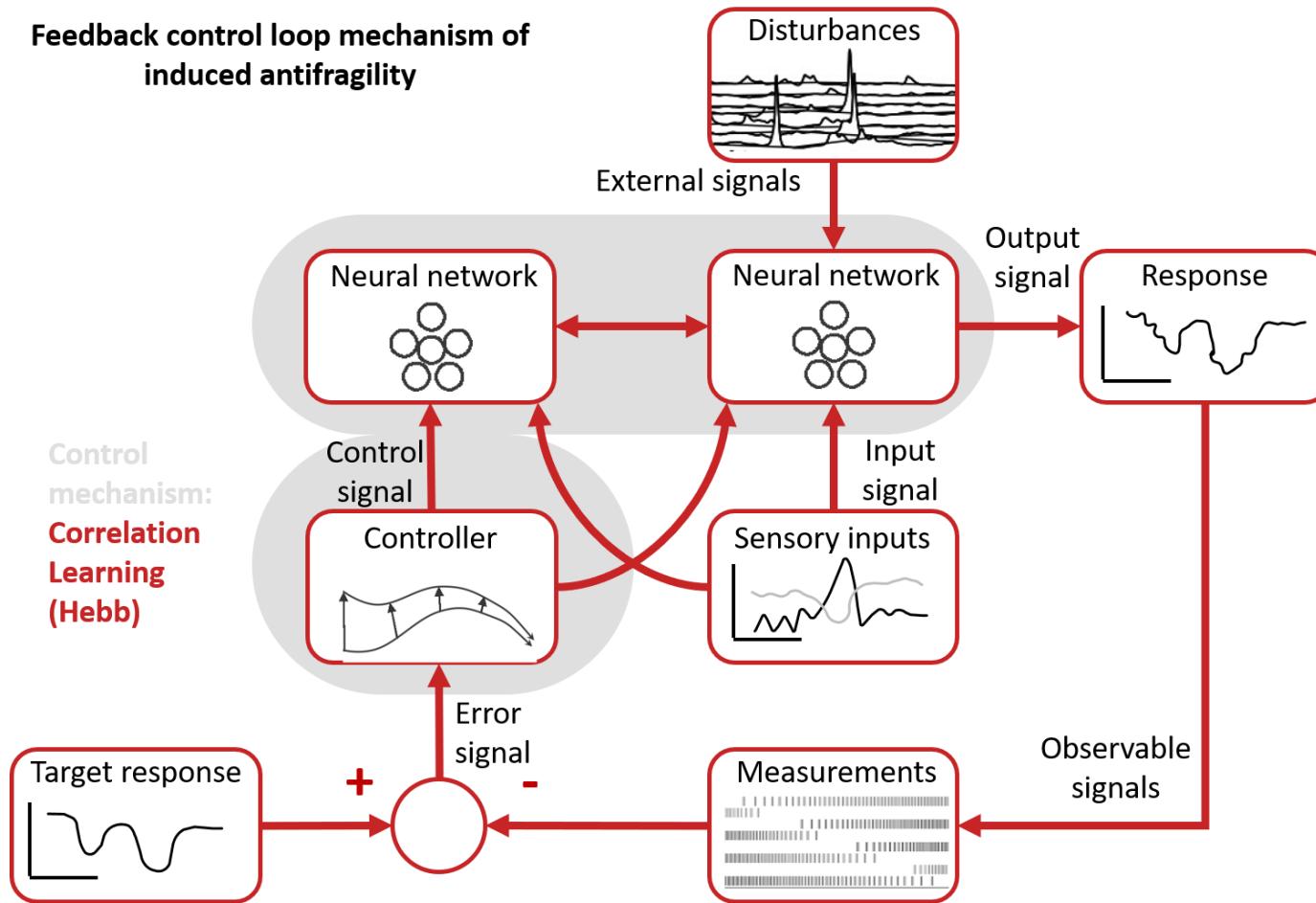
# Antifragile feedback control: learning from biology



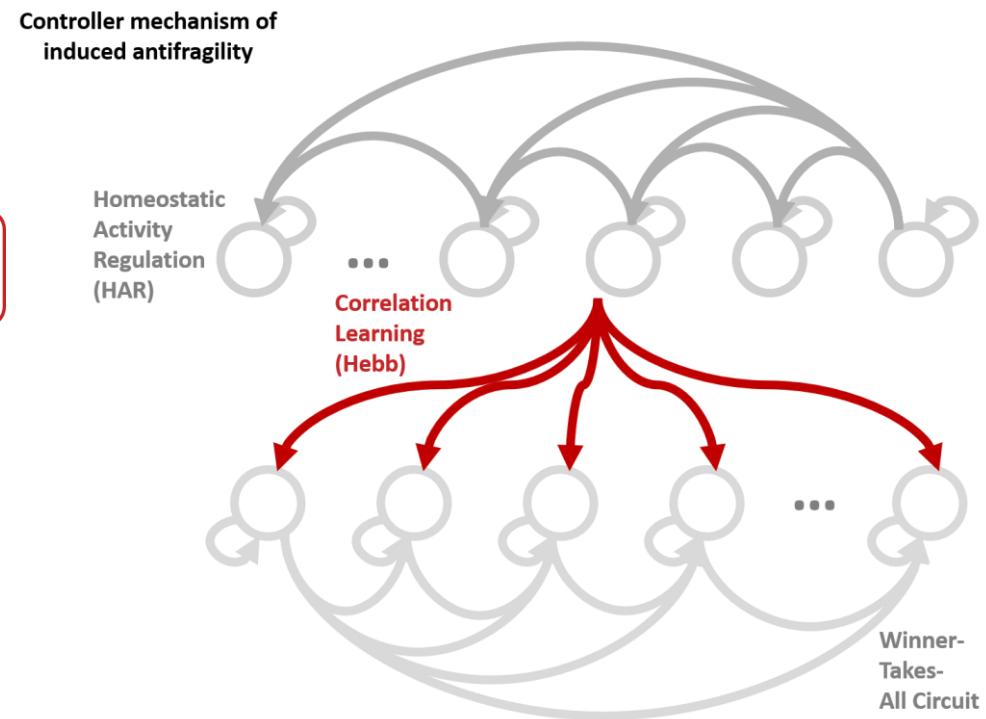
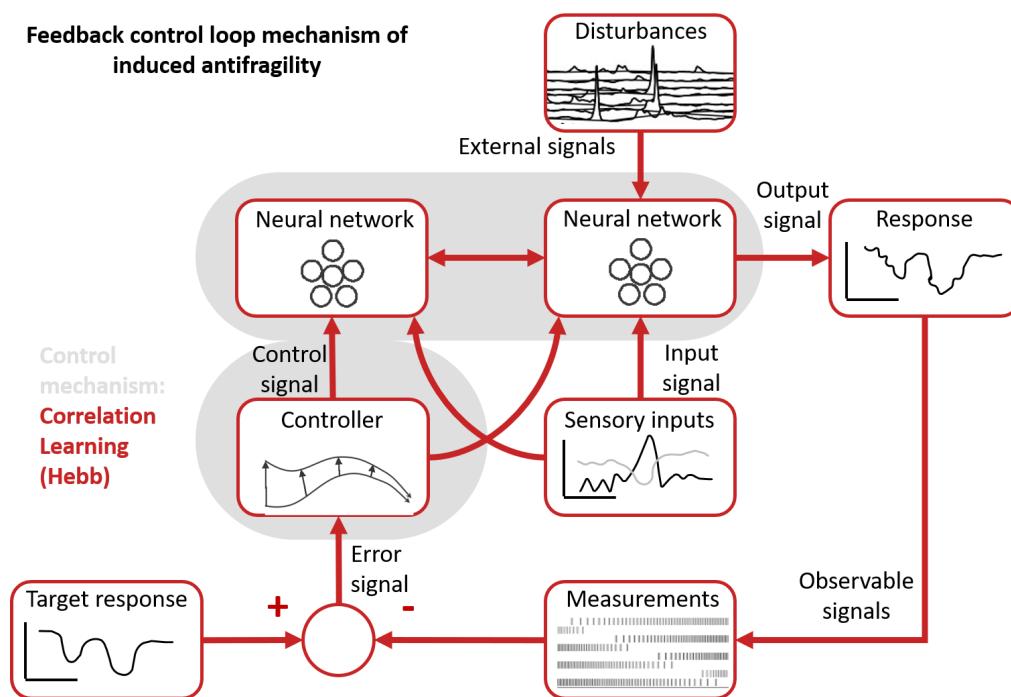
**Controller mechanism of inherited