
Practical Considerations for Generating Nonlinear Dynamic Blackbox Models using Neural Networks

A. J. Barber, Gary Sandlass
MTS Systems Corporation



Contents

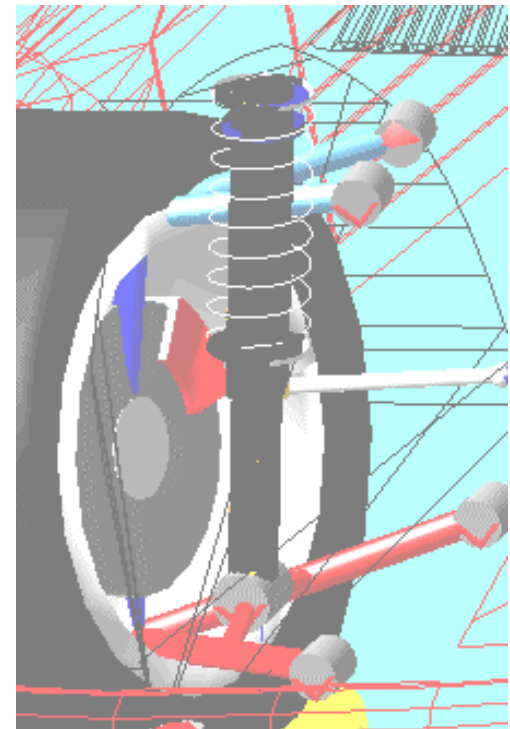
- **Review of Empirical Dynamics Modeling**
 - Status of Adams/EDM Interface development
- Choice of Model Input
 - Displacement
 - Velocity
- Representation of Inertial Forces
- Choice of Blackbox Boundaries

Empirical Dynamics Modeling

- **Blackbox**
- **Nonlinear, Amplitude-dependent**
- **Dynamic, Frequency-dependent**
 - Exhibits Hysteresis or Memory

Blackbox Models

- Obtained from experimental data
- Inputs & outputs alone
- Little or no physical detail

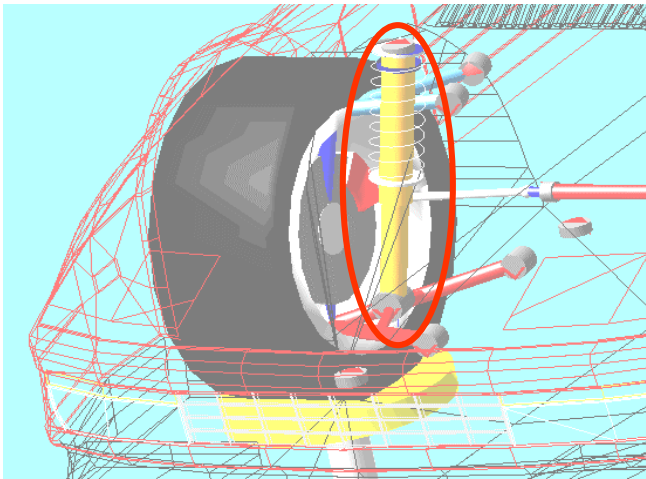
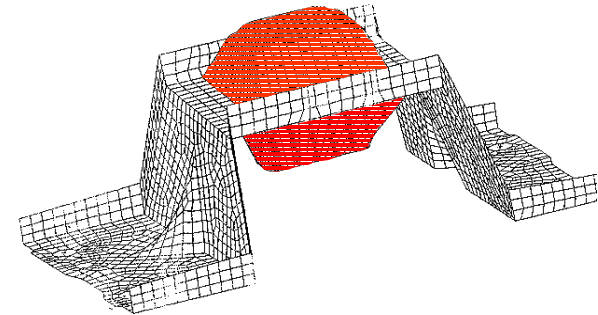


Whitebox Modeling

- **Uses information about the system**
 - Geometry
 - Physical properties
 - Physical laws
 - **Uses differential equations**
 - **Examples: ADAMS, FEM, CFD**
-

Why is blackbox modeling needed ?

- Too many variables
- Slow to solve

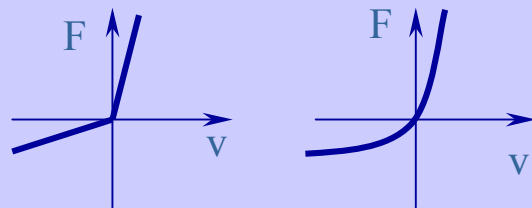


- Complex physics or math
- Expertise unavailable

Conventional Blackbox Methods

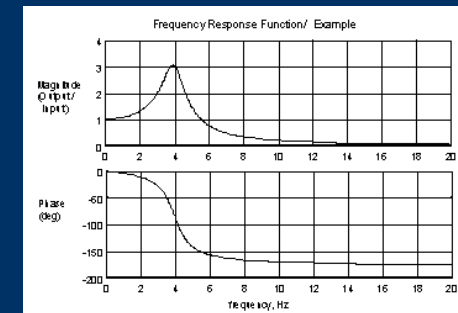
Amplitude Dependent Models

- Splines
- Polynomials



Frequency Dependent Models

- Frequency Response Functions (FRFs)



- ARMA Models

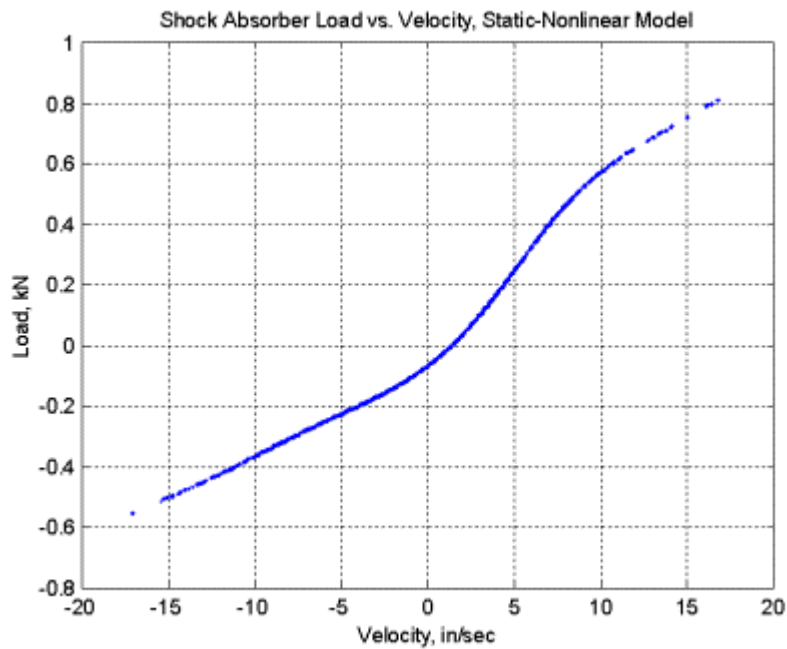
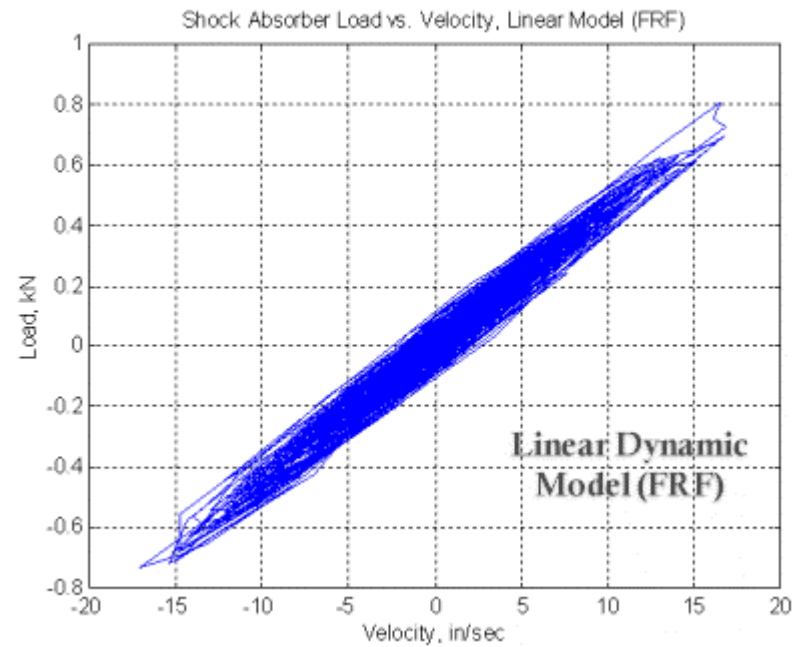
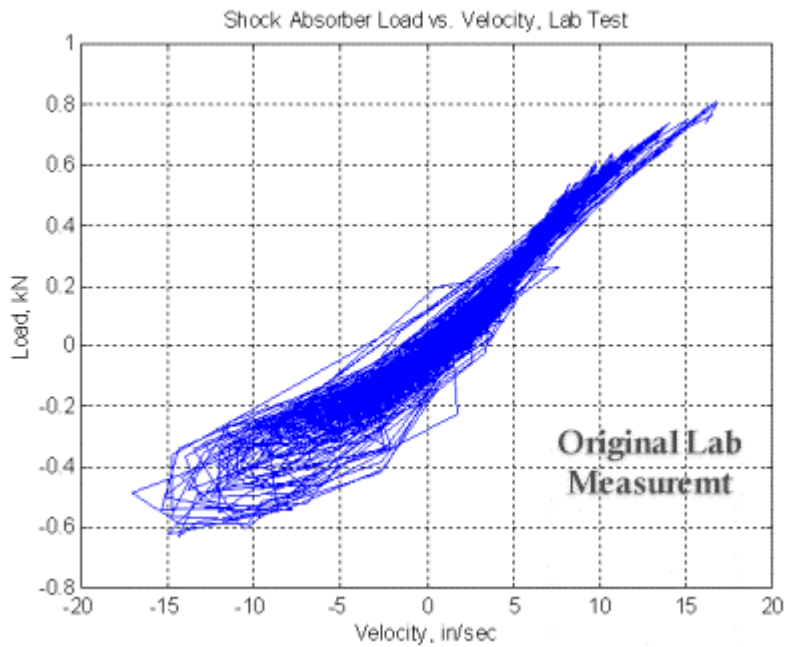
Conventional Blackbox Methods

