

Noise-sensitivity in stochastic nonlinear models with delays and discontinuities

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Noise reducing complexity:

$$dx = (\mu x - y^2 + 2z^2 - \delta z)dt$$

$$dy = y(x - 1)dt + \sqrt{2}\epsilon dW$$

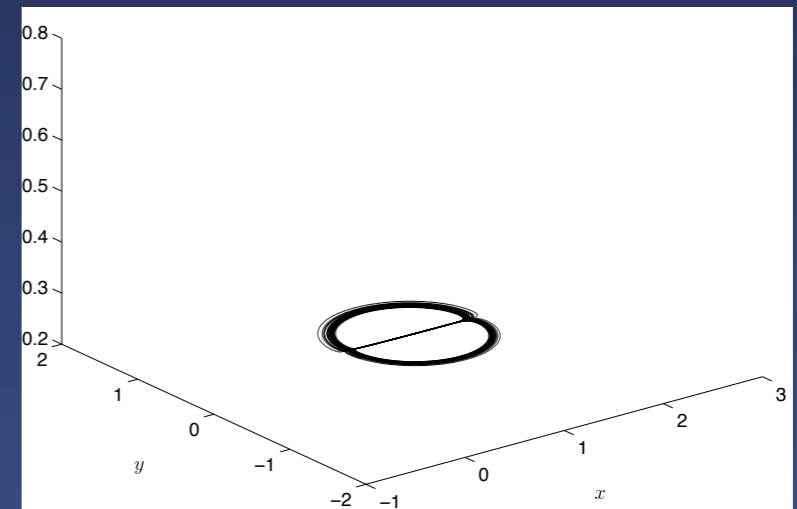
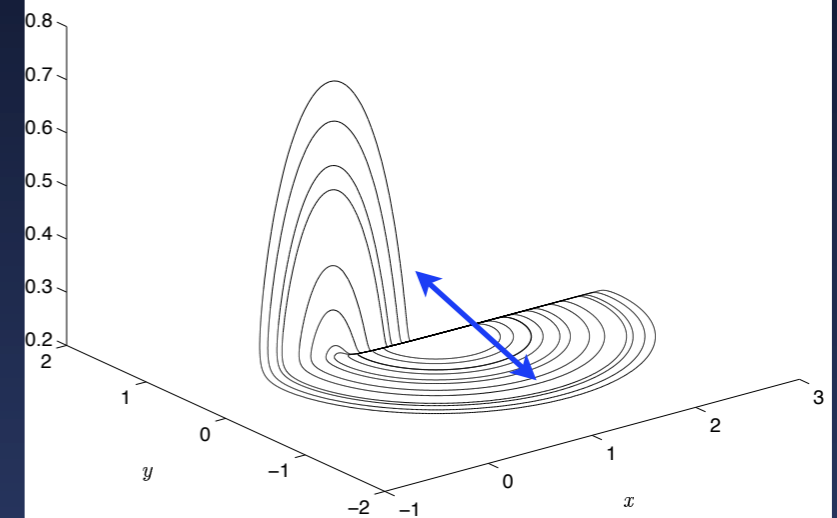
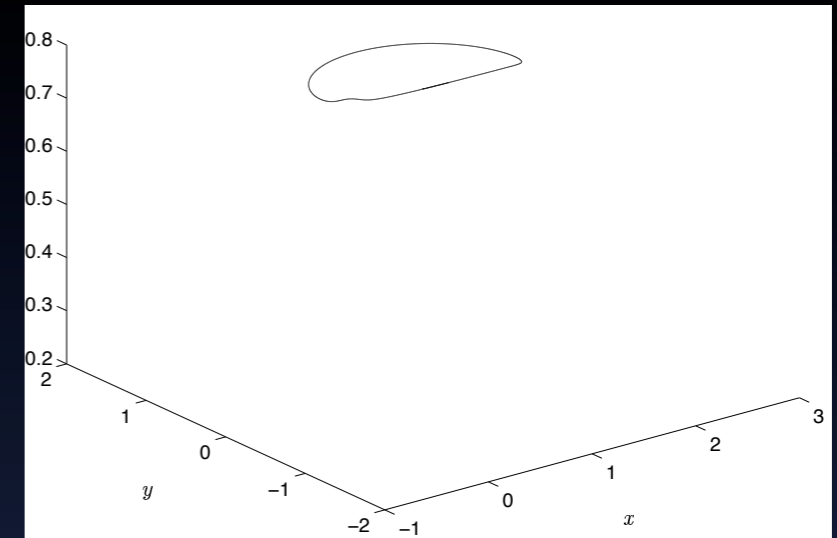
$$dz = (\mu z + \delta x - 2xz)dt,$$

Hughes, Proctor 1990

Noise ($dW =$ white noise)
perturbs trajectories near slow
manifold ($\mu \ll 1$)

Nearly periodic trajectory, well
approximated by 2D probability
density

Also, important in computations:
identify potential computational error



K., Papanicolaou, 1998

Computational questions:

- Convergence of numerical methods: Strong (pathwise) convergence vs. Weak convergence
- Dynamics of numerical schemes
- Stochastic bifurcations: qualitative changes of dynamics at specific parameter values
- Stability of schemes
- Questions can vary with types of noise

(Concentrating on SDE's)

Examples:

- Convergence of numerical methods: Strong (pathwise) convergence vs. Weak (in distribution) convergence

Strong: $E|X - Y_h|$

Weak: $|E[X] - E[Y_h]| \quad |f_X(x) - f_{Y_n}(y)|$

Forward, Backward Kolmogorov equations: PDE's

Karniadakis (2013), Schwab (2012) (DG, discontinuous dynamics)

Examples: $dY = a(Y)dt + b(Y)dW$

Strong:

$O(\sqrt{h})$ Euler-Maruyama $Y_{n+1} = Y_n + a(Y_n)h + b(Y_n)Z_n\sqrt{h}$

$O(h)$

θ - Maruyama method

Milstein

$Y_{n+1} = Y_n + a(Y_n)h + b(Y_n)Z_n\sqrt{h} + b(Y_n)b'(Y_n)/2(Z_n^2 - 1)h$

Weak:

$O(h)$ Euler-Maruyama

$O(h^2)$: Multi-step uses: $Y_n + ah + bZ_n\sqrt{h}$ $Y_n + ah \pm b\sqrt{h}$

Higher order: Additional r.v.'s, derivatives needed

Kloeden, Platen 1992

Recent examples:

- Nonlinearities:, tamed explicit methods
Hutzenhaler, 2012
- Stability of schemes: Questions can vary with types of noise, or quantities of interest

$dx = a(X)dt + b(X)dW$ Multiplicative noise: could have $X=0$ as an equilibrium

$dx = a(X)dt + bdW$ Additive noise: $X=0$ is not an equilibrium

Buckwar, Riedler, Kloeden, 2011

- Dynamics of schemes
- Dynamical behavior

Stability for nonlinear SDE's: contractive conditions

Buckwar, Riedler, Kloeden, 2011

Non-normal drift - interaction of drift and diffusion in discretized system (stability) Buckwar, et al 20

Ito vs. Stratonovich interpretation of dW :

endpt vs. midpt evaluation of integrand, Ito friendlier for coding, Stratonovich usually used for parametric noise in applications

Dynamical Questions:

Noise driven order: “Stabilized” transients

Can't ignore: Transients from the deterministic dynamics
“Small” random perturbations drive
qualitative changes

Stochastic facilitation: Constructive roles of biologically
relevant noise in the nervous system

McDonnell, Ward Nature Neuroscience Reviews 2011

Various types of dynamics: bifurcations + delays

Discontinuous, Piecewise Smooth, dynamics: Sliding,
grazing, impacts, virtual dynamics, control

Dynamical Questions:

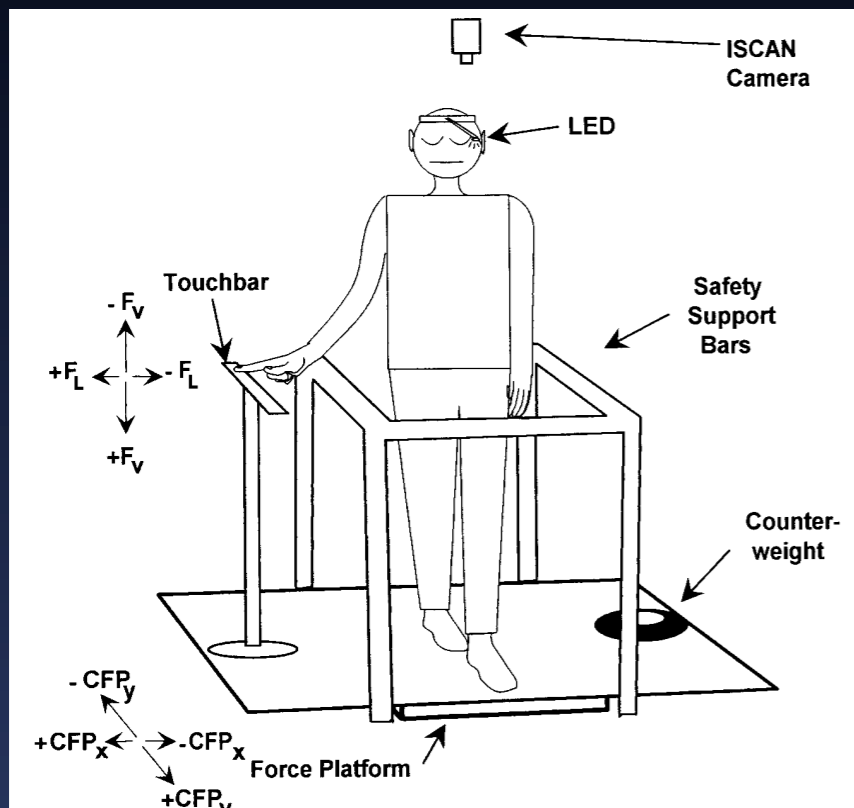
- Interplay of computational results and analysis
- Interplay of dynamics (and time scales) with stochastic perturbations - not necessarily separable
- Relatively fast, easy-to-code simulations to test and motivate “interesting” cases/parameter ranges
- Whose time is more valuable: researcher time or machine time? Significance of higher order methods