antifragility**



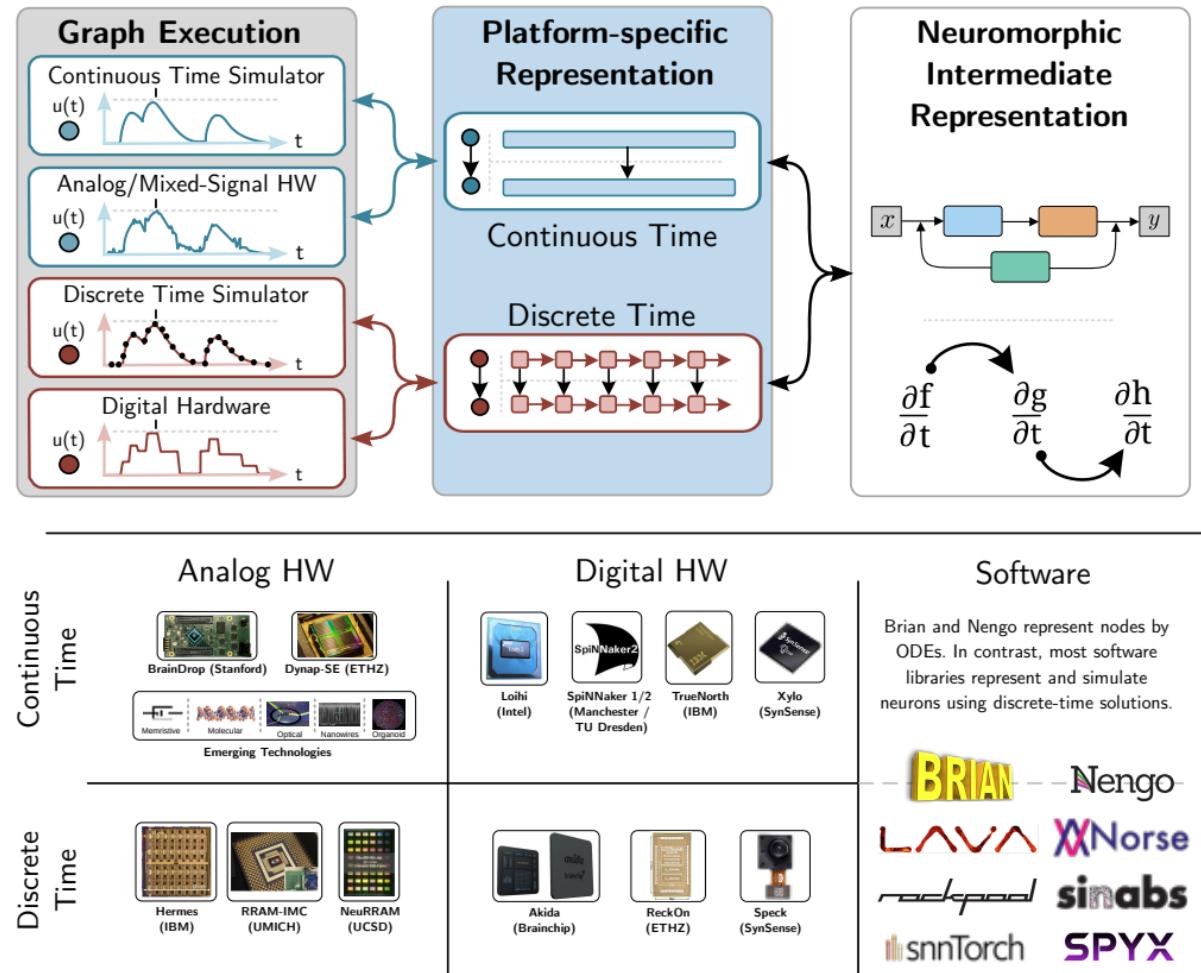
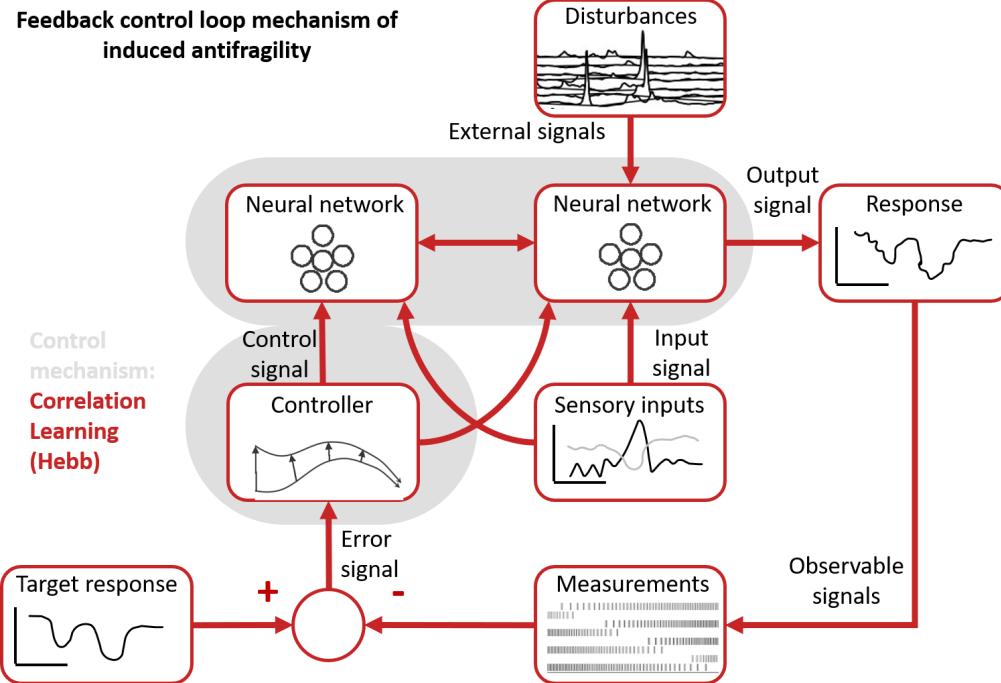
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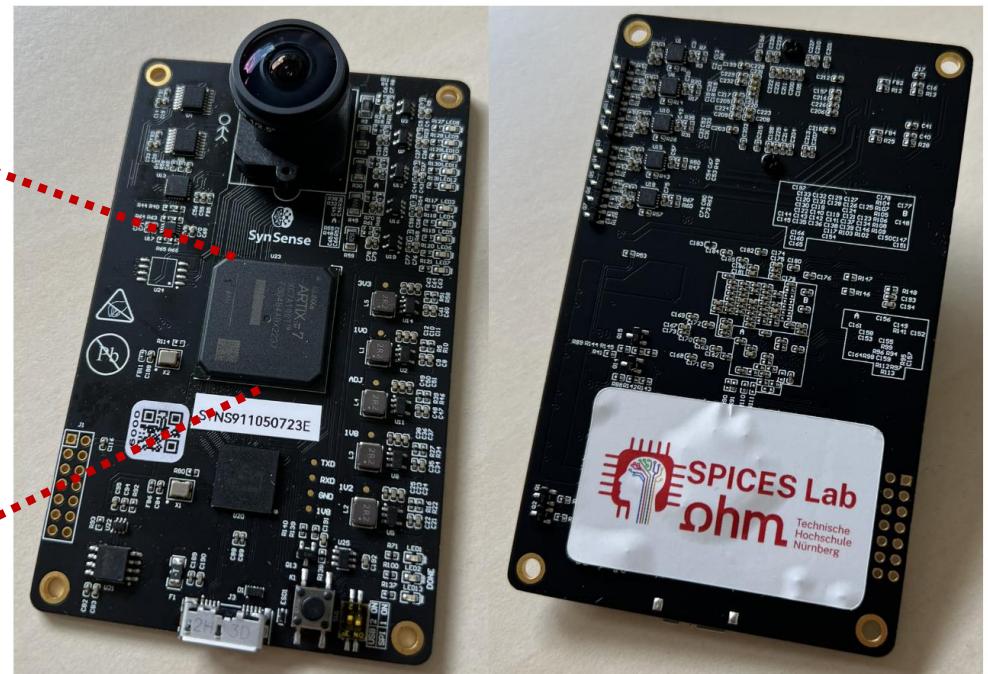
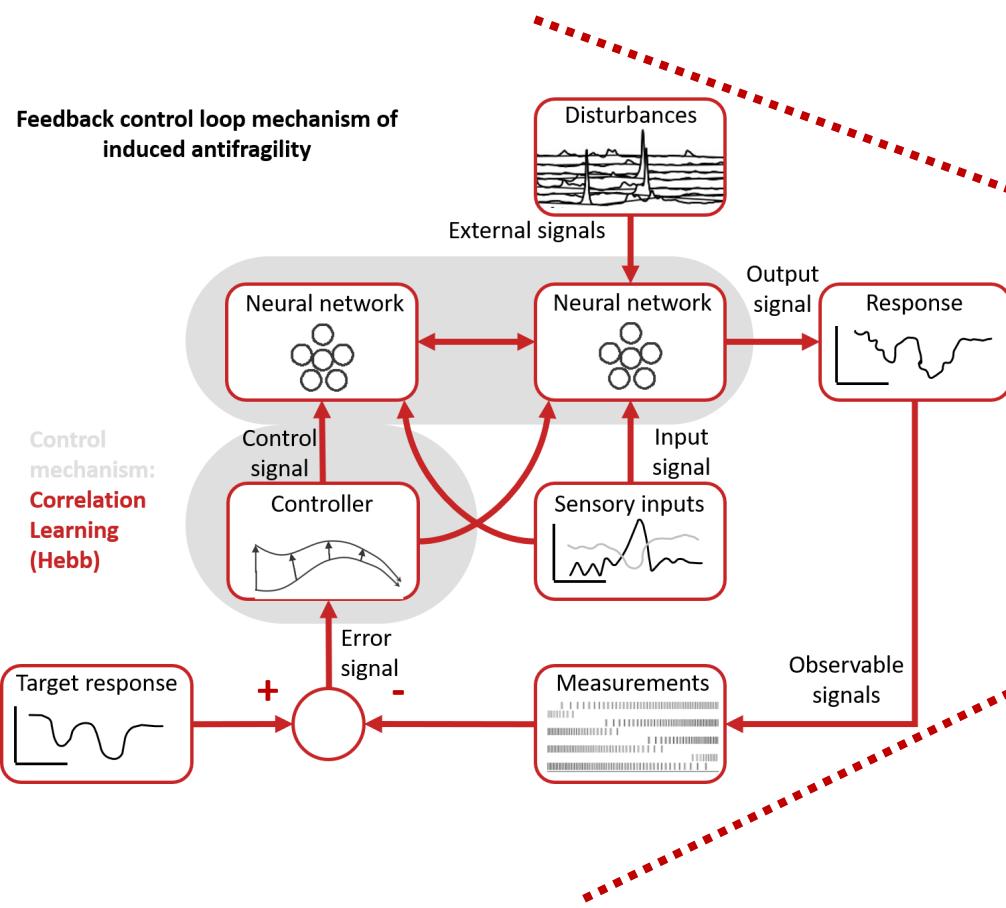