Amplitude Dependent Models

- nonlinear systems
- without hysteresis
- small number of inputs

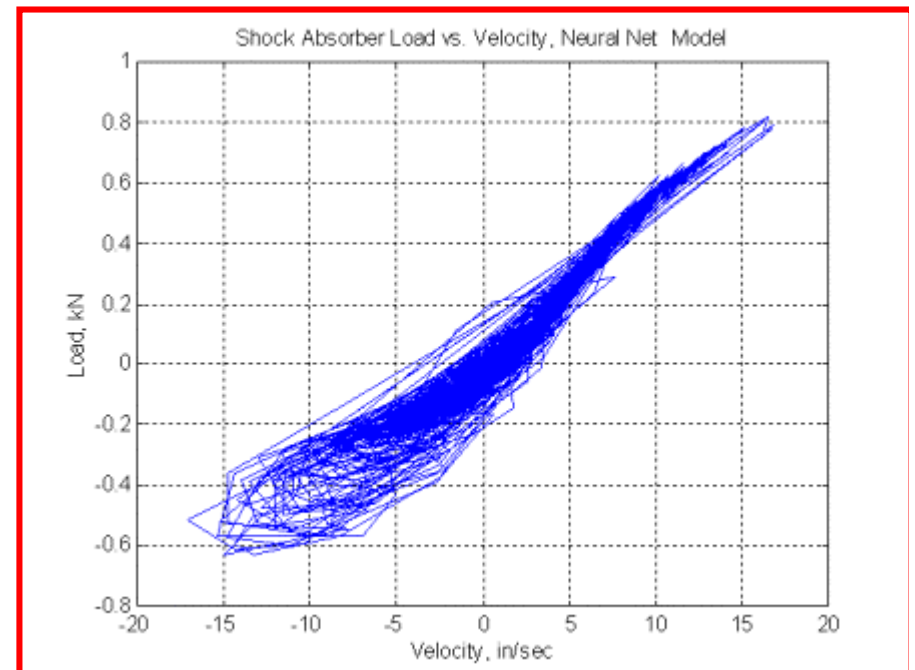
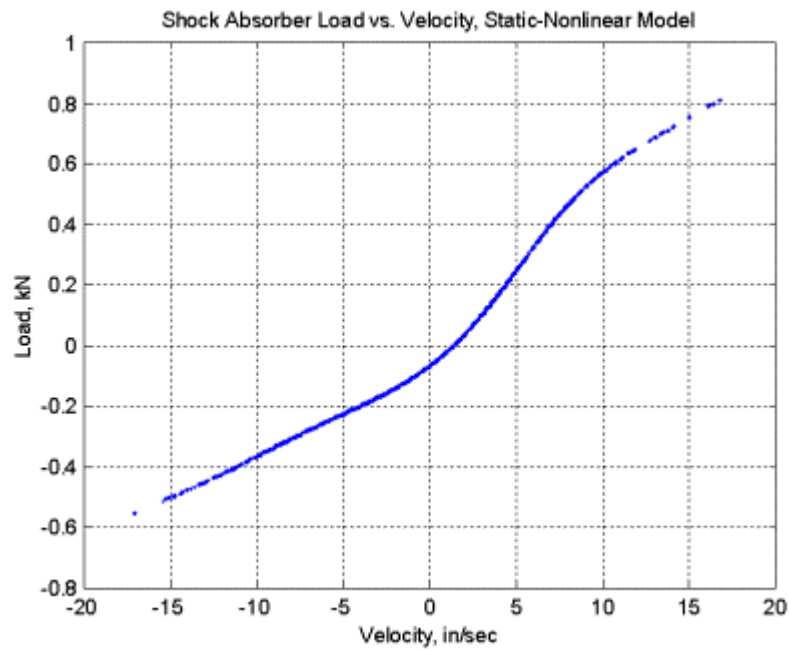
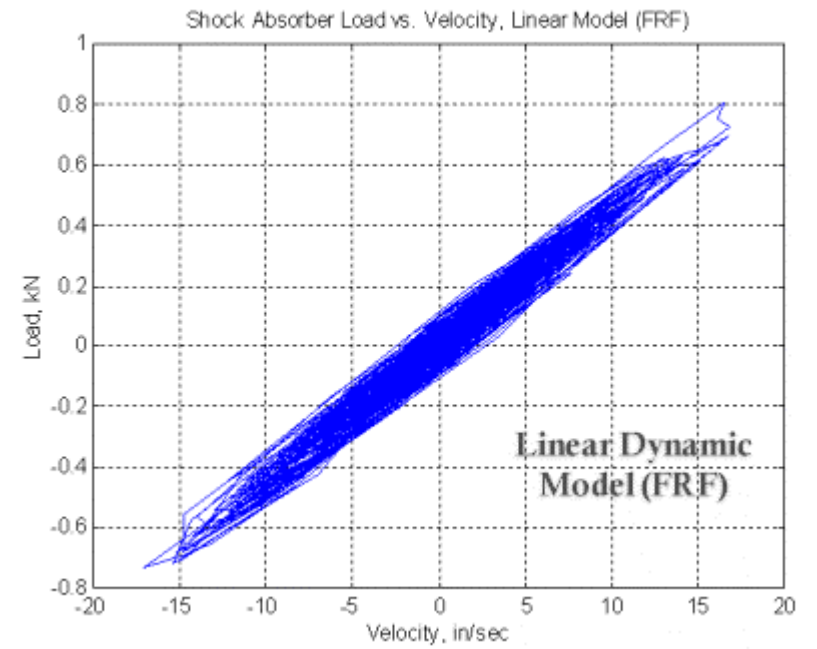
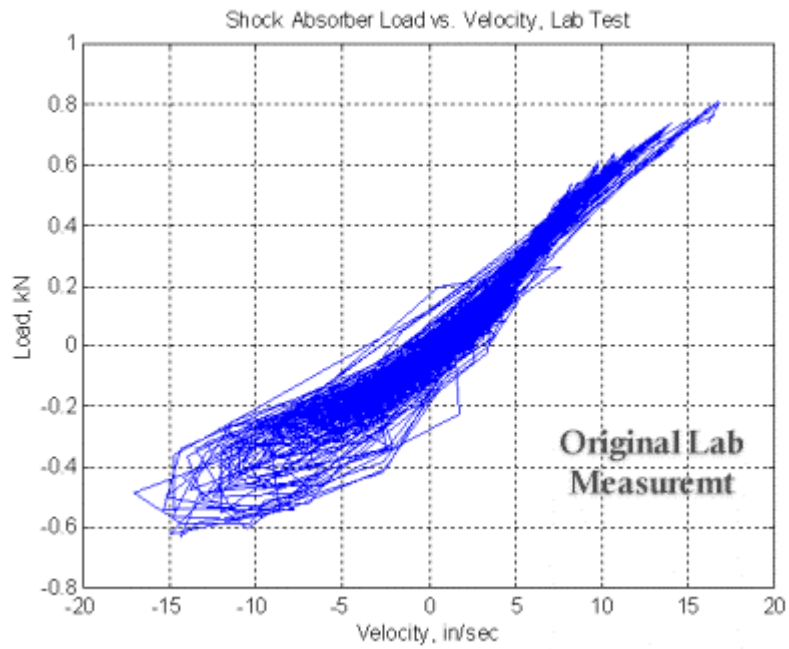
Frequency Dependent Models

- linear systems
- with hysteresis
- many inputs



Mathematical Complexity

CPU Power



Empirical Dynamics Modeling

- **Blackbox**
- **Amplitude-dependent**
- **Frequency-dependent**

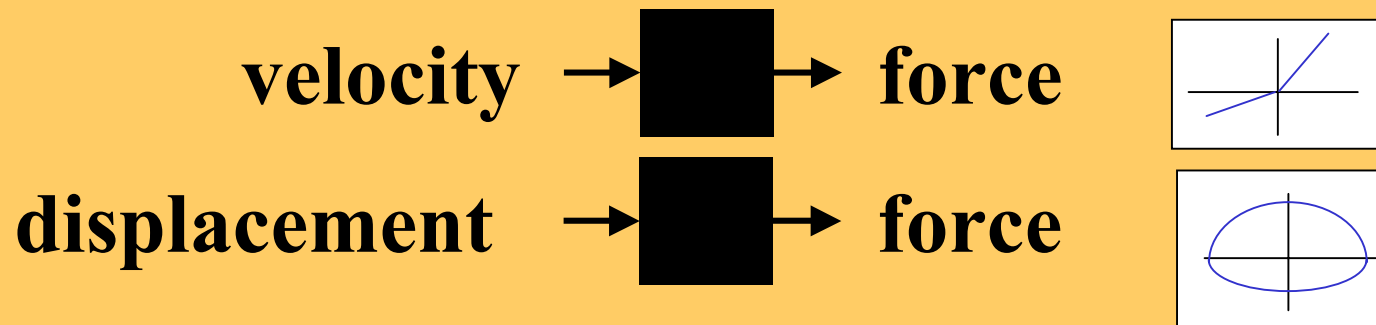
- **Random Waveform**
- **Multiple Input and Output**
- **Wide Class of Systems**

Practical Considerations in Nonlinear Blackbox Modeling

- Review of Empirical Dynamics Modeling
- **Choice of Model Input**
 - Displacement
 - Velocity
- Representation of Inertial Forces
- Choice of Blackbox Boundaries

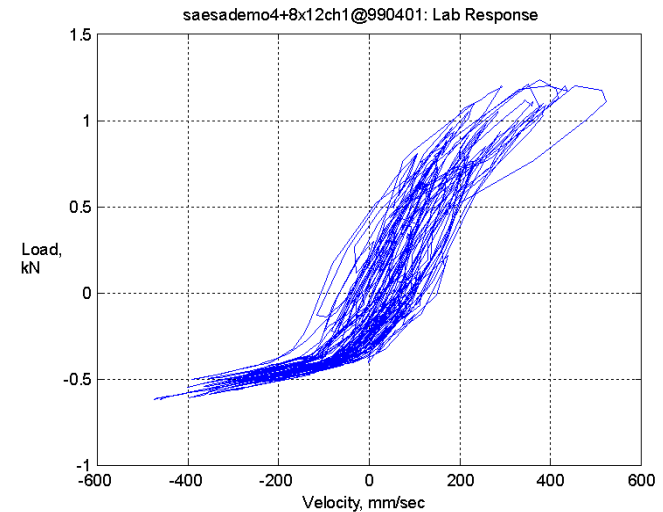
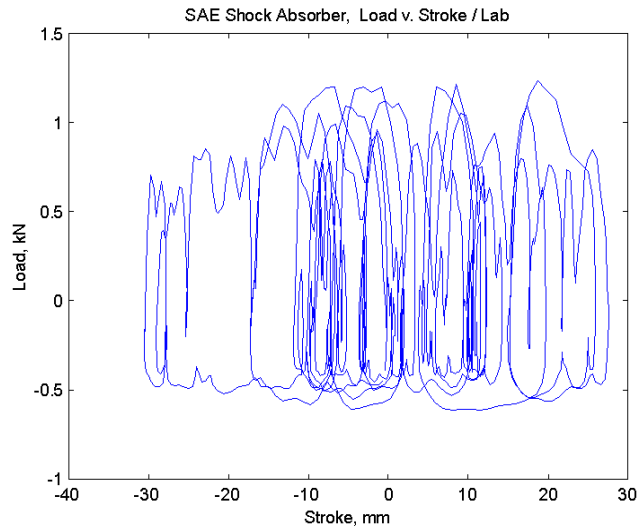
Choice of Model Input