Delays: Models of Balance

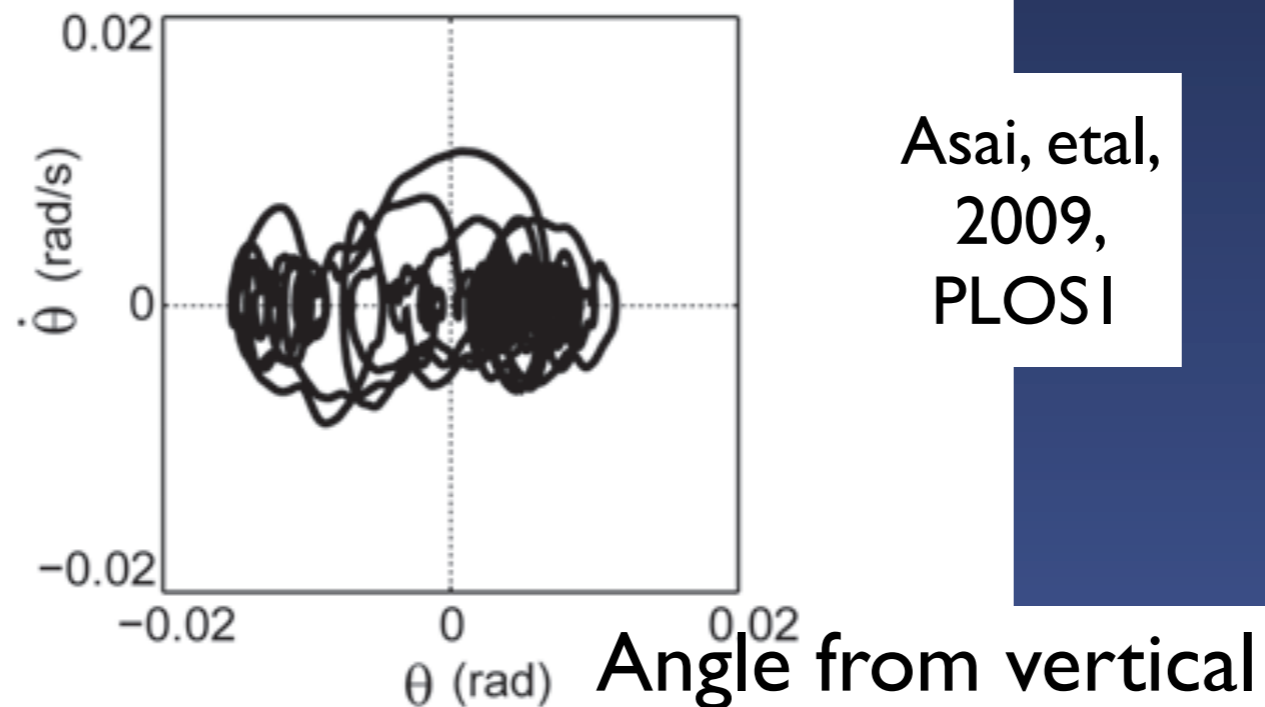
- Applications: Human Postural Sway, Stick Balancing, Robotics
- What are the contributing factors to stability, instability, balance, sway, other behaviors?

Transfer ideas between models in mechanics/
optics and biological applications:
transients sustained by stochastic effects

- Applications: Human Postural Sway, Stick Balancing, Robotics

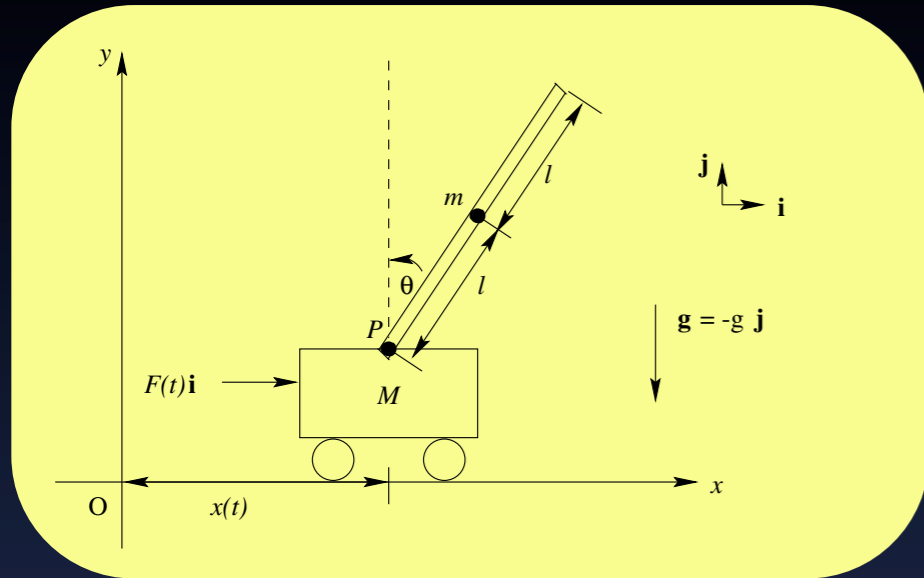


Derivative of the Angle from vertical



Asai, etal,
2009,
PLOS I

Simple model: inverted pendulum

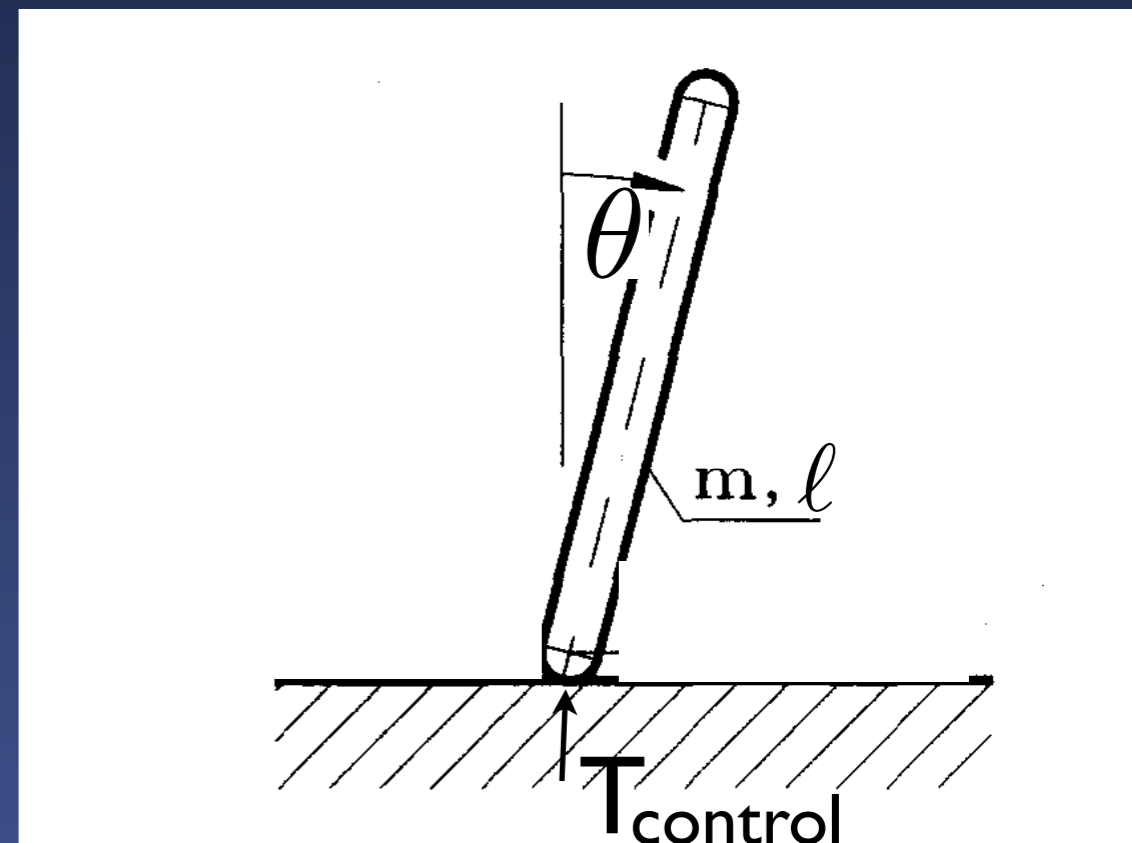


Stabilized on a cart

$$\left(1 - \frac{3m}{4} \cos^2 \theta\right) \ddot{\theta} + \frac{3m}{8} \dot{\theta}^2 \sin(2\theta) - \frac{3}{2} \frac{g}{L} \sin \theta + \frac{3F}{2L(M_p + M_c)} \cos \theta = 0.$$

$$\frac{4}{3} m \ell^2 \ddot{\theta} - m g \ell \sin(\theta) = T_{\text{control}},$$