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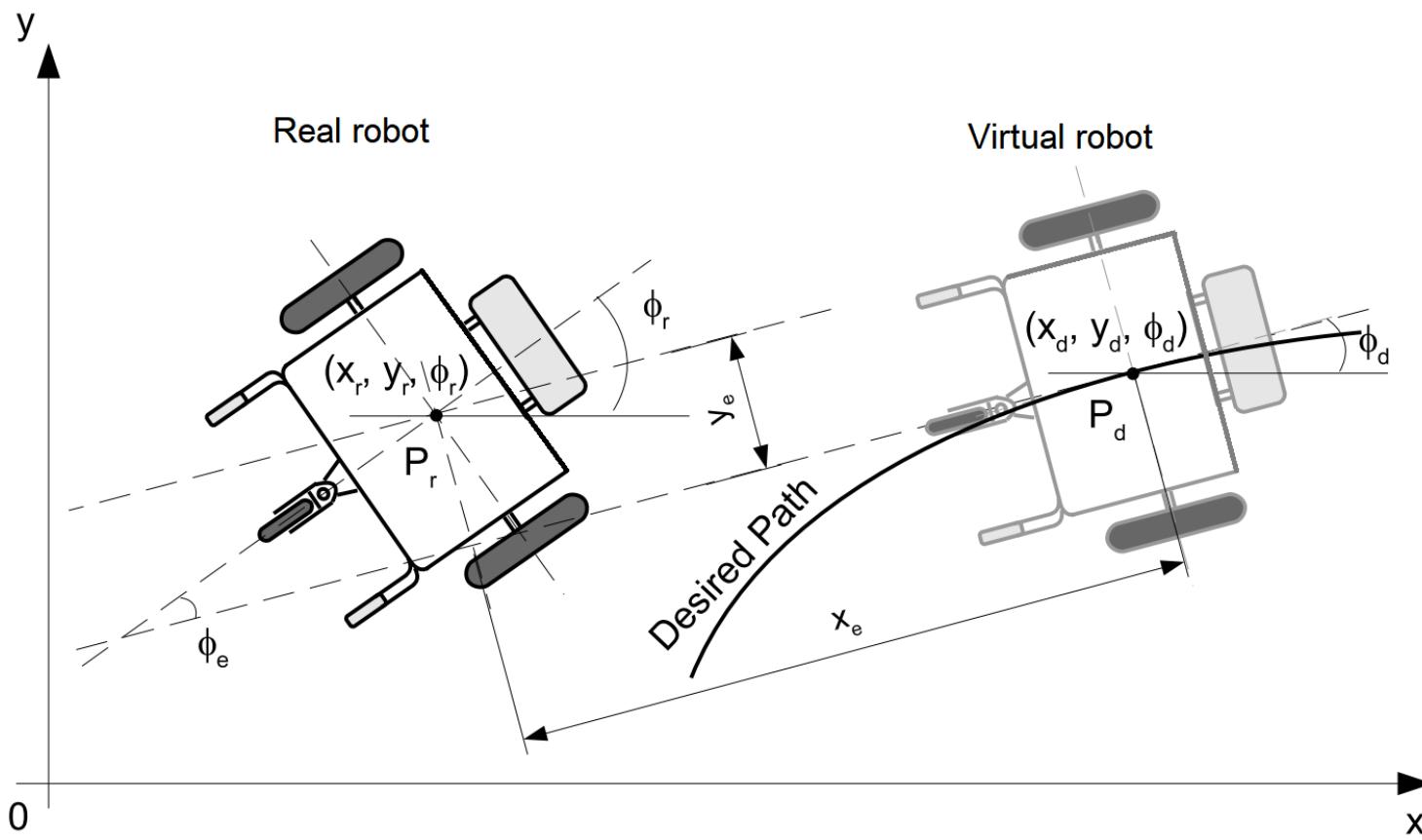
# Antifragile feedback control: transfer to technology



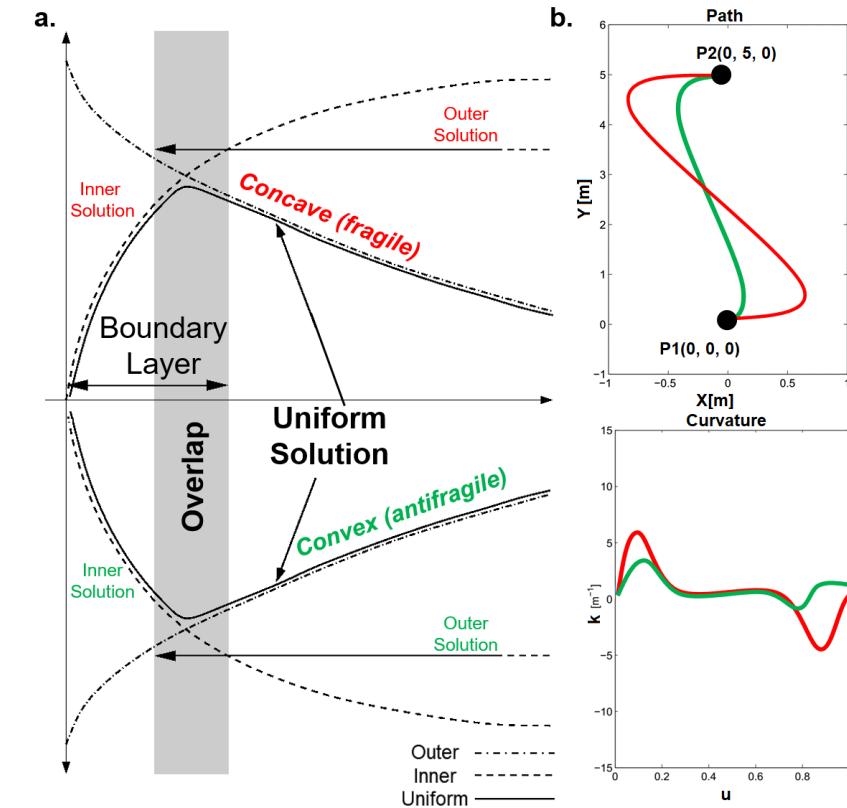