- Even a simple damper can be modeled several ways as a blackbox, by using different input and output signals:



Choice of Model Input

- **Damper: Displacement v. Velocity Input**



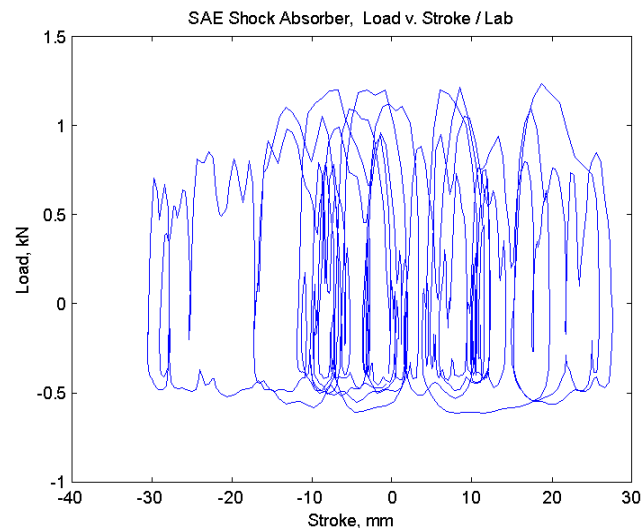
Choice of Model Input

- Empirical Dynamics *damp*er models use displacement input, force output:

displacement → EDM → force

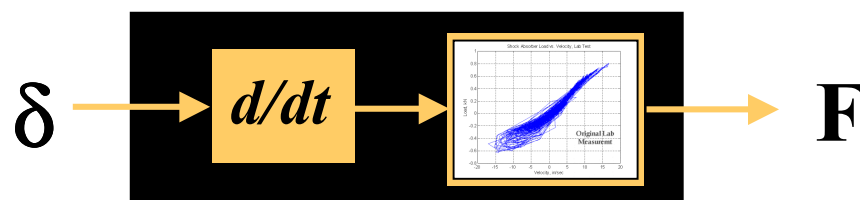
Modeling Dampers with Displacement Input

Normally, it's not feasible to fit curves to random force-displacement data

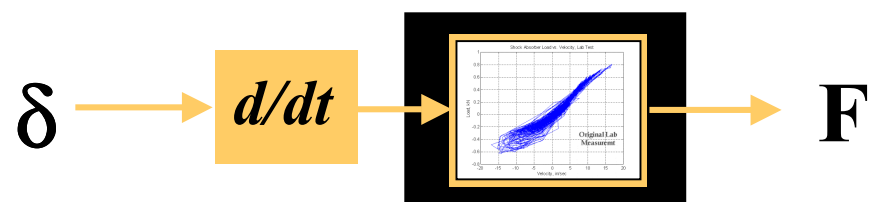


Modeling Dampers with Displacement Input

- The EDM blackbox is equivalent to differentiator and a force-velocity function



Empirical Dynamics damper model

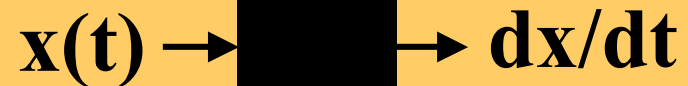


conventional damper model

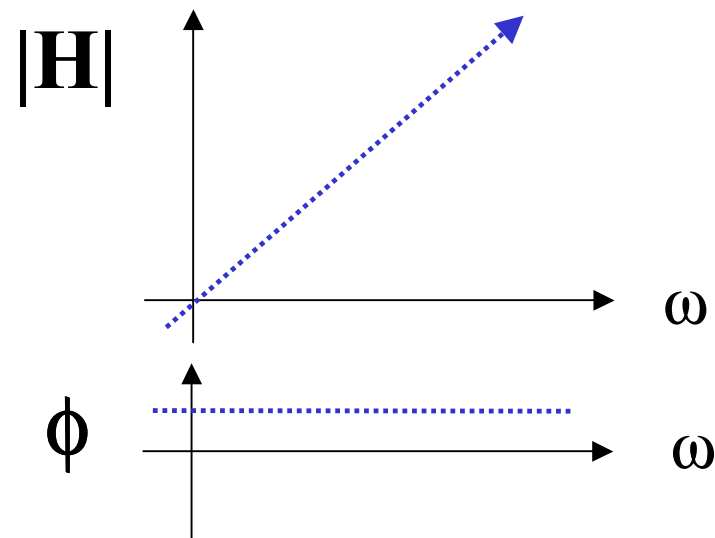
EDM Differentiation

- **Benefits**
 - Reduce differentiation errors
 - Noise immunity
 - Model different types of derivatives.

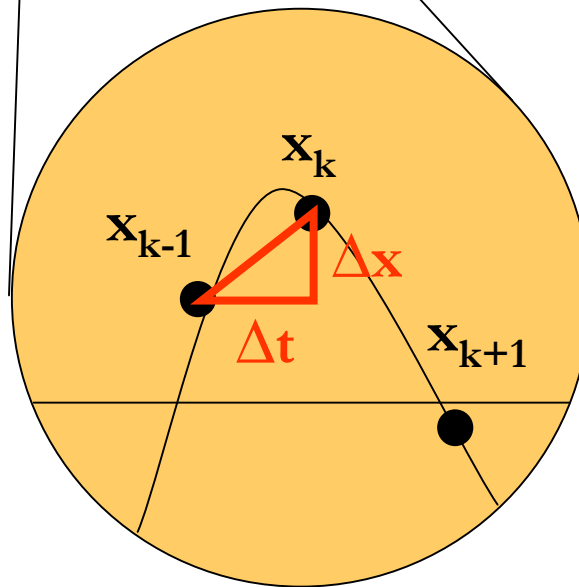
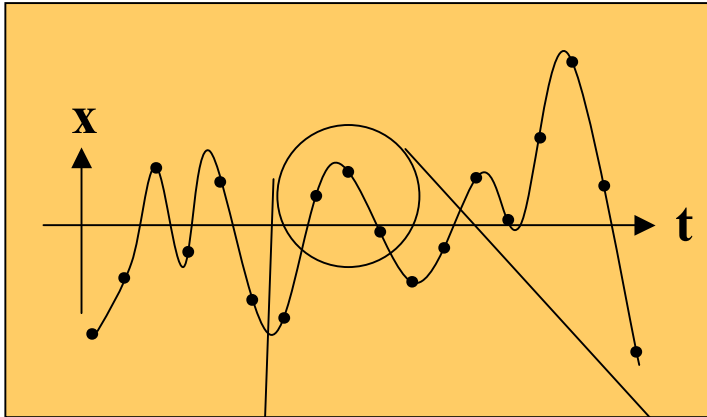
Ideal Differentiator



- Frequency response: $dx/dt \Leftrightarrow j\omega X = H$



Finite Difference Differentiator

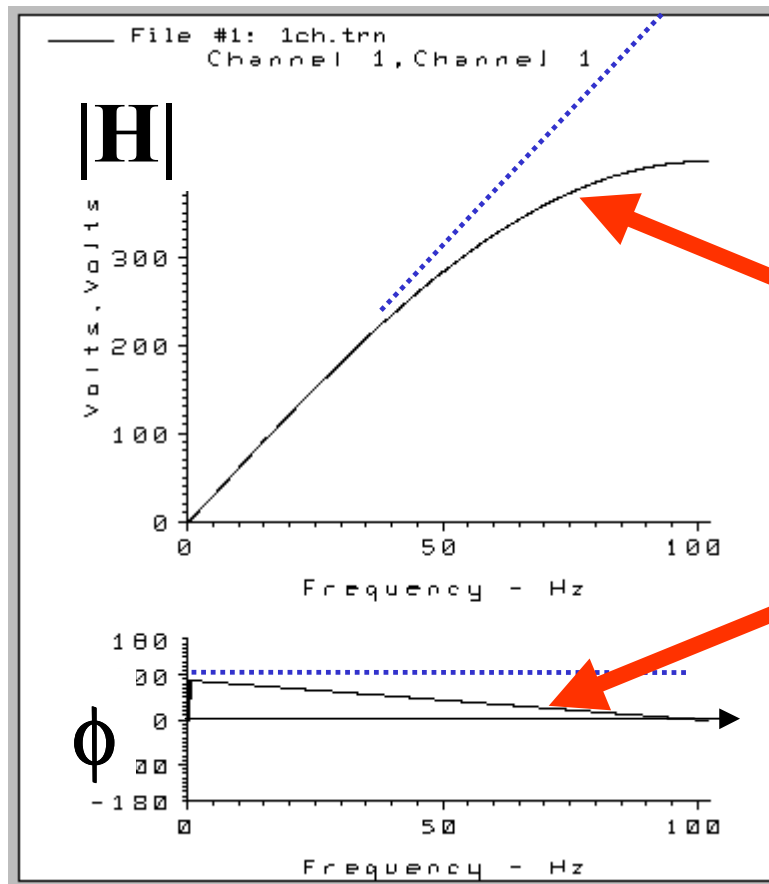


Backward Difference:

$$\dot{x}_k = \frac{x_k - x_{k-1}}{\Delta t}$$

Finite Difference Differentiator

Frequency Response



..... Ideal

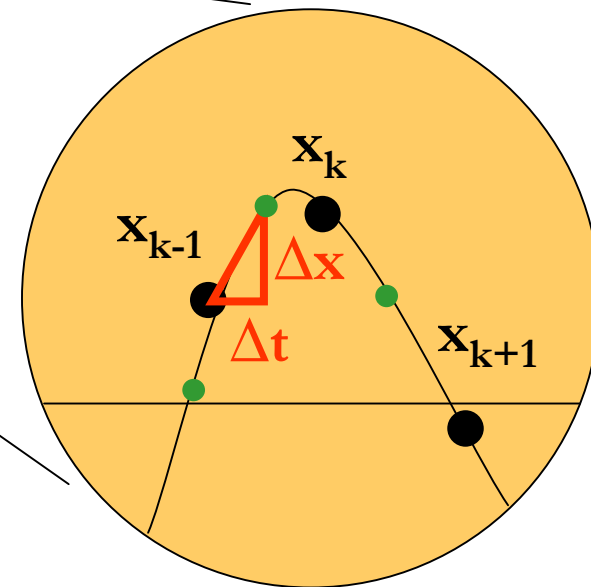
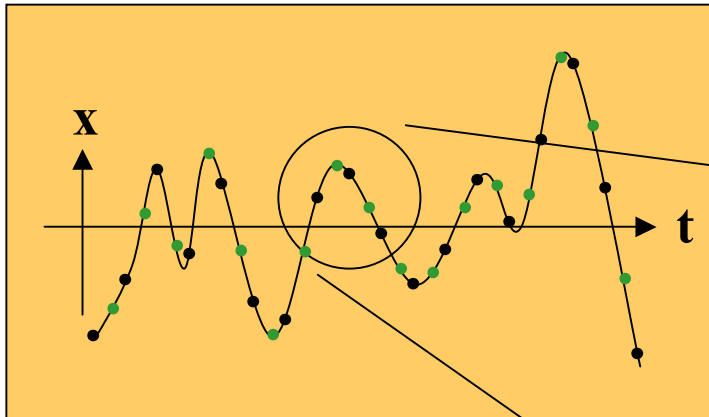
— Finite Diff.