Even more simple model:
inverted pendulum w/
pivot control (torque at
the pivot)



Reduced to essentials

$$\begin{aligned}\dot{\theta} &= \phi \\ \dot{\phi} &= \sin \theta - F(\theta, \dot{\theta}) \cos \theta \\ F &= a\theta(t - \tau) + b\dot{\theta}(t - \tau)\end{aligned}$$

$$F = \overset{(P)}{a\theta} + \overset{(D)}{b\dot{\theta}}$$

Proportional Derivative

PD control

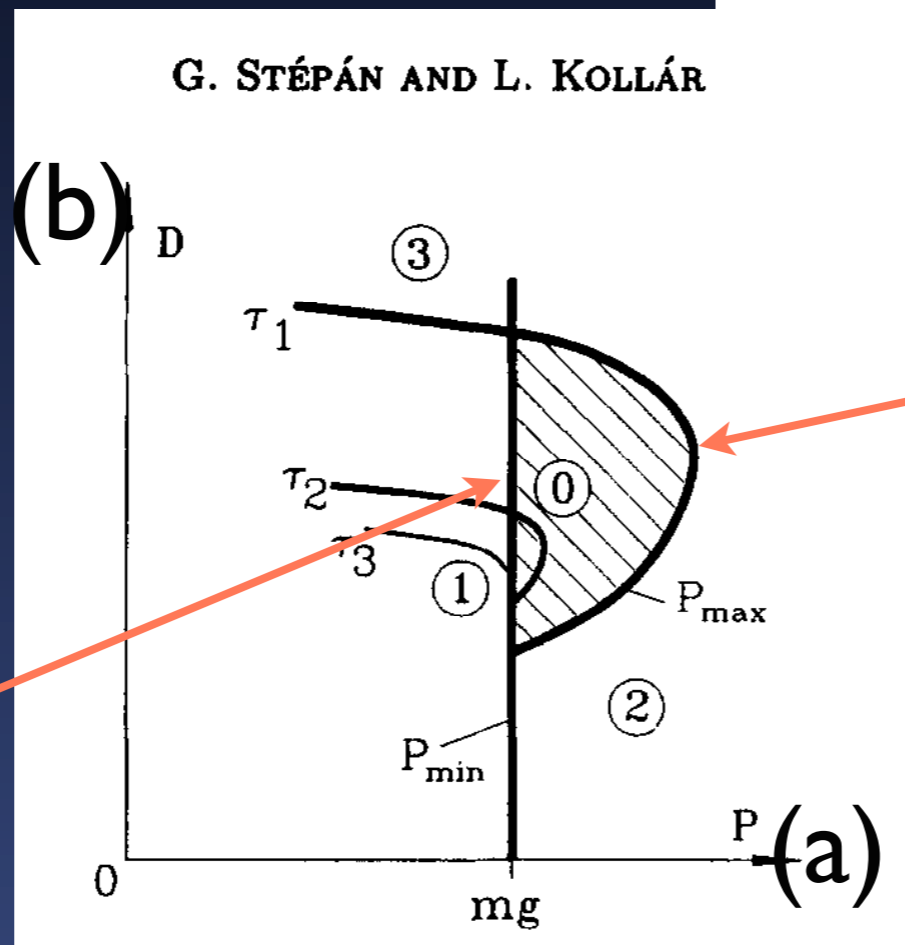
Biological considerations:

- Delay in the application of the control: neural transmission
- On-off control: not active control all the time
- Noise: Small random fluctuations can result in large changes in certain circumstances

Balance model:

Stability diagram for PD control (w/delay), control always on

$$\begin{aligned}\dot{\theta} &= \phi \\ \dot{\phi} &= \sin \theta - F(\theta, \dot{\theta}) \cos \theta \\ F &= a\theta(t - \tau) + b\dot{\theta}(t - \tau)\end{aligned}$$



Pitchfork bifurcation

Hopf bifurcation: oscillations for larger values of the control parameters a,b

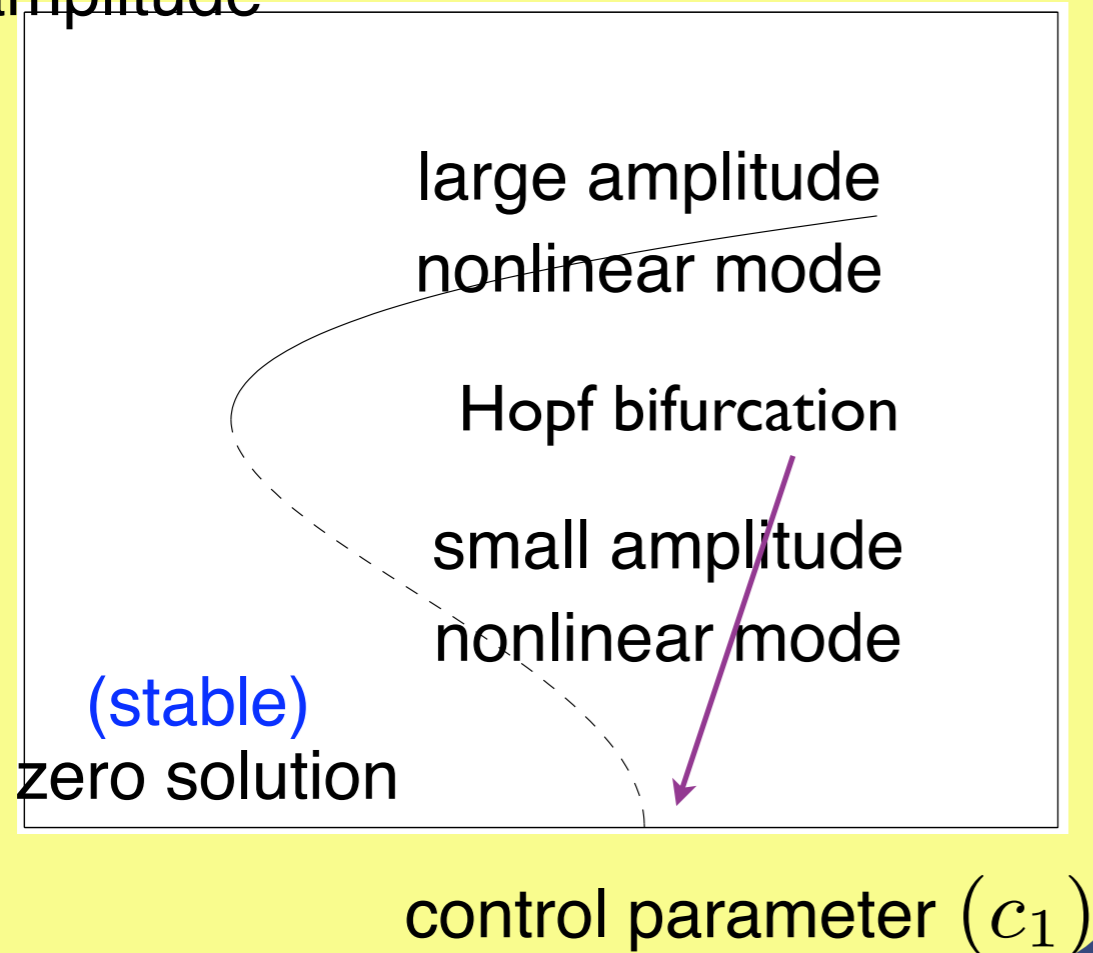
$\theta = 0$ is stable in shaded area, **D-shaped region**

Pitchfork bifurcation: stability of non-zero fixed point

Hopf bifurcation, stability of oscillatory behavior

Fully nonlinear model: w/ Hopf bifurcation

maximum
amplitude



Hopf bifurcation: oscillatory solutions for certain delay

Smaller nonlinear oscillatory solutions unstable

Larger nonlinear oscillatory solutions bi-stable with $x=0$ solution

Mathematical Challenges:

- Nonlinearity: non-uniqueness, complex dynamics
- Delays: Delay Differential Equations (DDEs)
- Noise: Stochastic DDEs (SDDEs), minimal theory, limited computational methods