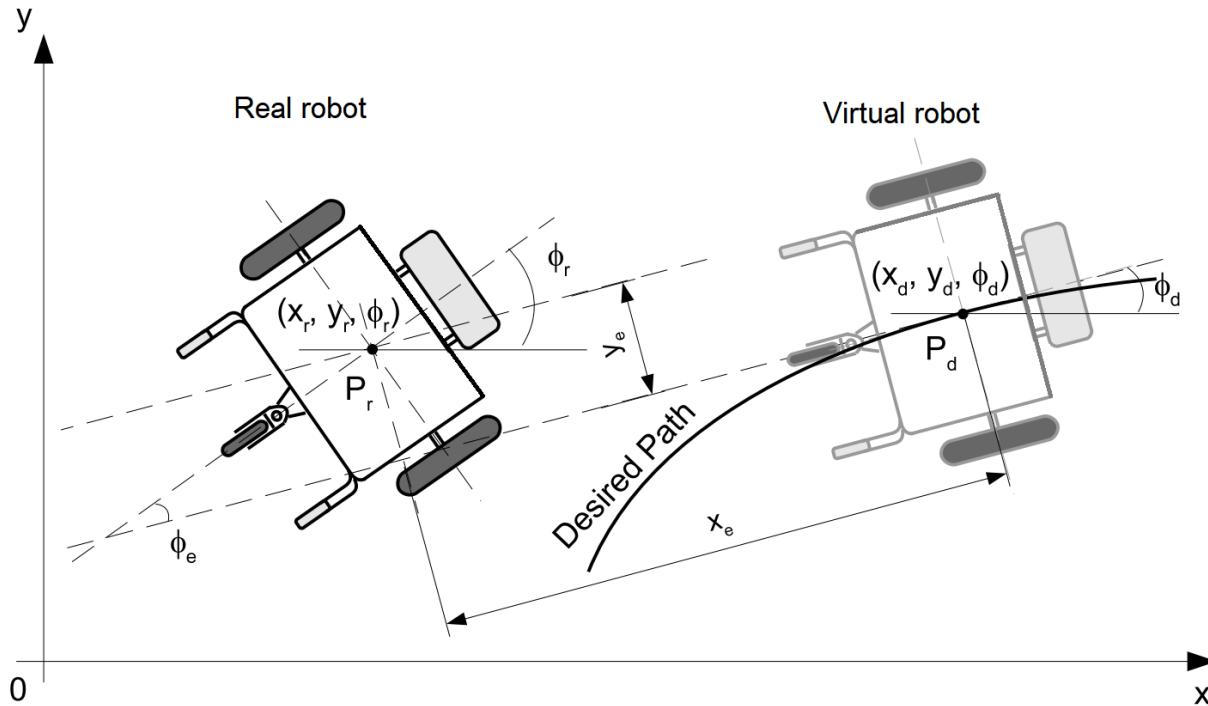
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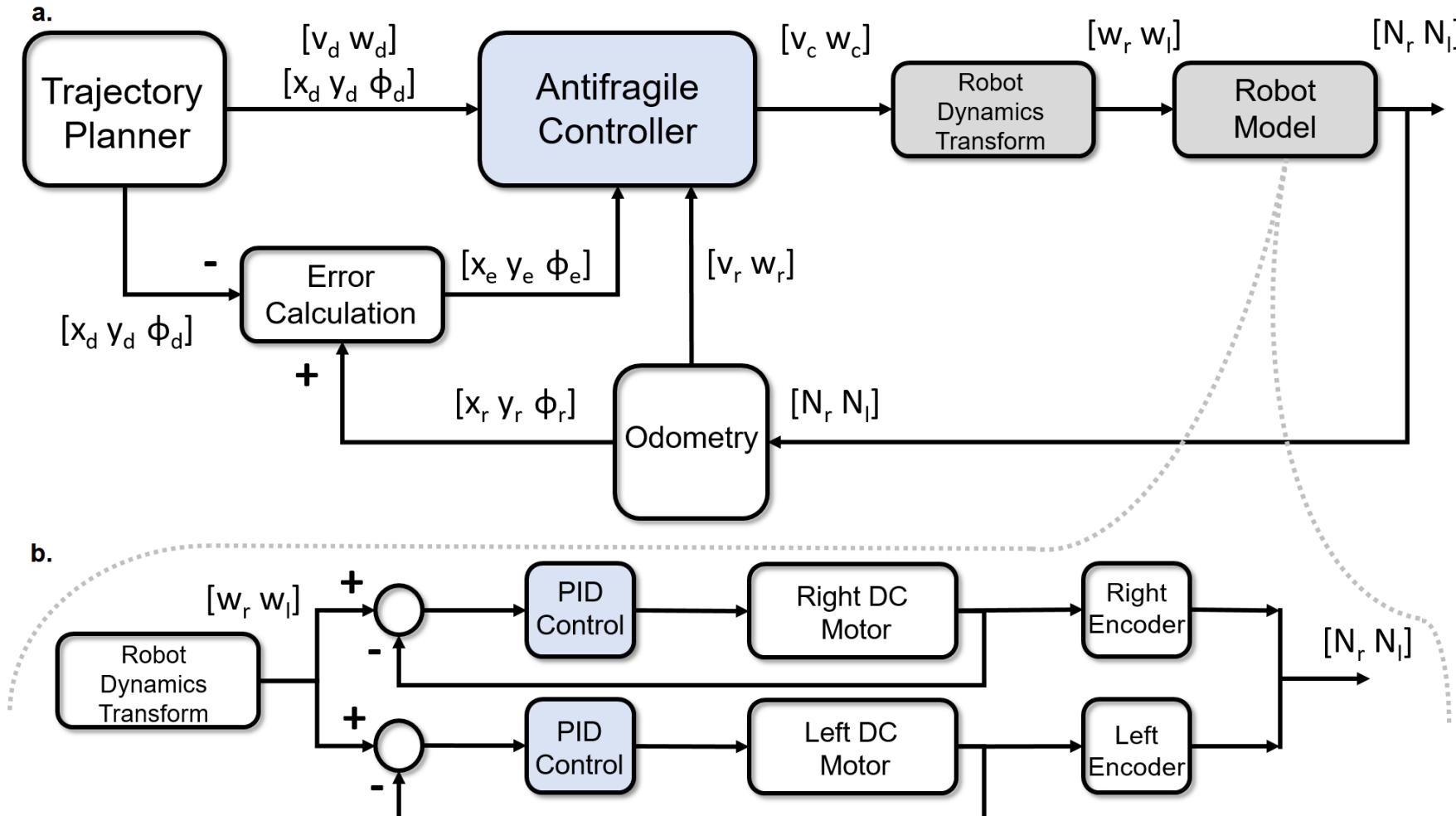
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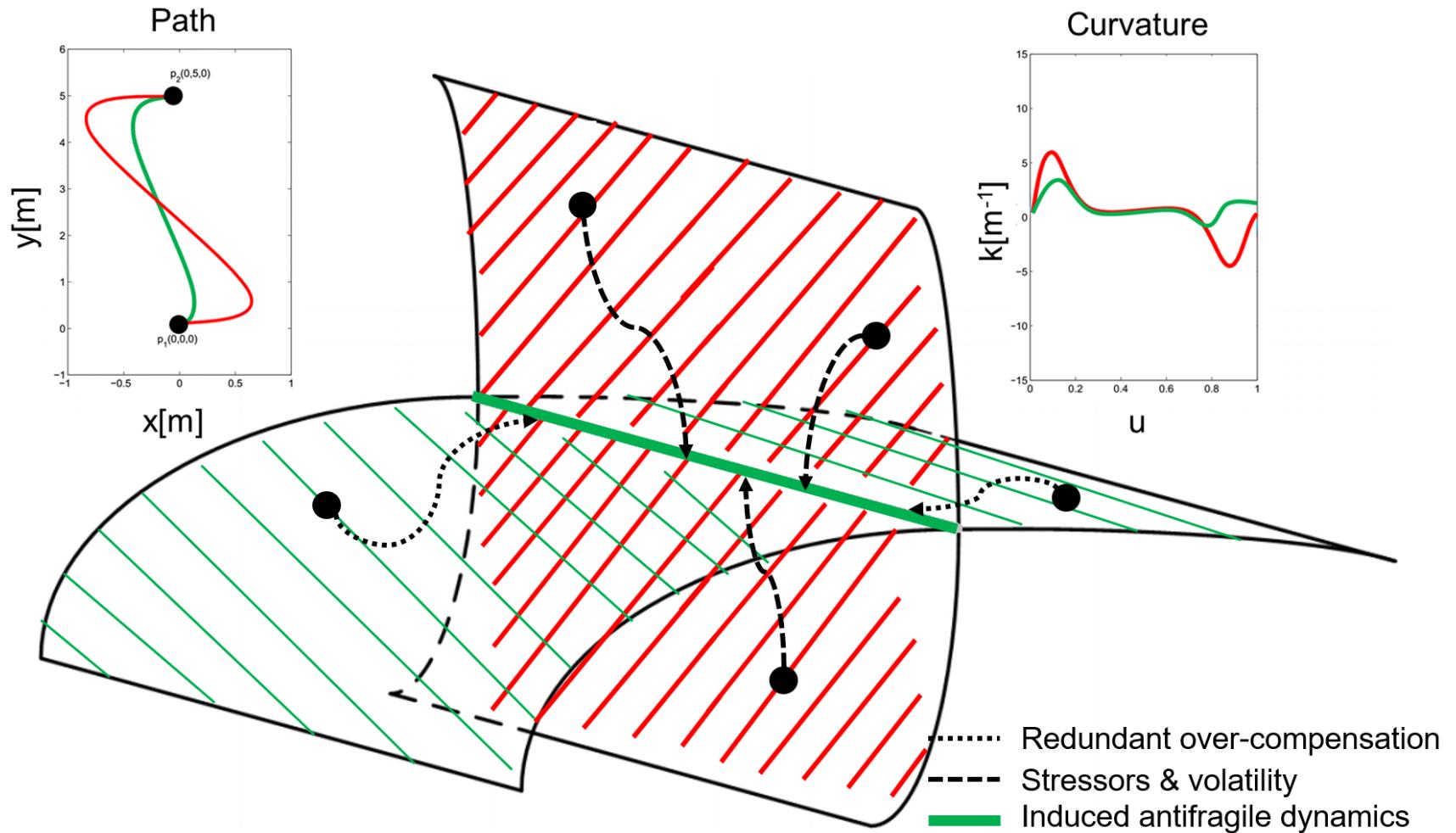