High frequency
attenuation

Large phase shift

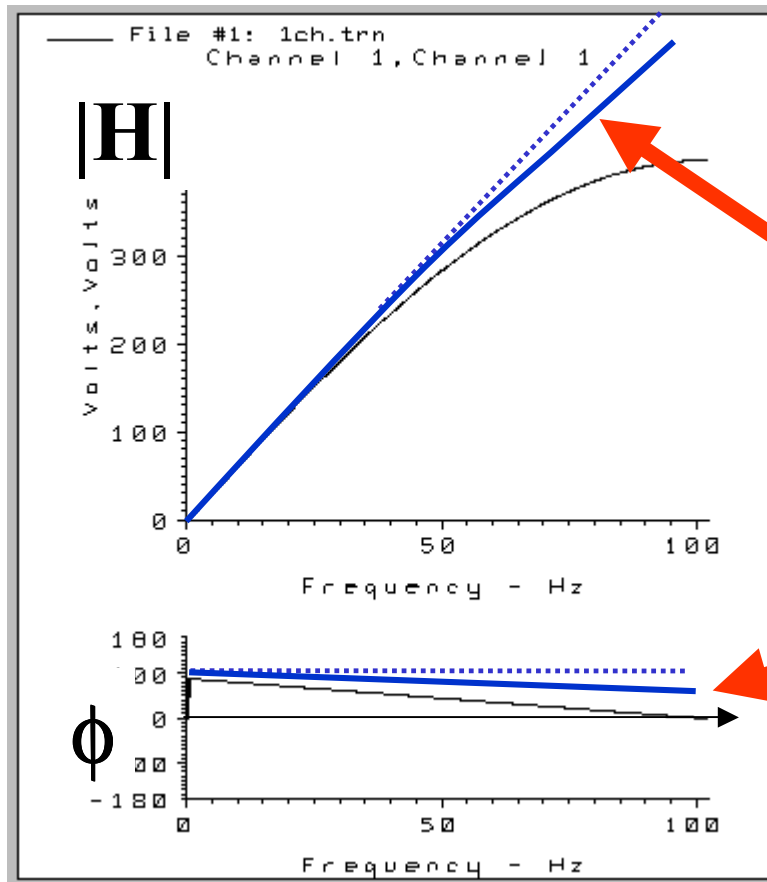
Improved Differentiator

- Use higher sample rate



2x Upsampled Differentiator

Frequency Response



- Ideal
- Finite Diff.
- 2x Sampling

Better high frequencies

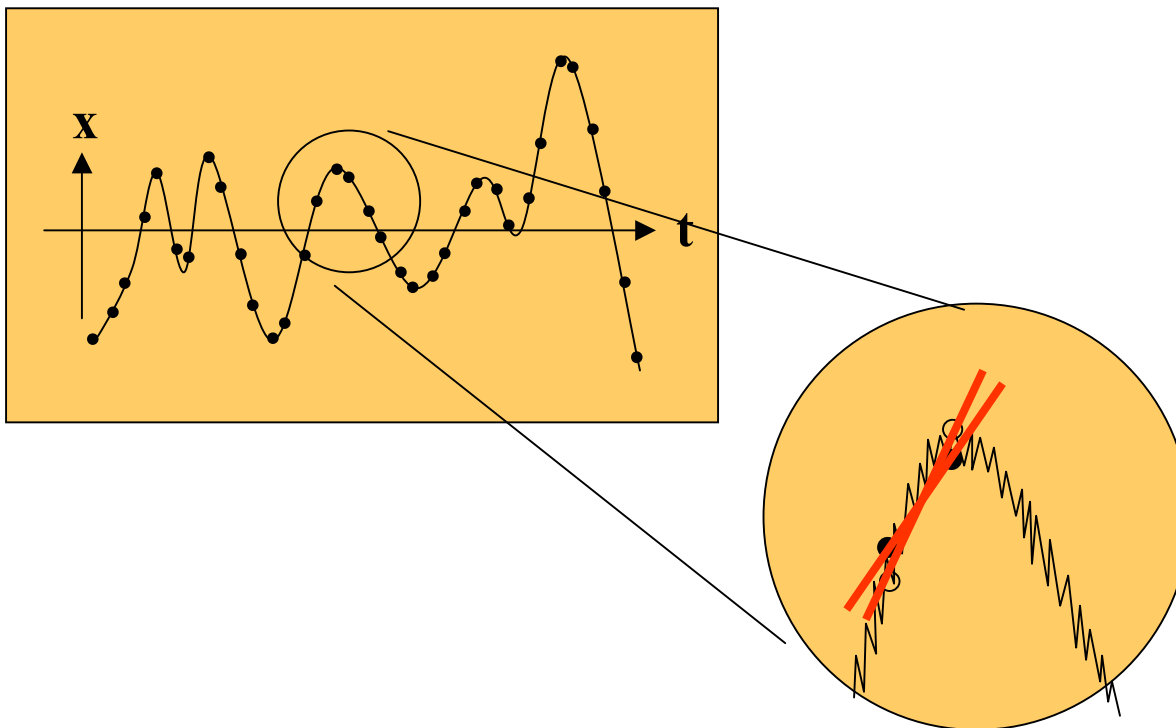
Less phase shift

Higher Sample Rate Differentiation

- **Shortcomings**
 - More samples => more calculations
 - Improved high frequencies => more susceptible to noise

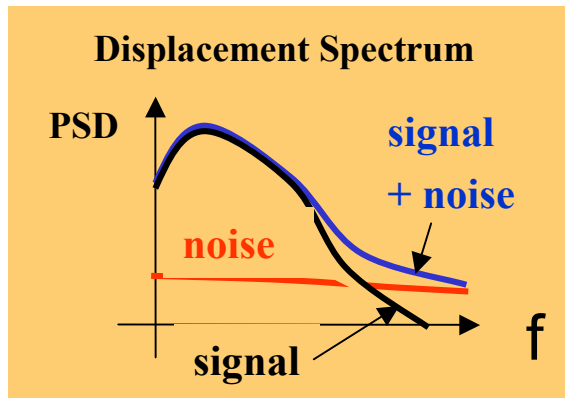
Differentiators: Dealing with Noise

- typically, noise problems occur at higher frequencies

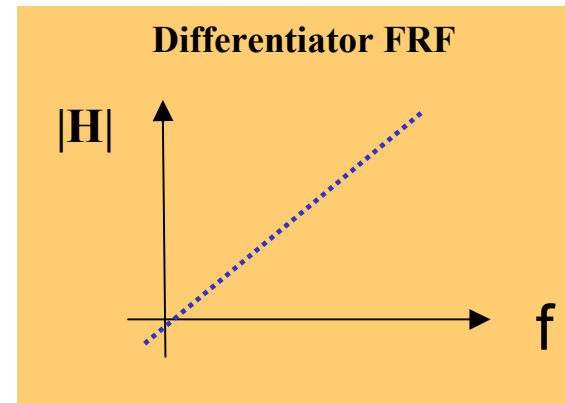


Differentiators & Noise

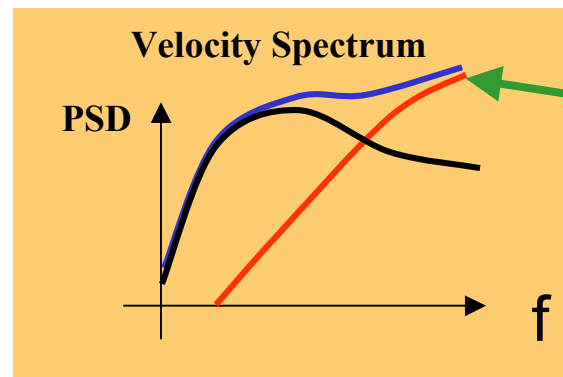
- Frequency Domain Effects



*



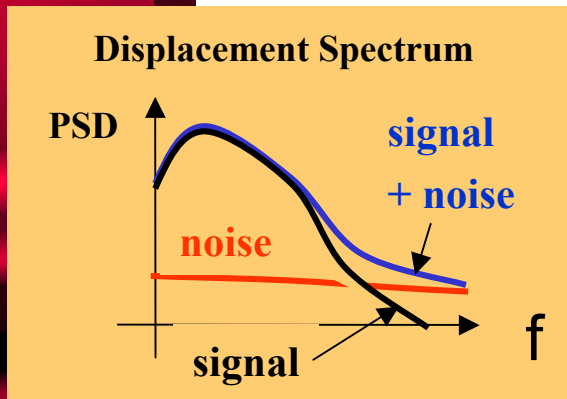
=



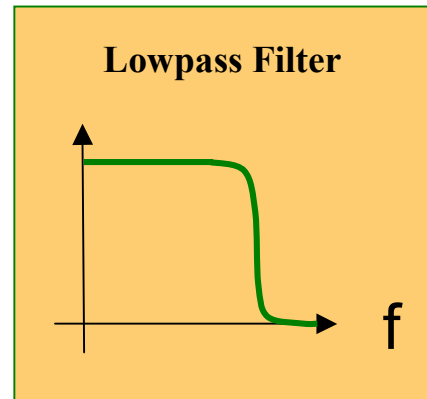
High frequency noise dominates the differentiated signal