Dynamical References: Campbell, Milton, Ohira, Sieber, Krauskopf, Stepan, K. others

SDDE's: $dx = f(x(t), x(t - \tau))dt + g(x(t), x(t - \tau))dW(t)$

- Dynamics: e.g. Mean Square Stability: Linear system w/ delays, Buckwar et al 2013
- Numerical methods

Milstein method ($O(h)$ strong convergence) for SDDE's:
Kloeden, Shardlow, 2012 (Taylor-like expansions)

Euler-Maruyama, ($O(h)$ weak convergence) for SDDE's:

Distributed delays:

Buckwar, et al, 2005 (linear), Clement, et al 2006

Anticipating (Malliavin) calculus: tame Ito formula on segments of solution

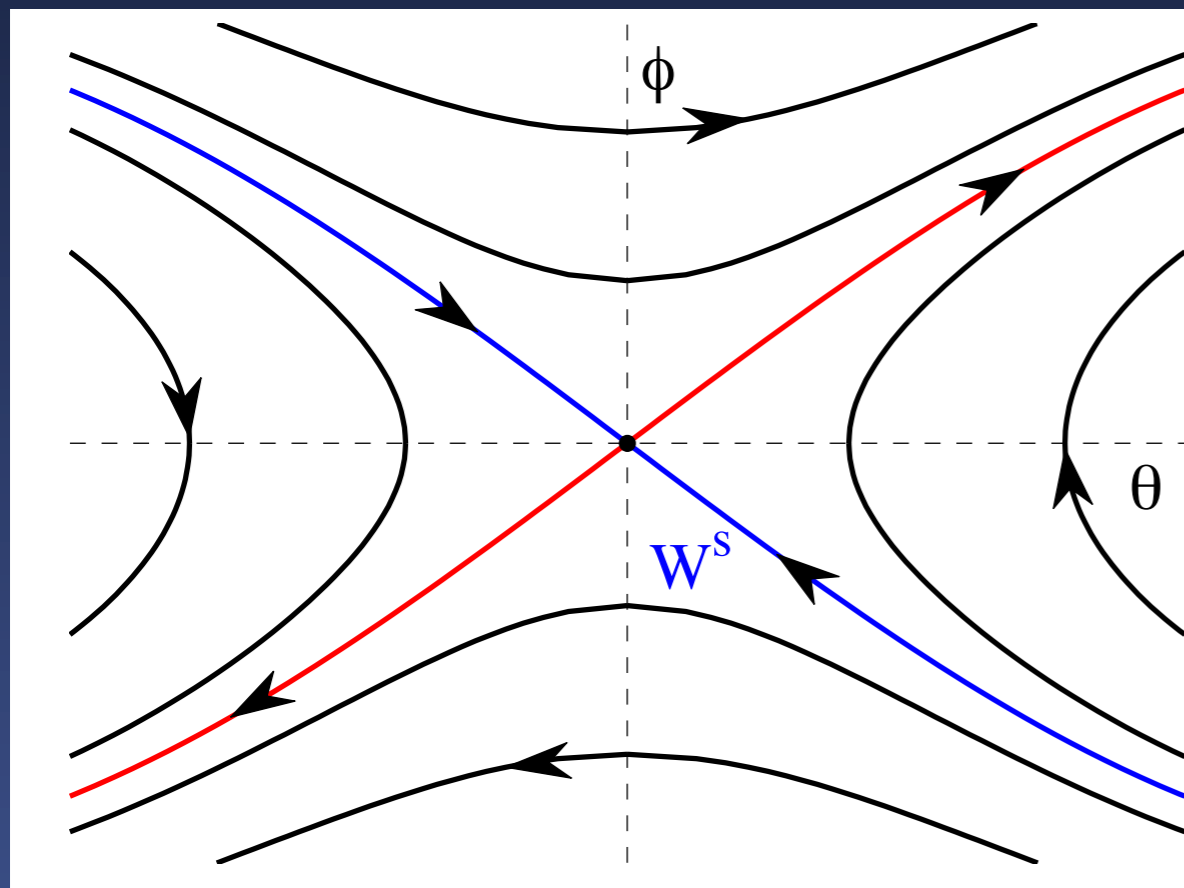
$$\int_{t_1}^{t_2} f^{(i)}(Y(u - \tau), Y(u)) \dots dW(u - \tau)$$

Buckwar, K., Mohammed, Shardlow, 2008

State-dependent control in balance models:

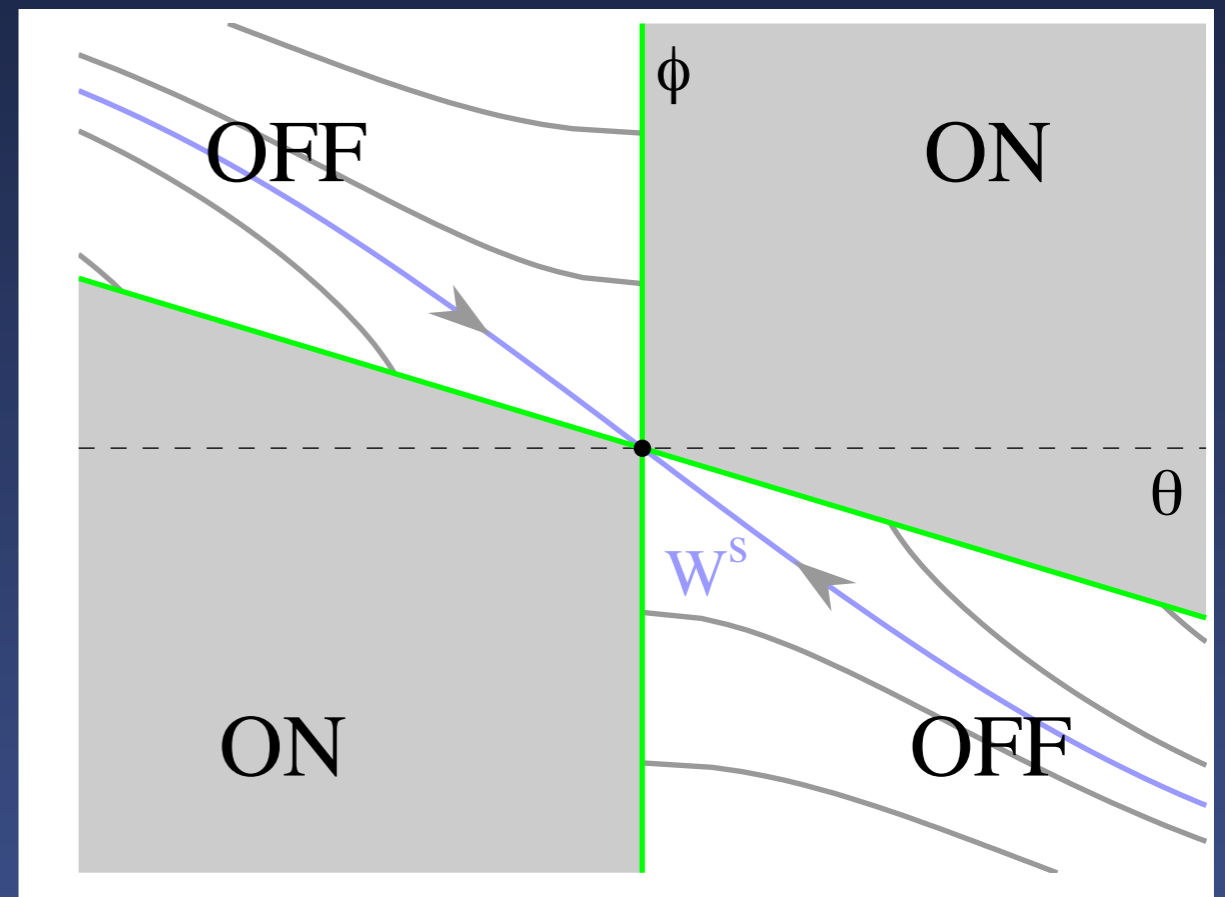
PD control with delay

Phase plane for system
with control off



$$\begin{aligned}\dot{\theta} &= \phi \\ \dot{\phi} &= \sin \theta - F(\theta, \dot{\theta}) \cos \theta \\ F &= a\theta(t - \tau) + b\dot{\theta}(t - \tau)\end{aligned}$$