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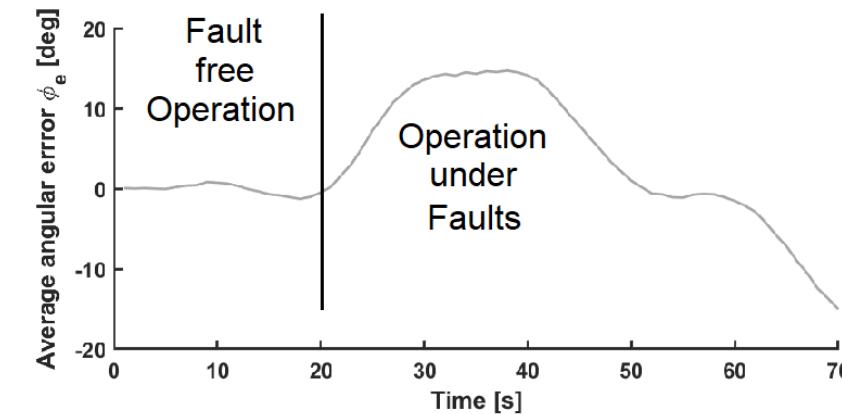
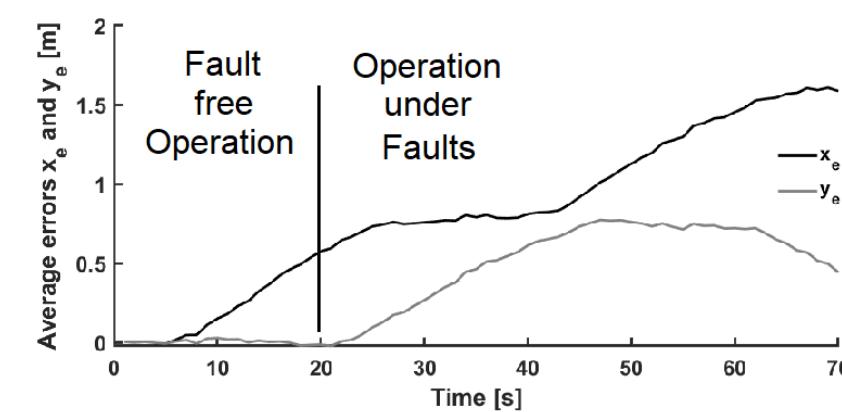
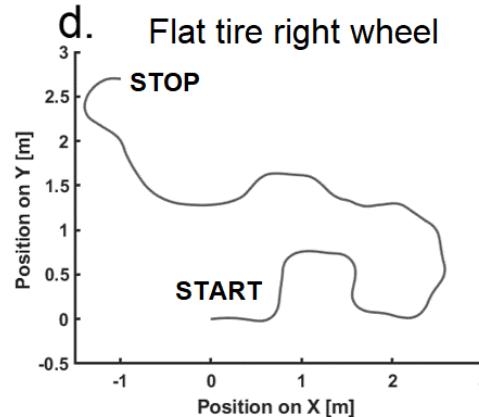
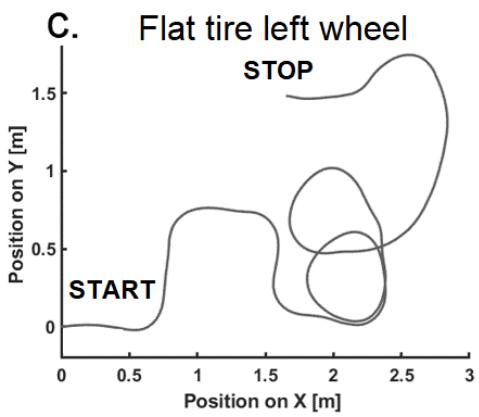