Differentiators & Noise

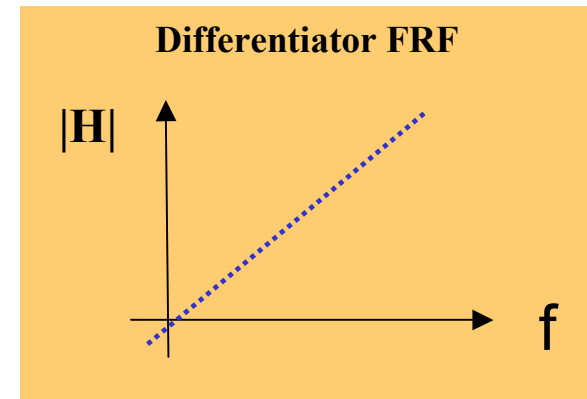
- Use of Lowpass Filter



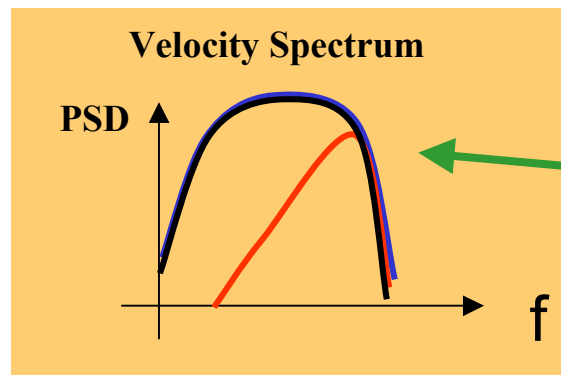
*



*



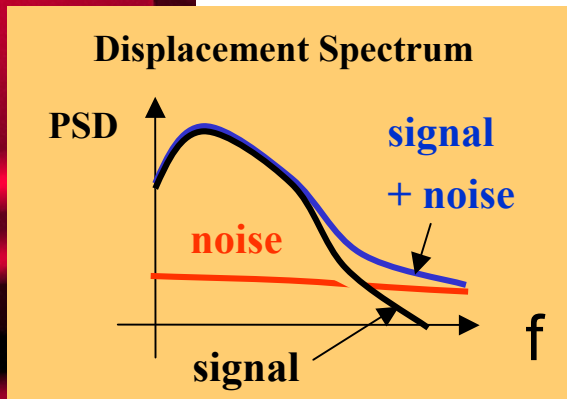
=



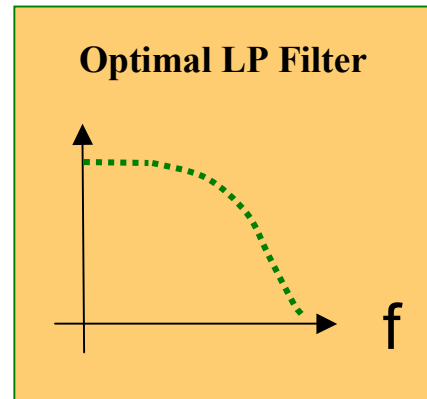
**High frequency
noise contribution
is reduced**

Differentiators & Noise

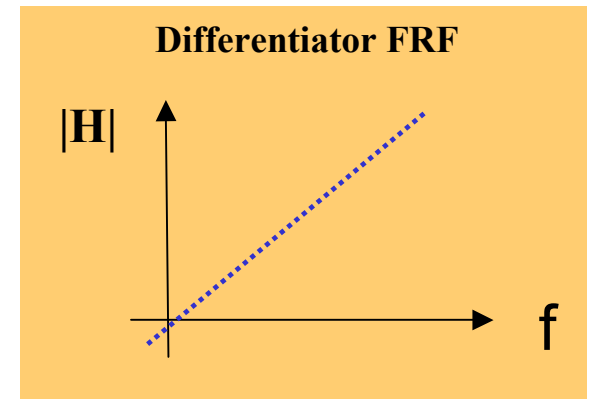
- Use of Optimal Lowpass Filter



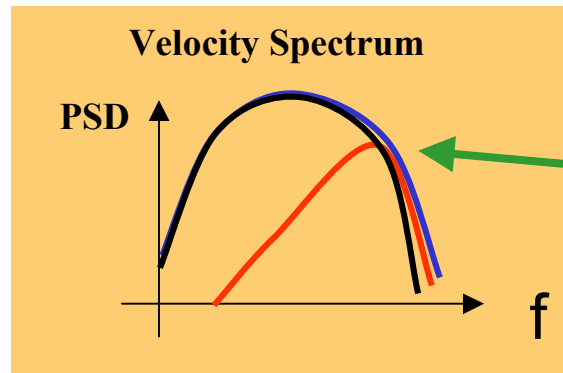
*



*

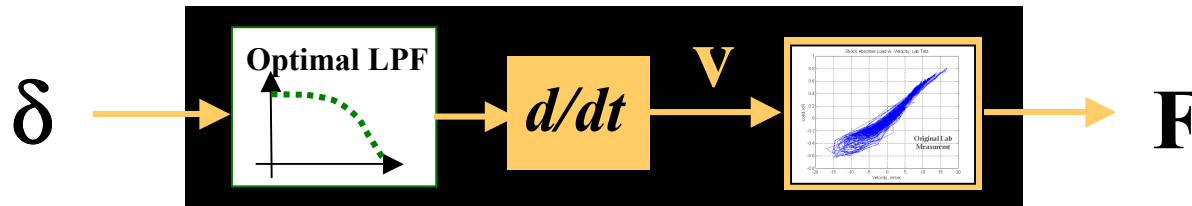


=



**Improved
high frequency
signal response**

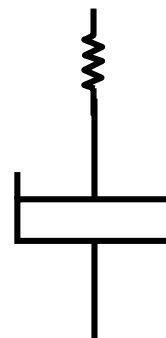
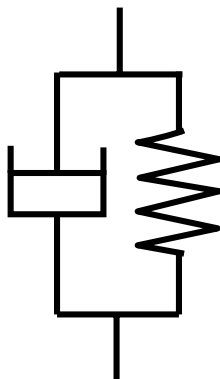
Differentiation using EDM



- EDM differentiation effectively includes upsampled, optimal LPF

Alternate Derivatives

- Higher order derivatives (d^k/dt^k)
- Derivative + proportion



Choice of EDM Damper Input

- EDM differentiation
 - more effective than typical finite difference, LPF schemes
 - ability to handle complex phenomena
 - no additional modeling cost
 - ~~Conclusion: For dampers, displacement input is preferred over velocity.-----~~
 - Stop the press !!
 - For using EDM in ADAMS, a better choice is a combination of displacement & velocity
-

Practical Considerations in Nonlinear Blackbox Modeling

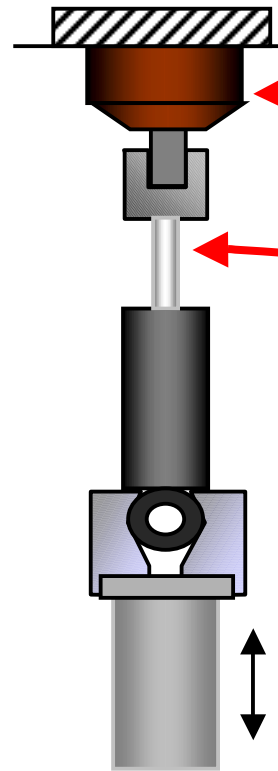
- Review of Empirical Dynamics Modeling
- Choice of Model Input
 - Displacement
 - Velocity
- **Representation of Inertial Forces**
- Choice of Blackbox Boundaries

Representation of Inertial Forces

- A mechanical lab test may exclude important inertial force components
- These components may affect ED model accuracy at high frequencies

Representation of Inertial Forces

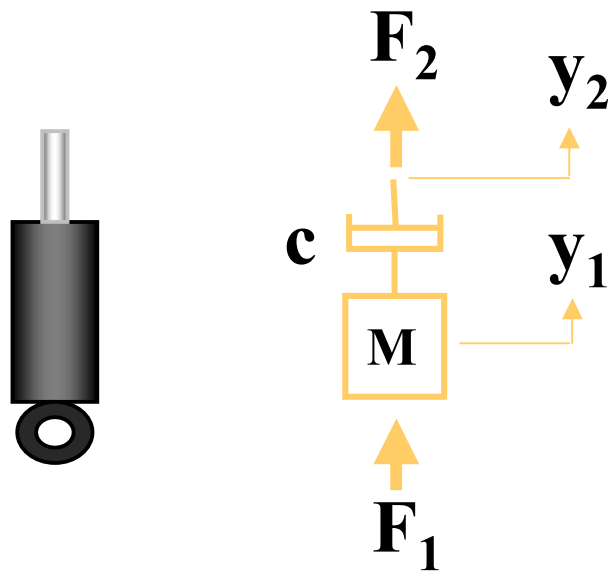
- Example: Std damper test rig



- Load cell between specimen and ground
- Typically, low mass end (rod) attaches to load cell
- Actuator attaches to opposite end

Representation of Inertial Forces

- Equations of Motion



$$\Sigma \mathbf{F} = M\ddot{y}_1 = \mathbf{F}_1 + \mathbf{F}_2$$

$$\mathbf{F}_2 = c (y_1 - y_2)$$

Representation of Inertial Forces

- Equations of Motion

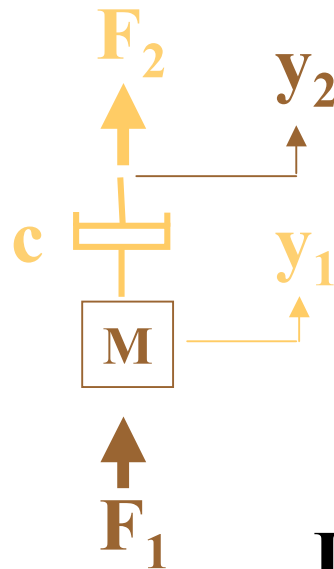
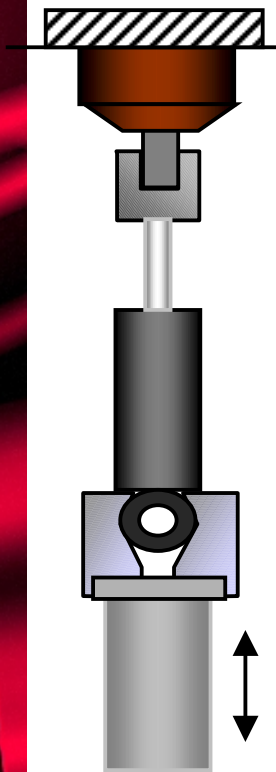
Damper Test:

$$y_2 = 0$$

F_1 is not measured

$$\Sigma F = M\ddot{y}_1 = F_1 + F_2$$

$$F_2 = c(y_1 - y_2)$$

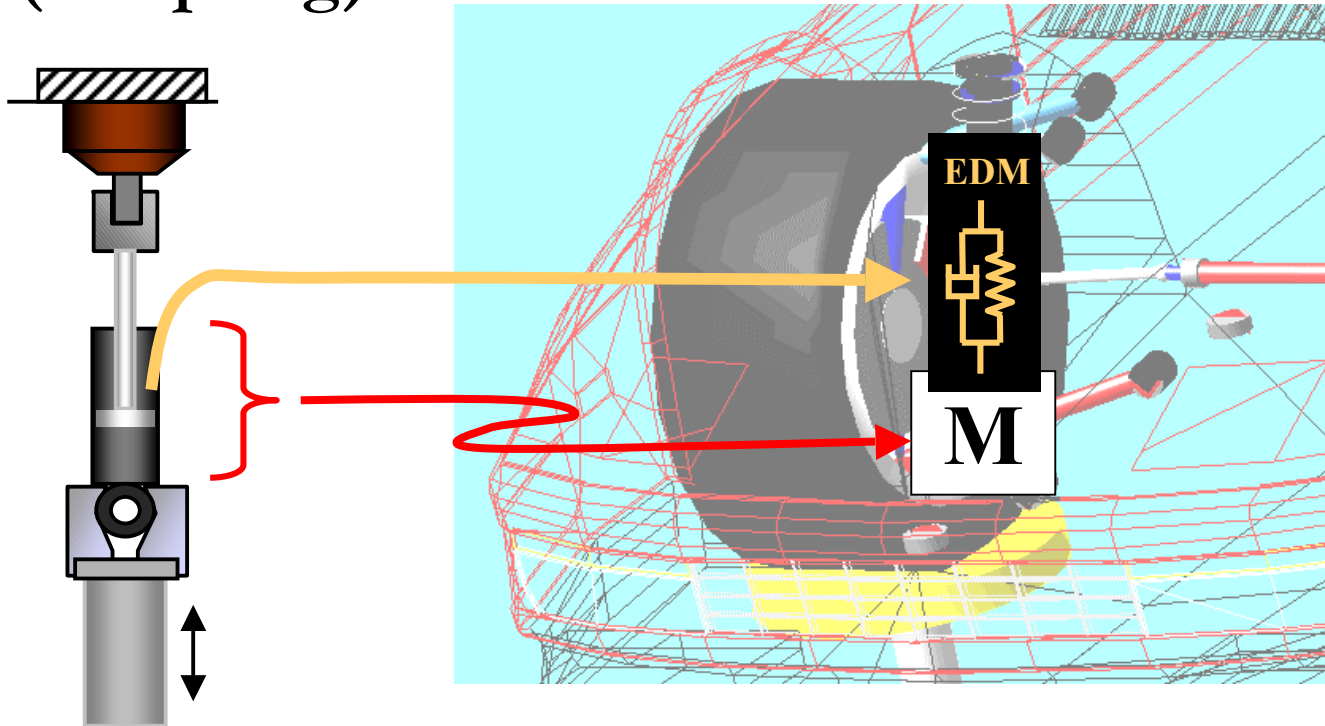


Inertial force $M\ddot{y}_1$ is ignored
Only damping is modeled

Inertial Forces, Modeling Workarounds

Simple
Damper
(no spring)

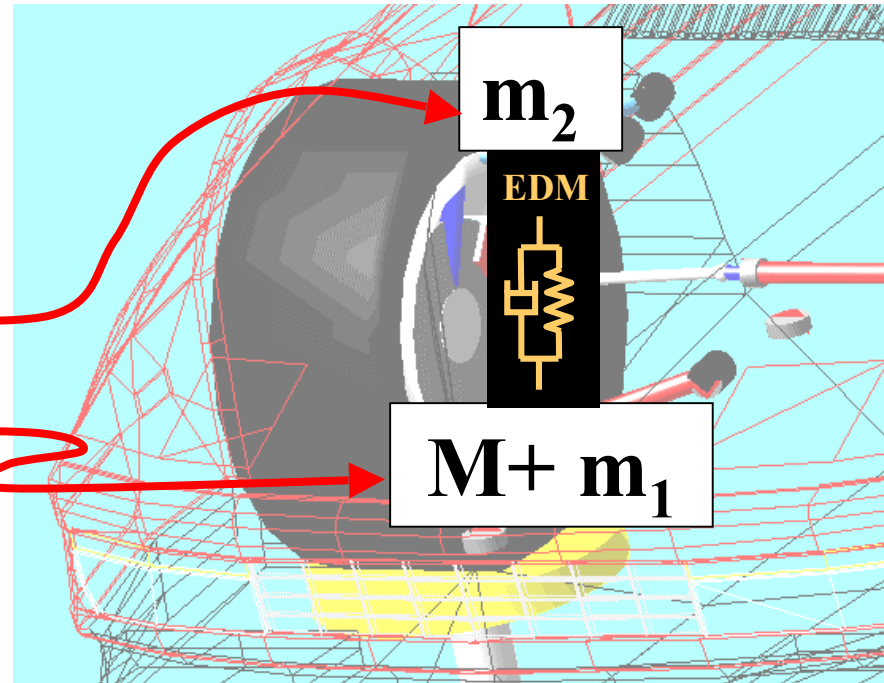
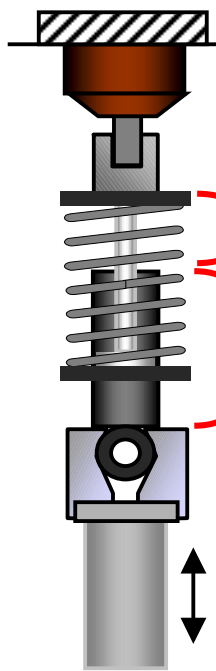
Add lumped mass of s/a body to
one end



Inertial Forces, Modeling Workarounds

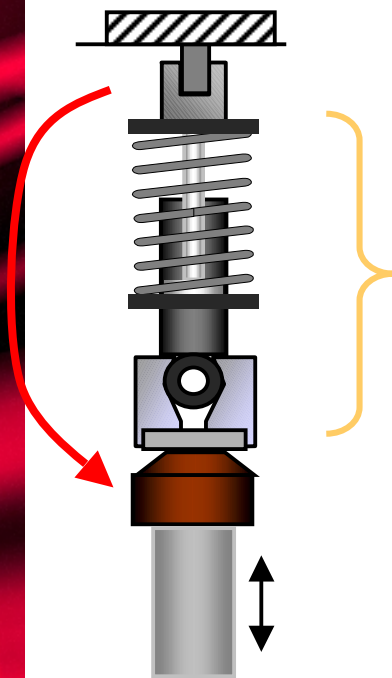
**Strut
(incl spring)**

Assign a fraction of mass to each end

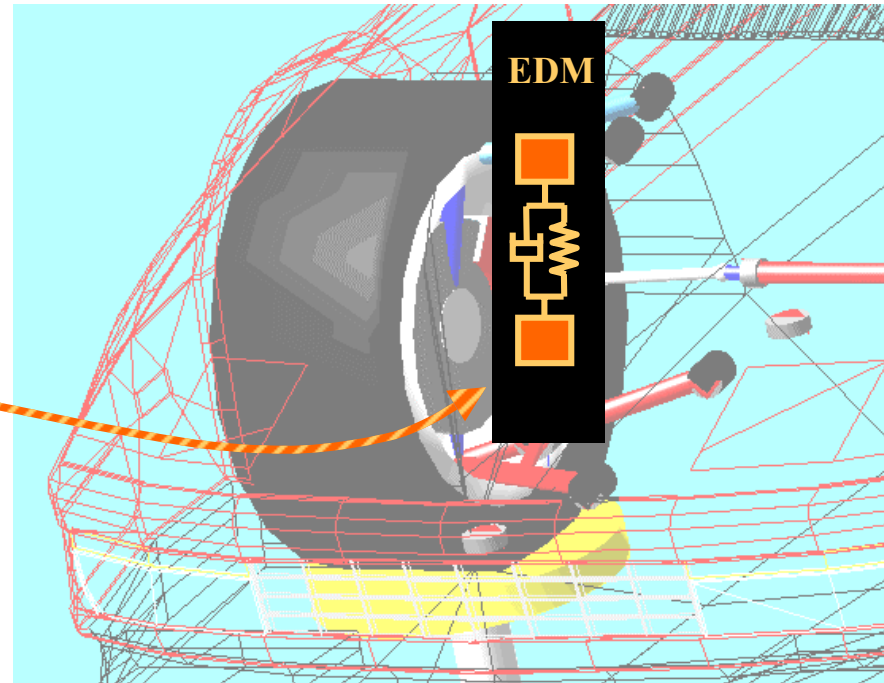


Inertial Forces, Modeling Workarounds

**Damper or
Strut**



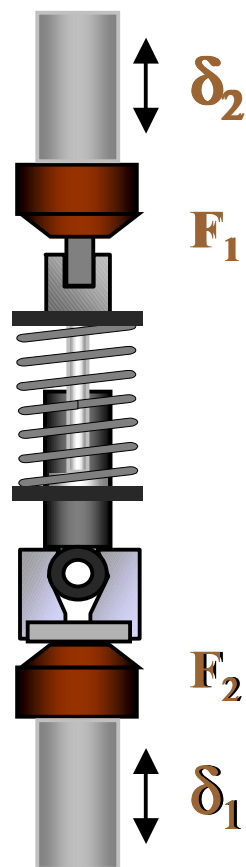
Use load cell at the active end



- (+) Some inertia forces are included in the blackbox
- (-) Difficult to measure - moving load cell

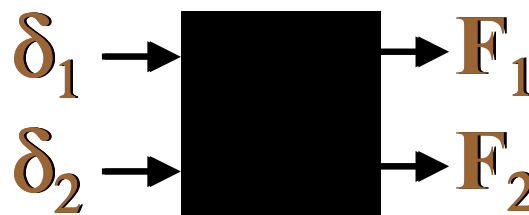
Inertial Forces, Modeling Workarounds

Damper or Strut



Use alternative test rig

- two actuators
- two load cells
- => 2 input, 2 output blackbox



- (+) Inertial forces are included in the blackbox
- (-) *Excess inertia from grips*
- (-) *Difficult to measure - moving load cell*

EDM Representation of Inertial Forces

- Other specimens => similar issues

- Problem is inevitable at sufficiently high accelerations

Inertial Forces and EDM

Summary

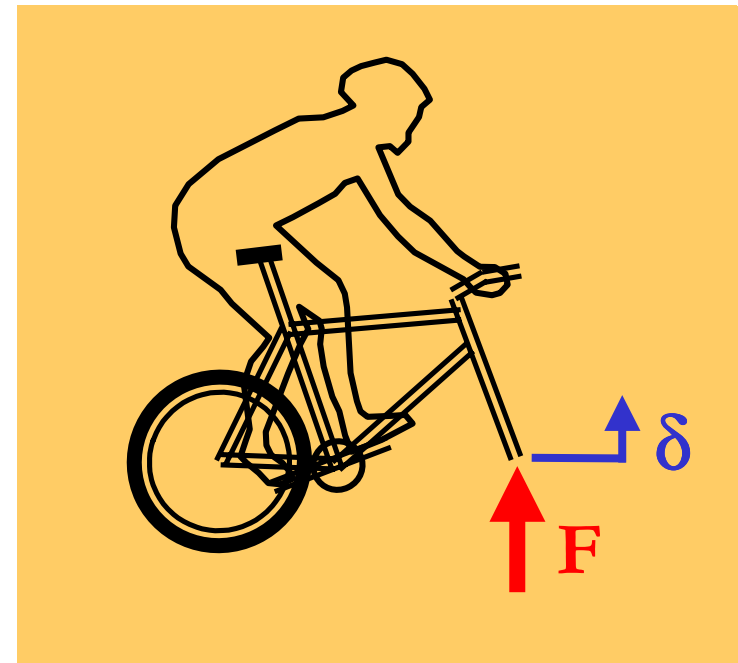
- For accurate EDM characterization at high frequencies, a special test rig may be necessary
- Remember to consider the test rig used to generate ED model and include the “effective” mass in your Adams simulation.