State-dependent on/off control:
control is on in state drifting
away from origin



Asai, etal, 2009, PLOS I

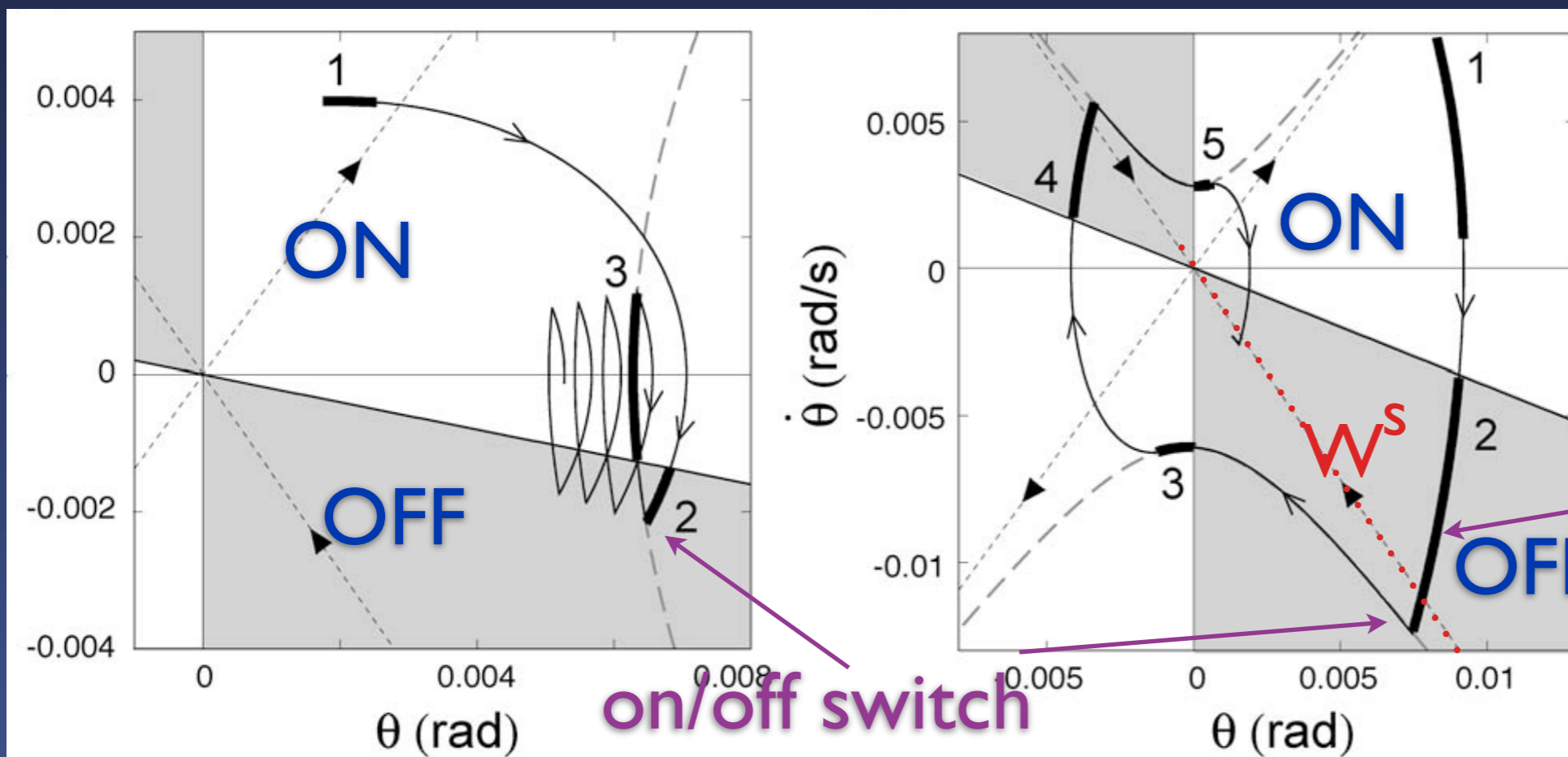
Zig-zags and spirals:

PD control with delay

$$\ddot{\theta} - \sin \theta + F(\theta, \dot{\theta}) \cos \theta = 0$$
$$F = a\theta(t - \tau) + b\dot{\theta}(t - \tau)$$

On/Off Control :
Note delay in switching from on/off, system enters off region before control is switched off

Weaker control: zig-zag behavior



Larger delay/
Stronger control:
yields spiral

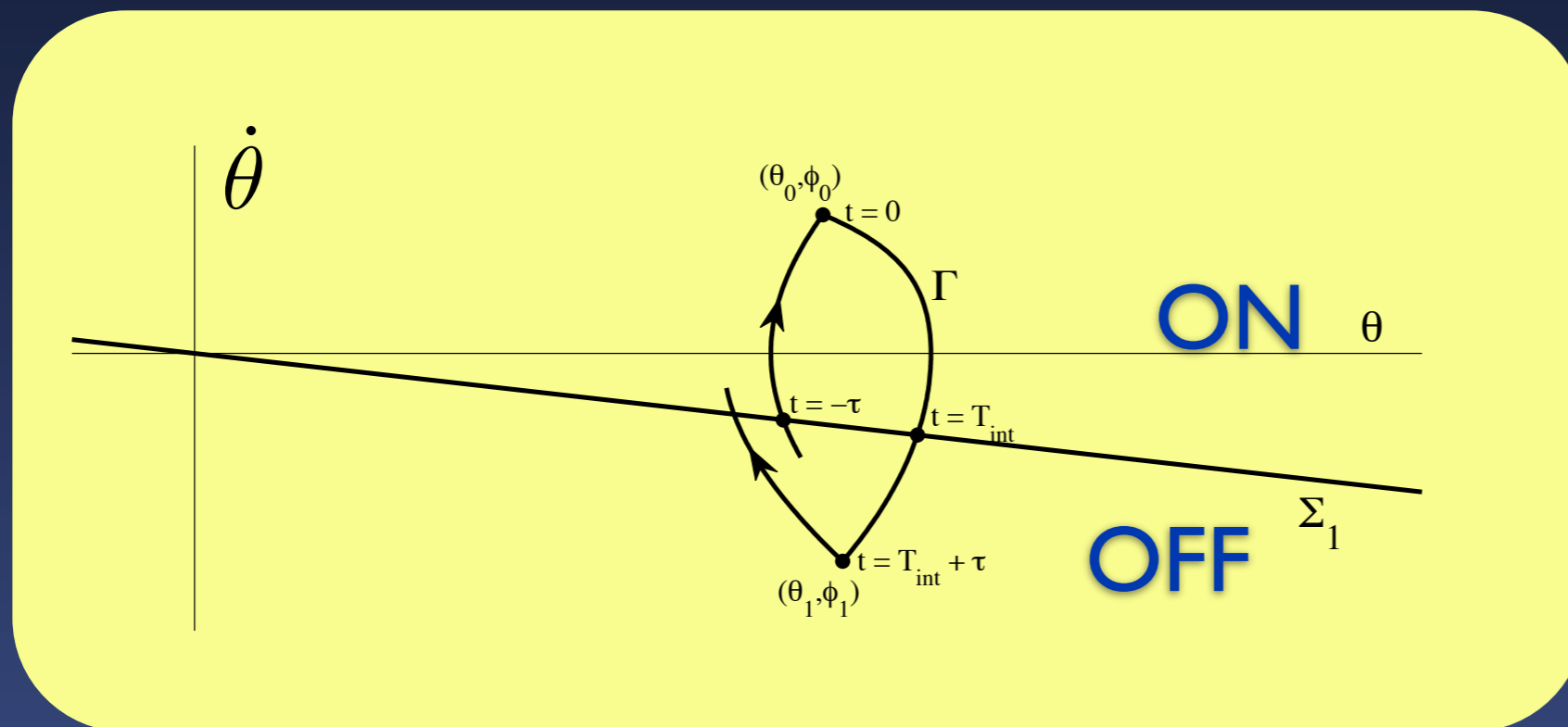
Asai, etal,
2009,
PLOS I

Periodic behavior:

$$\ddot{\theta} - \sin \theta + F(\theta, \dot{\theta}) \cos \theta = 0$$
$$F = a\theta(t - \tau) + b\dot{\theta}(t - \tau)$$