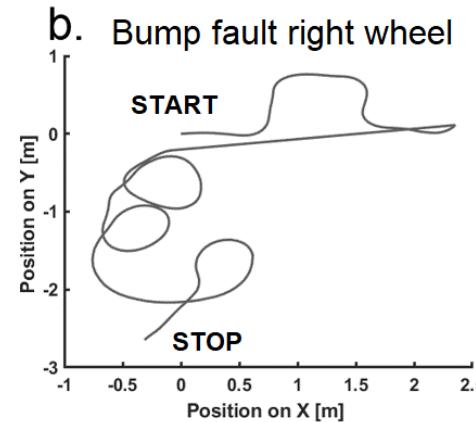
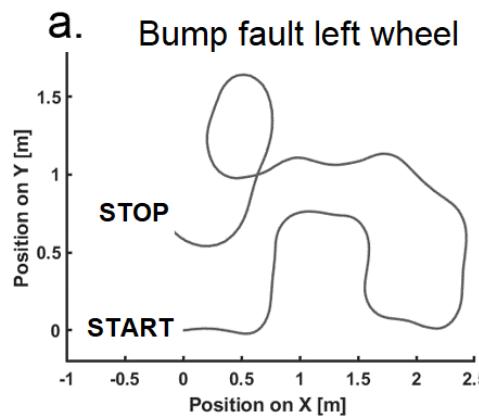
# Antifragile feedback control: transfer to technology



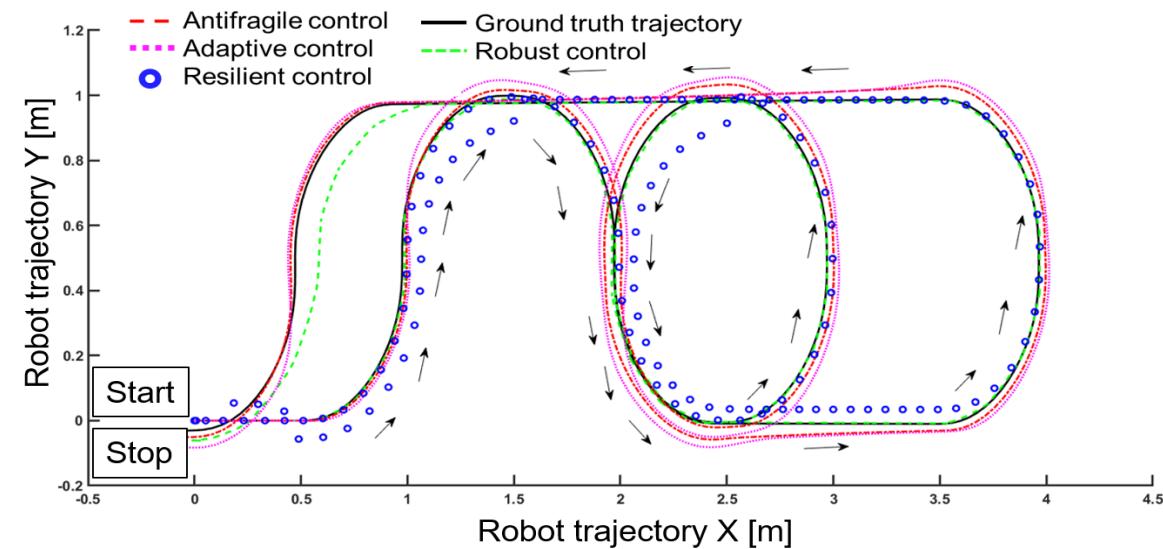
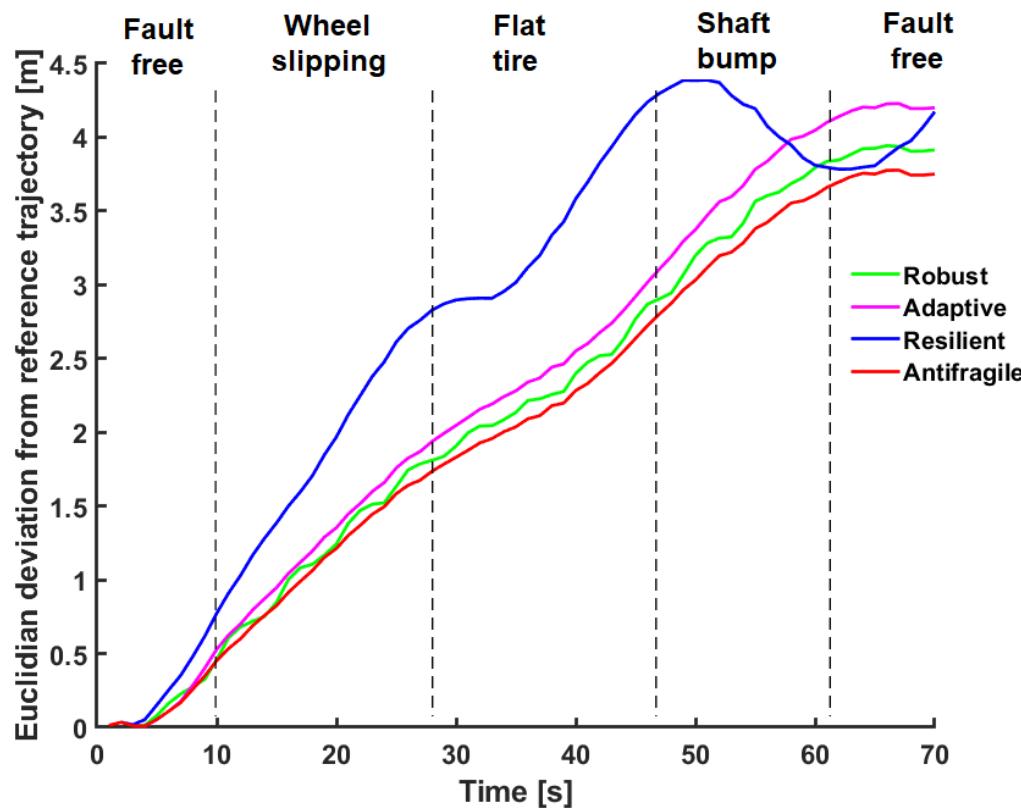
# Antifragile feedback control: transfer to technology



# Antifragile feedback control: transfer to technology



# Antifragile feedback control: transfer to technology





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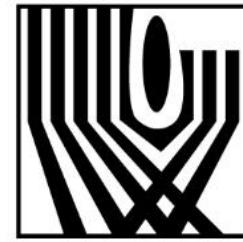
Cristian  
Axenie



Meisam  
Akbarzadeh

UNIVERSITY OF  
SURREY

Roman  
Bauer



**iimas**  
Oliver Lopez-  
Corona

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