Practical Considerations in Nonlinear Blackbox Modeling

- Review of Empirical Dynamics Modeling
- Choice of Model Input
 - Displacement
 - Velocity
- Representation of Inertial Forces
- **Choice of Blackbox Boundaries**

Choice of Blackbox Boundaries

- System = bicycle
w/ suspension
fork, + rider
- ED model :
predict vertical
force into fork, for
any road profile
input

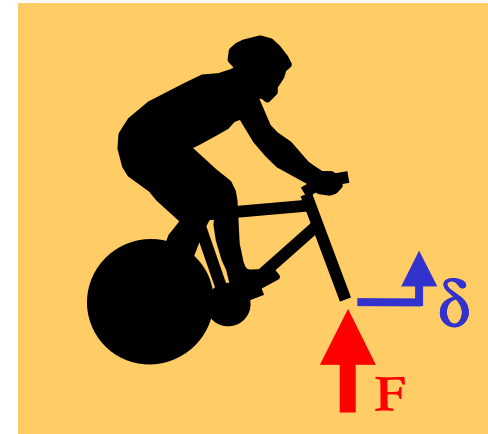


Choice of Blackbox Boundaries

Scenario 1

Blackbox =

Bike + Rider combination



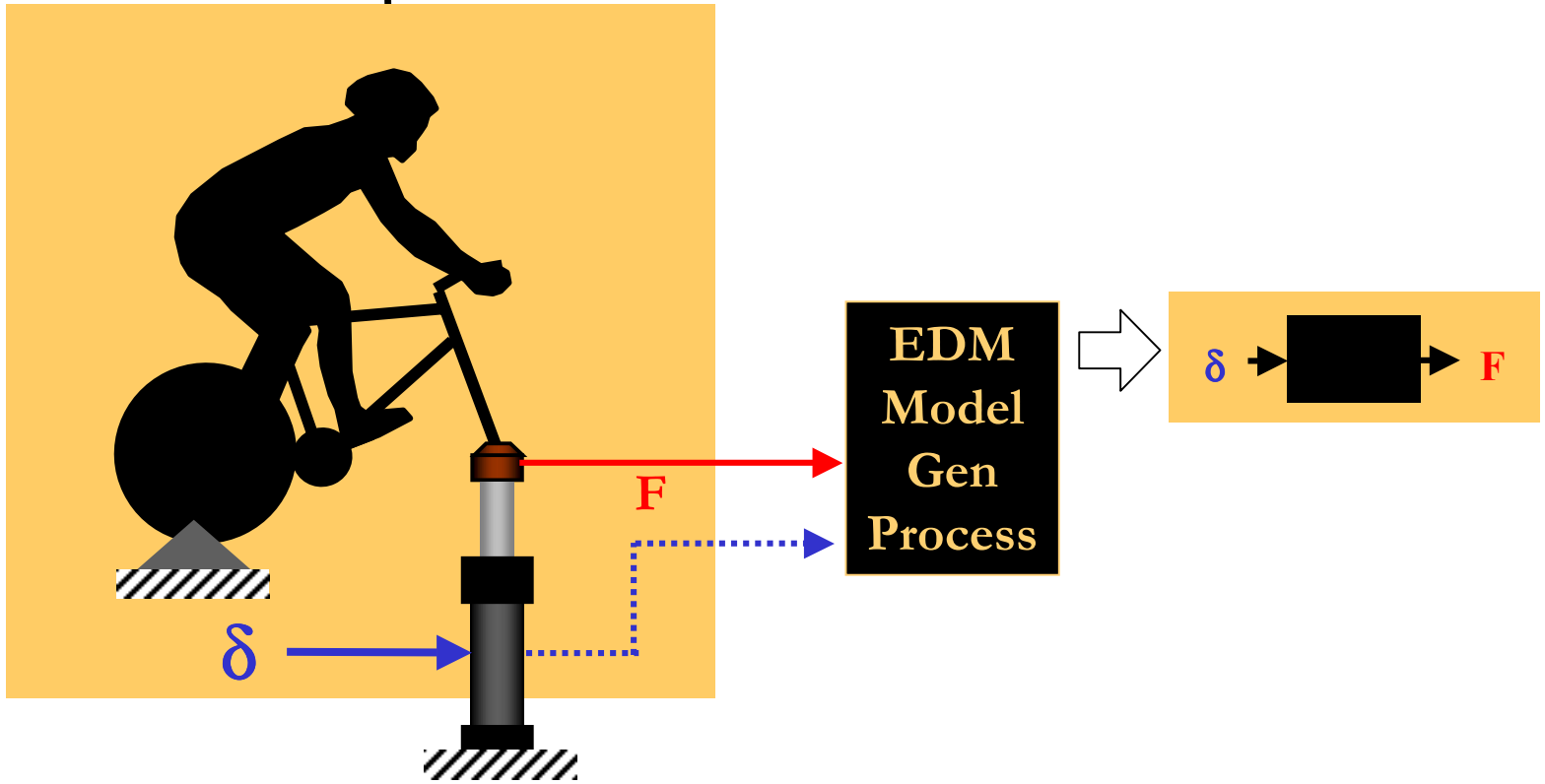
(+) can model displacement
=> force directly
(SISO model)



(-) model is valid only for
one particular Rider

Lab Test Proposal

- Single Actuator
- Rider (mechanically consistent !)
- Load & displacement measurements



Choice of Blackbox Boundaries

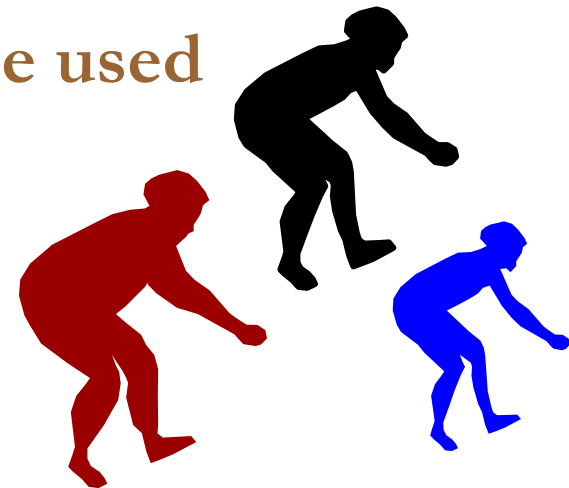
Scenario 2

1 Blackbox = Bike

1 Other model = Rider

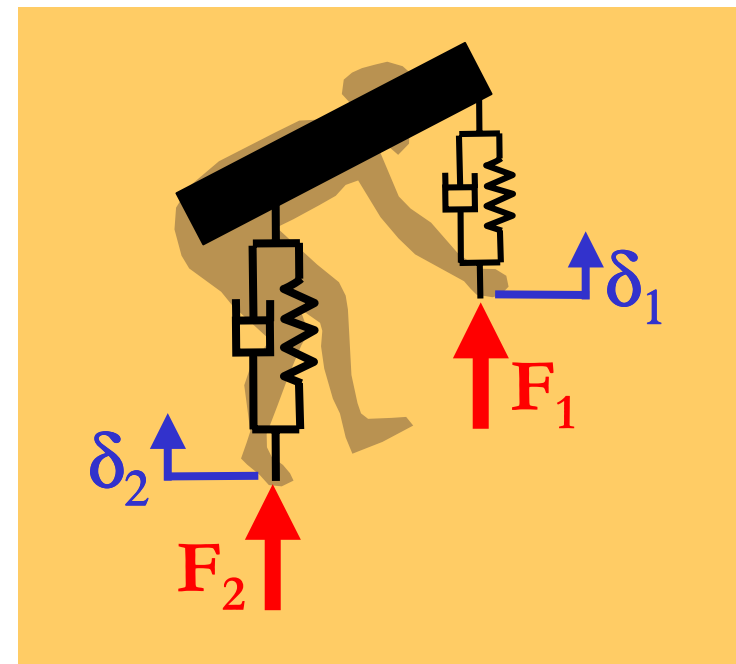


(+) Different Riders can be used



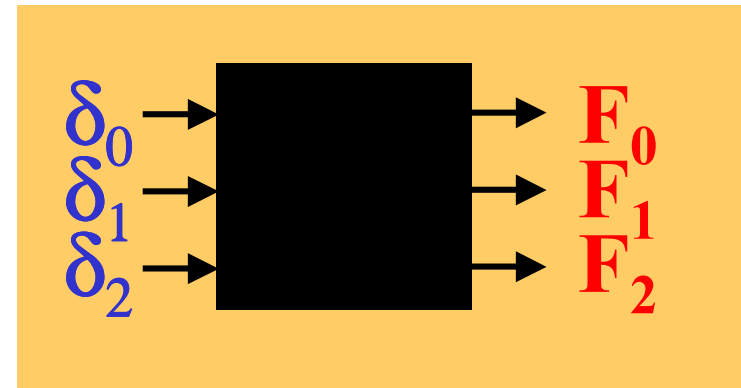
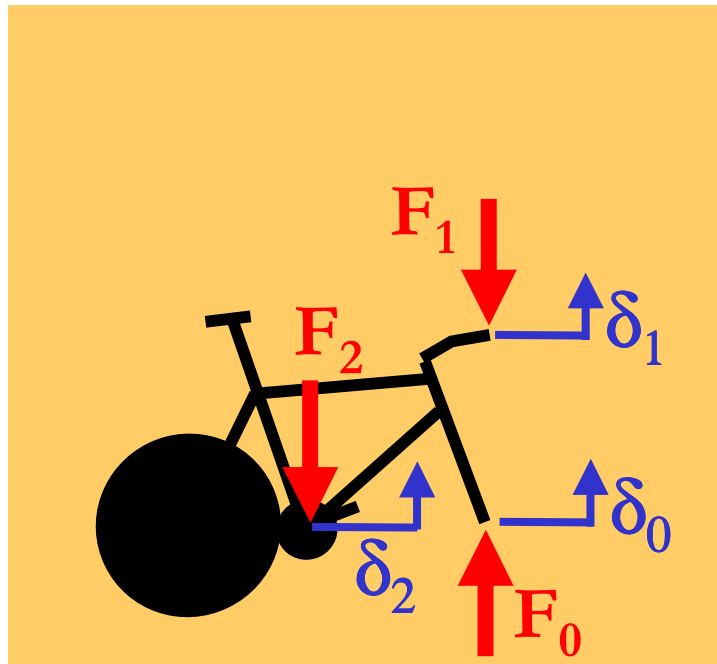
Rider Model