PD control with delay

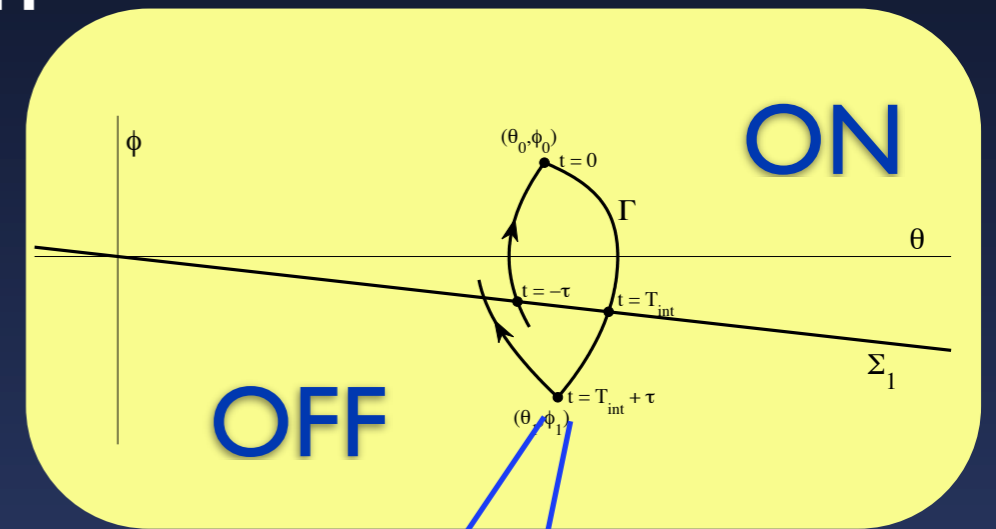
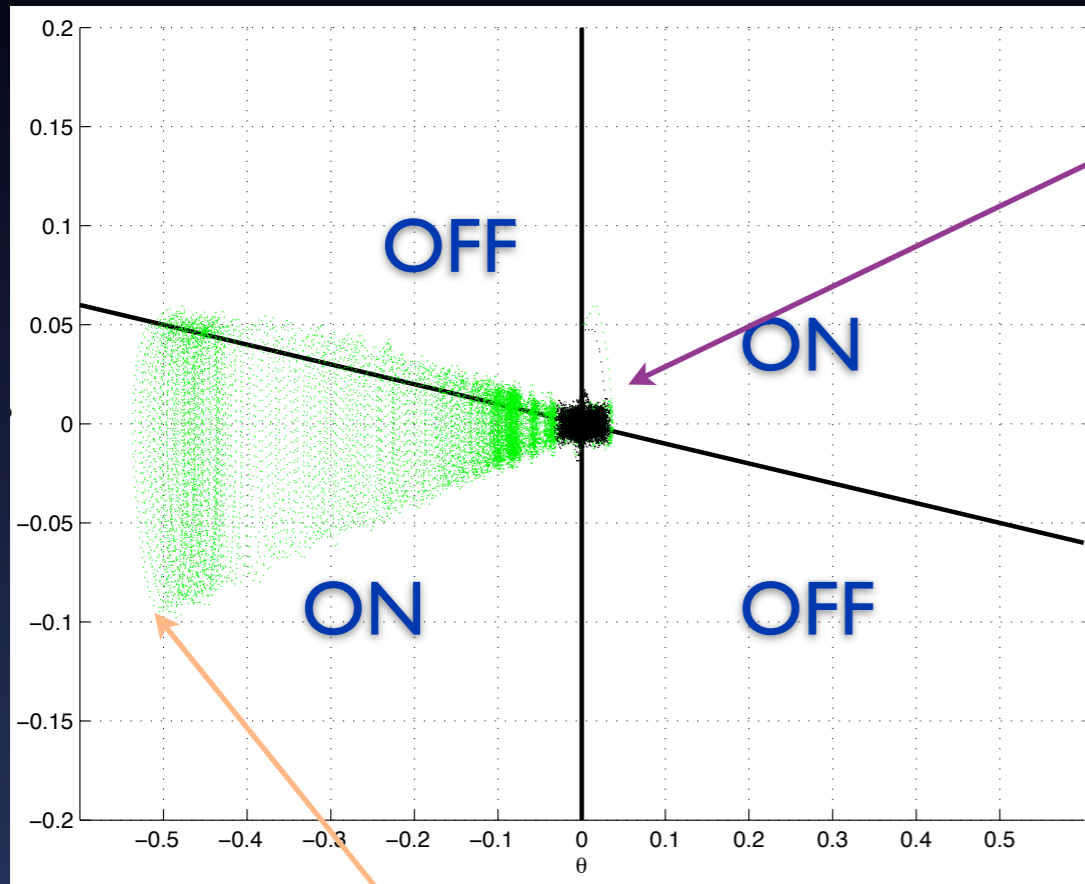
Zig-zag orbits - periodic solutions away from origin, moving back and forth from on/off



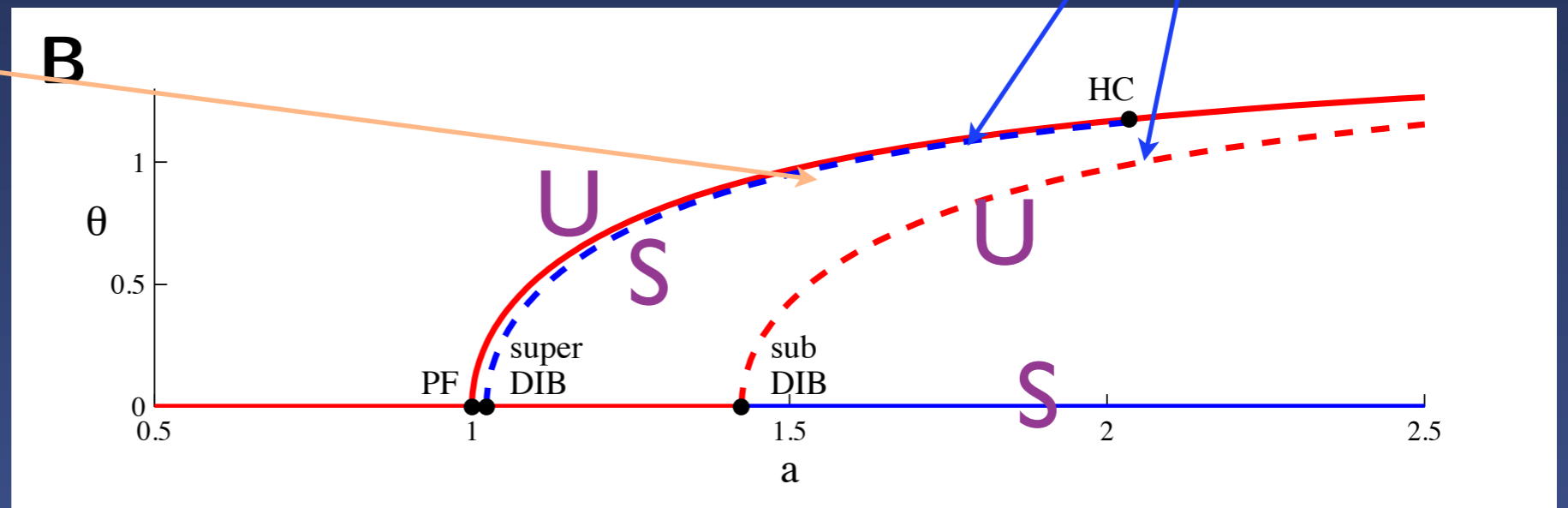
Note: w/o delay: PWS system is Filippov, solution slides along the switching manifold, can calculate analytically

Balance model w/ noise: small random fluctuation

For larger control:
oscillations near the origin:
weak attraction to the origin

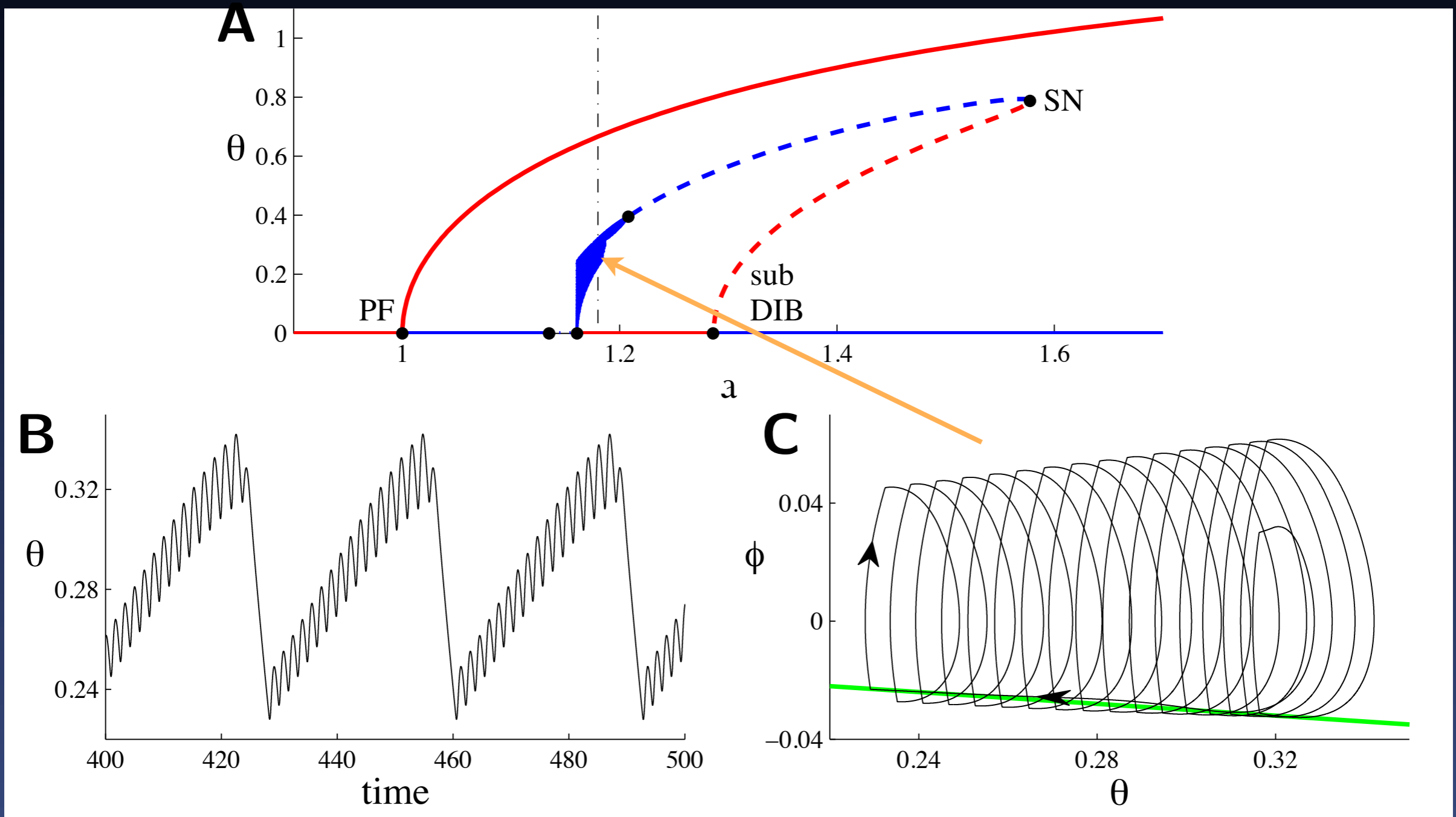


For smaller control, noise can cause transition to zig-zag orbit



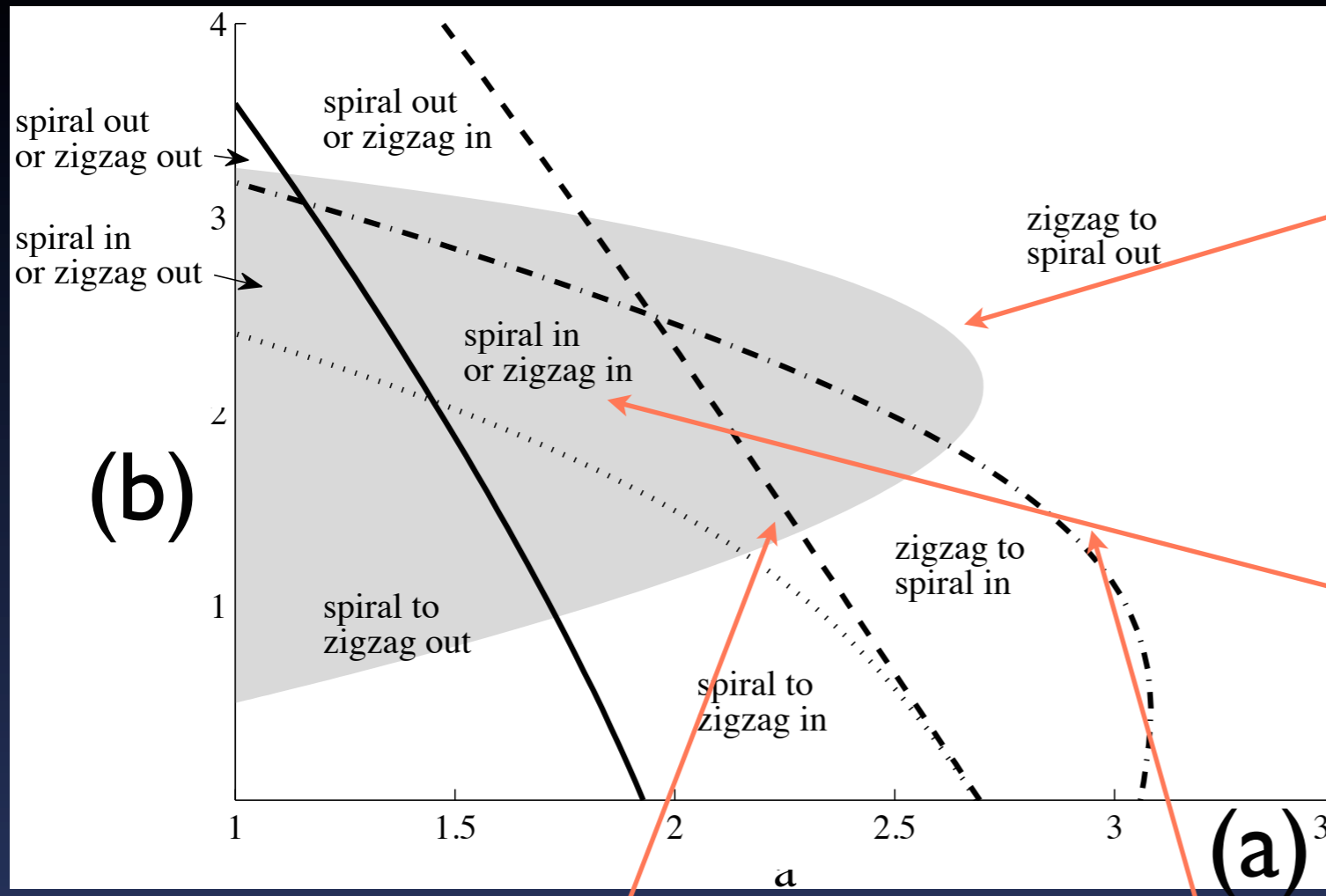
Bifurcation diagram: control a vs. θ

Experimental θ Bursting-like oscillations



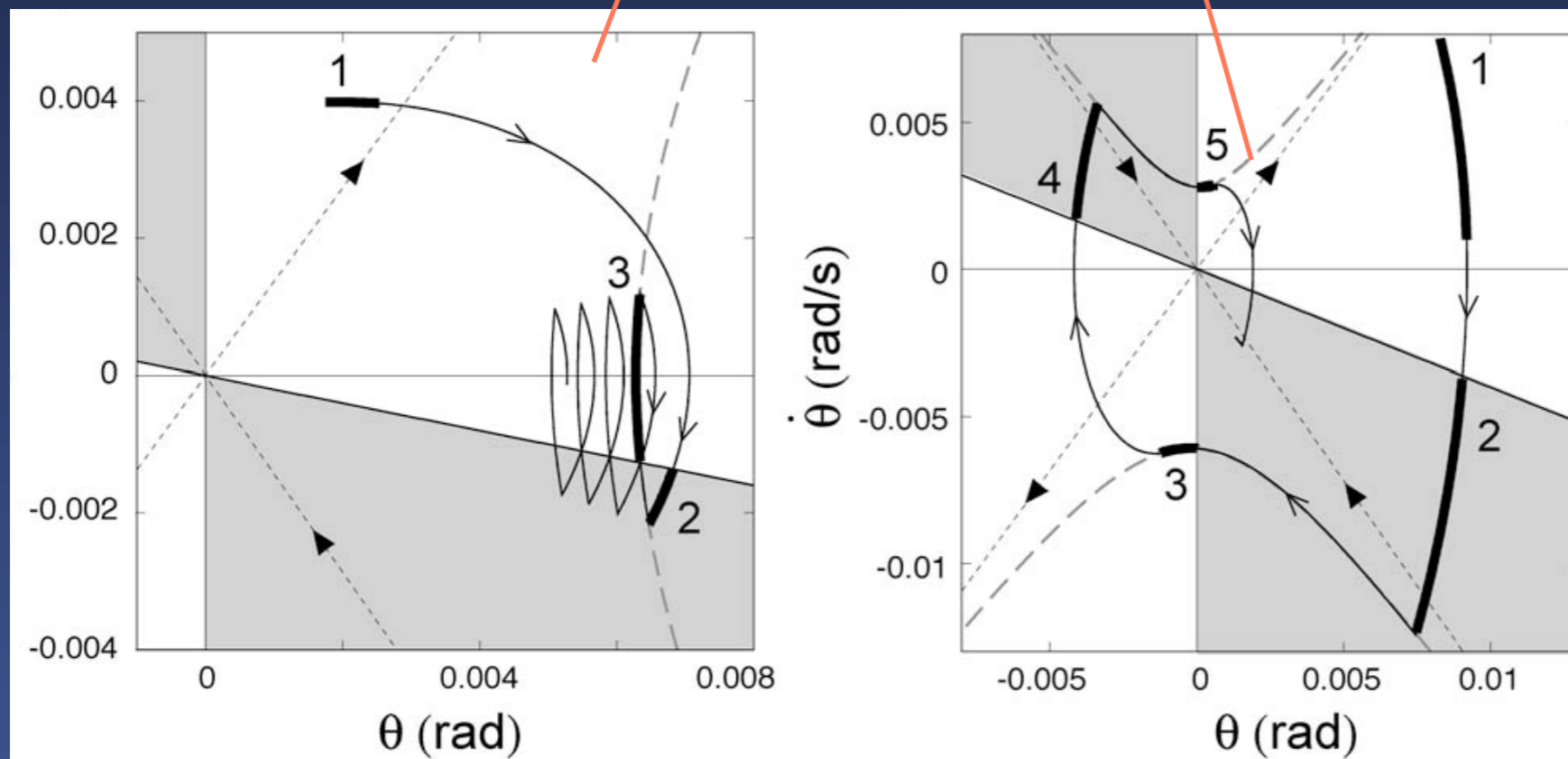
Also prominent in cart model with significant mass

Transitions from zig-zag to spirals?