- Simple:
 - mass
 - bio-suspension (arm & knee compliance)



Choice of Blackbox Boundaries

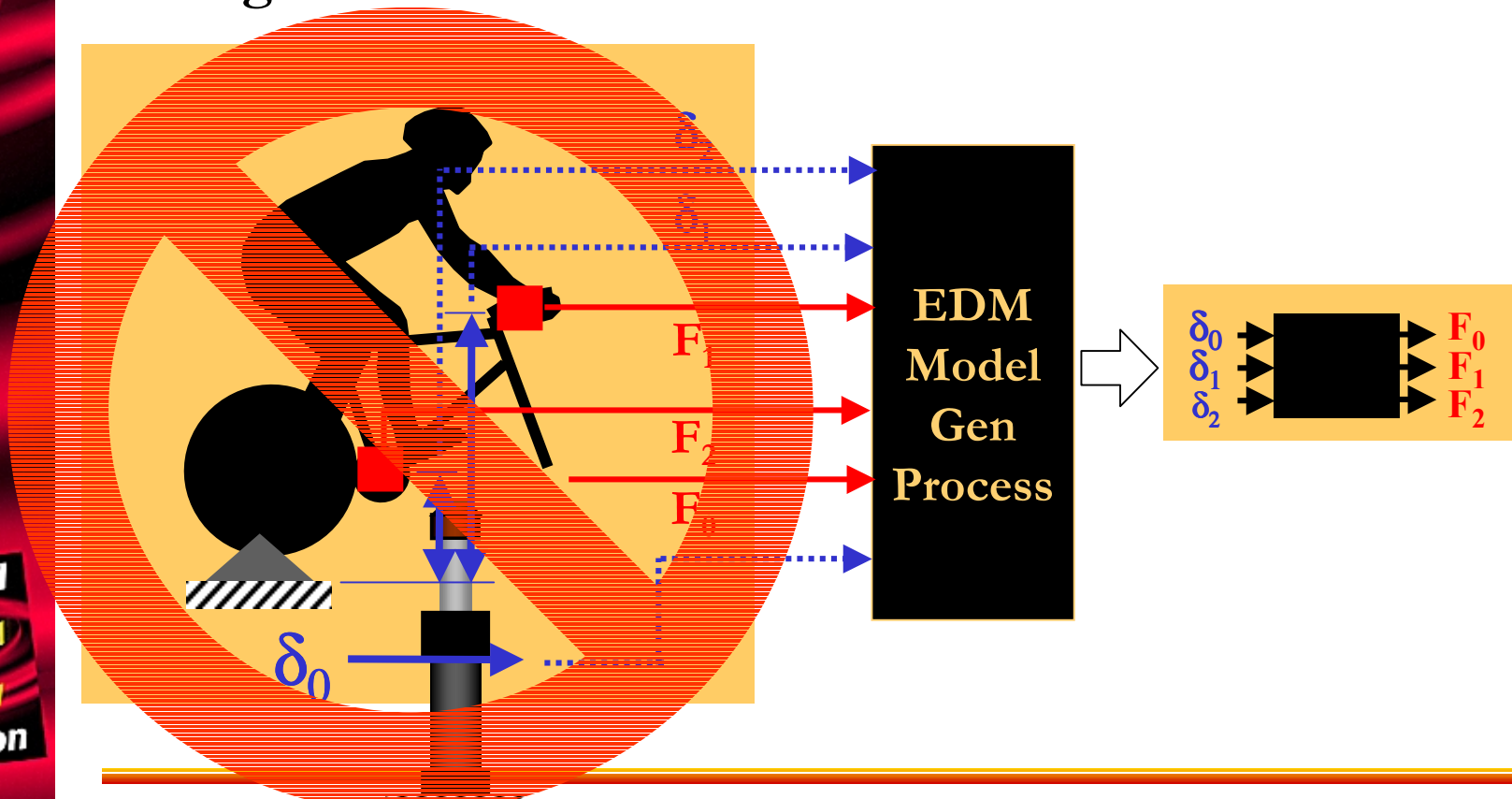
Blackbox = Bike



(-) Requires more
blackbox inputs !

Lab Test Proposal

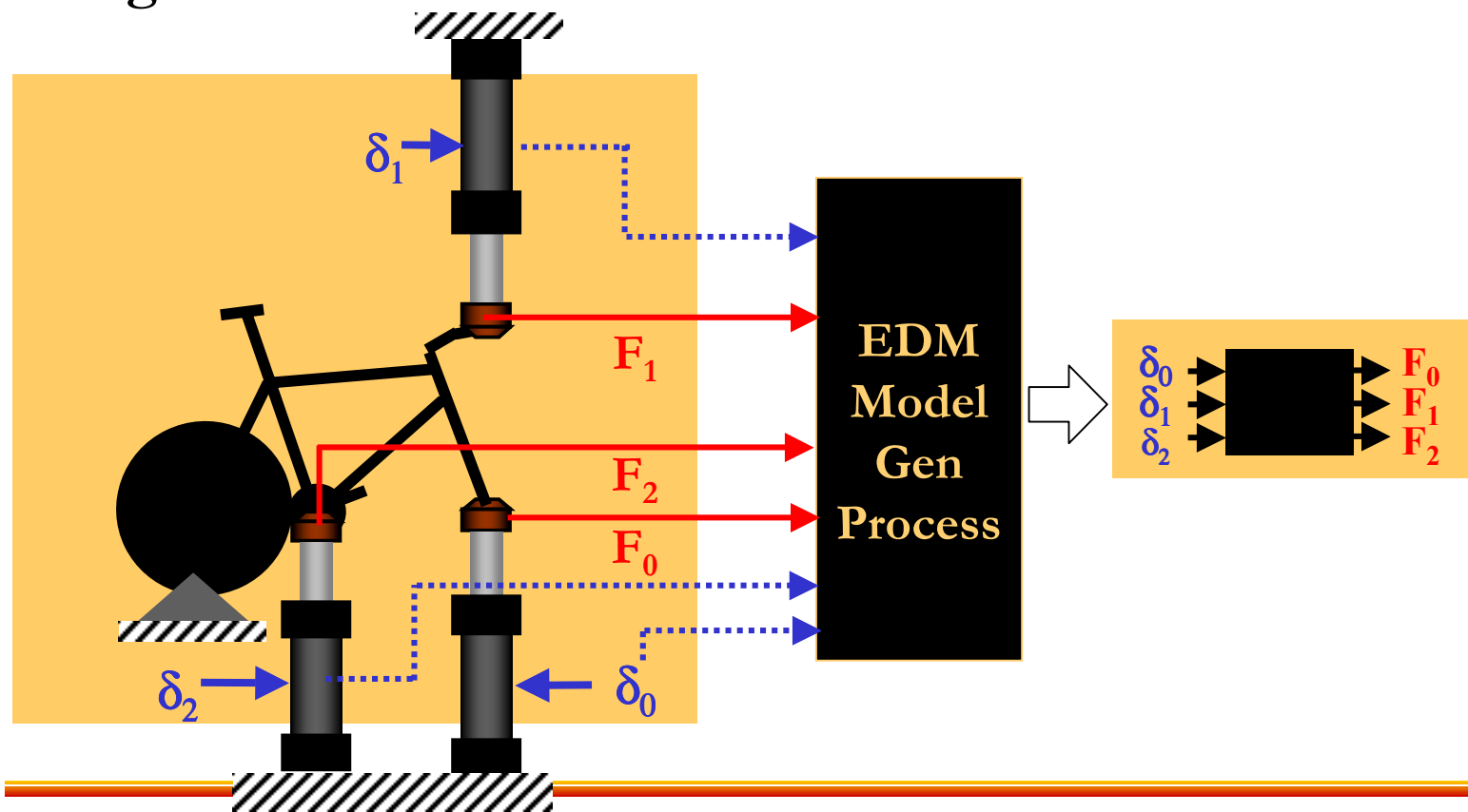
- Single Actuator
- Rider
- 6 signal measurements



Integrated
analytical
physical
virtual
Simulation

Lab Test Proposal 2

- 3 actuators, independently controlled
- No rider !
- 6 signal measurements



Lab Test Requirement

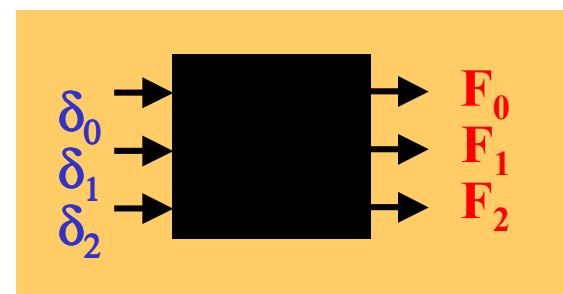
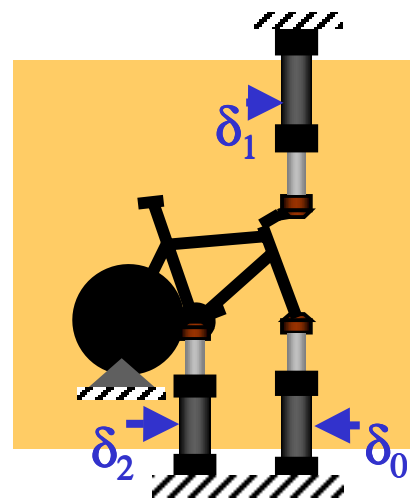
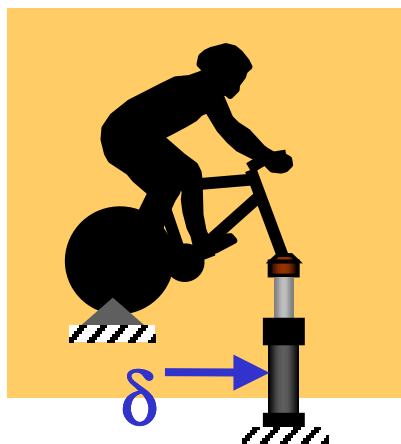
**Number of independently
controlled actuators**

=

Number of ED model inputs.

Choice of Blackbox

- **Simplicity vs. Adjustability**



Choice of Blackbox Boundaries

- Apply same thinking to more advanced systems

Choice of Blackbox Boundaries

Summary

- **Choice of blackbox boundaries determines:**
 - the adjustability of a model
 - the number of model inputs
 - the number of independent actuators for the lab test

Summary

- **For successful nonlinear blackbox modeling, understand the limitations:**
 - Choice of model input
 - Inertial force measurement
 - Blackbox boundaries

Reference

**A. J. Barber,
Accurate Models for Bushings and Dampers using
the Empirical Dynamics Method,
ADAMS International Users Conference, 1999
(Berlin)**

<http://support.adams.com/usercon/euc99/euc99-thu.htm>