D-shaped region:
stable vertical position
with cts control

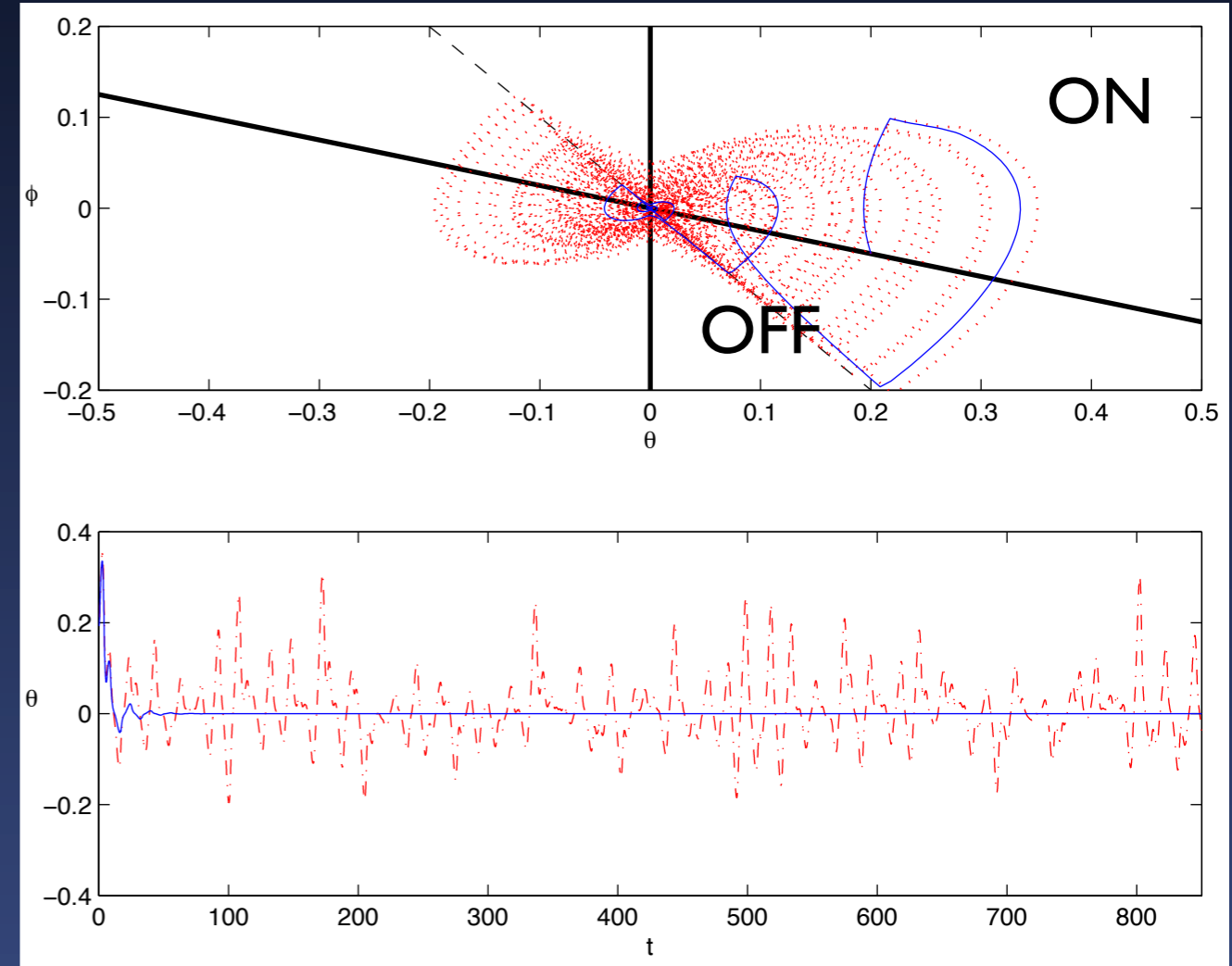
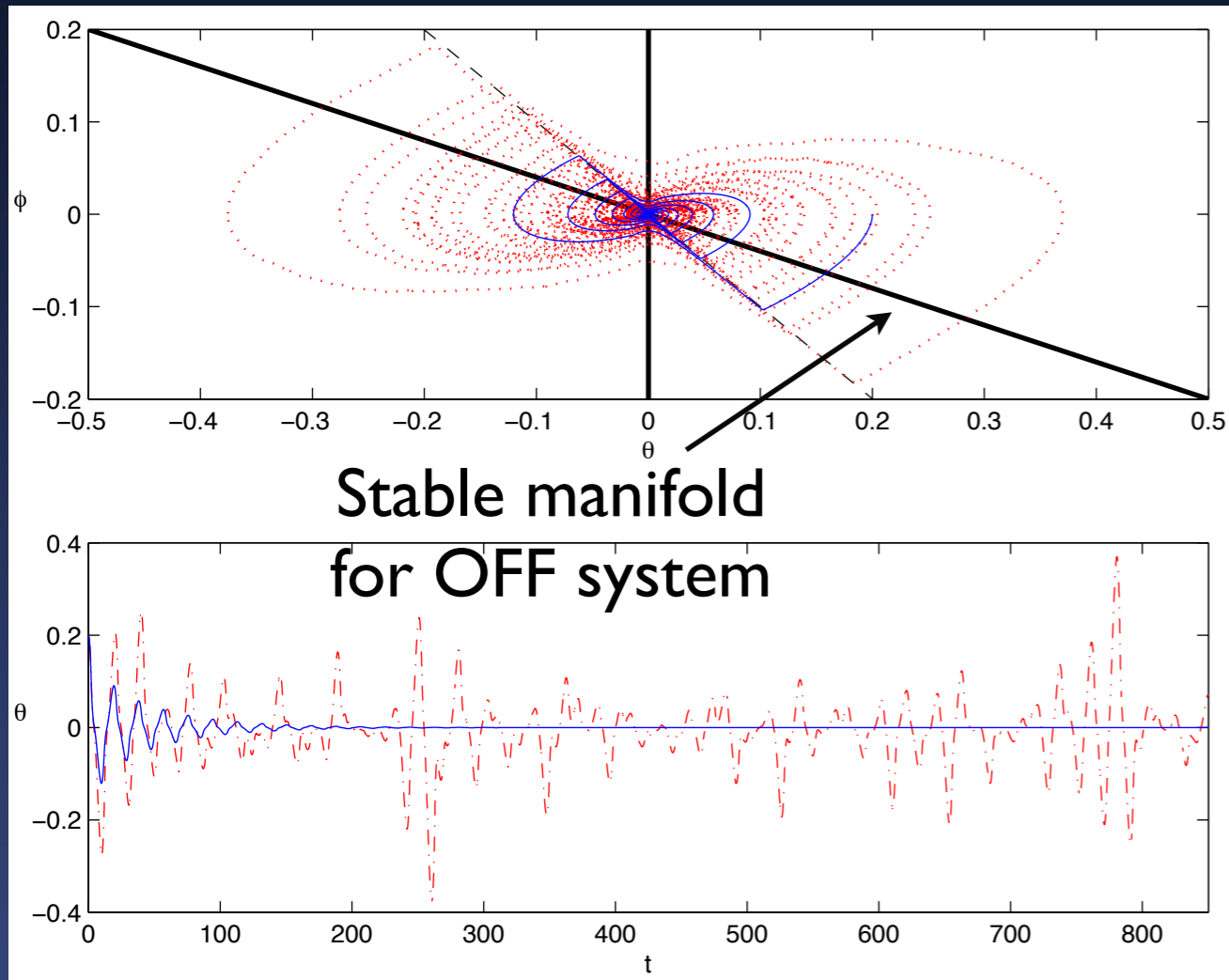
Regions: analysis of
linearized model



(unstable)
zigzag and
spirals w/
discts control

Transitions: sensitivity to noise

Noise driven/amplified spirals, stabilized transients:
complex dynamics expected for larger systems



unstable spiral, sustained by noise
- similar to perturbed spiral

unstable ZZ sustained as
spiral via noise

Results for stochastic + discontinuous dynamics

Discontinuous noise sources : Jump-diffusion: R-K methods: Mean square (strong) convergence and Lipschitz-type conditions for increment functions
Buckwar, et al 2011

Stochastic Flows for SDE's with singular coefficients: Many results carry-over for non-smooth, measurable drifts, as long as noise is "nice" (e.g. Brownian) Mohammed et al 2013

This result is counter-intuitive since the dominant 'culture' in stochastic (and deterministic) dynamical systems is that the flow 'inherits' its spatial regularity from the driving vector fields

Results for stochastic + discontinuous dynamics

Smoothed systems (approximating discrete system):

Elemgard et al 2013, Simonsen et al 2013

Deterministic:

Acary, Brogliato, Numerical Methods for Nonsmooth, DS, 2008

Dieci, Lopez, Survey for IVPs, 2012

Weak existence and uniqueness vs. strong uniqueness,
discontinuous drift bounded away from zero Pascu, 2013

Smooth vs. Piecewise Linear (PWL) models: Complex dynamics in higher dimensions or noise driven?

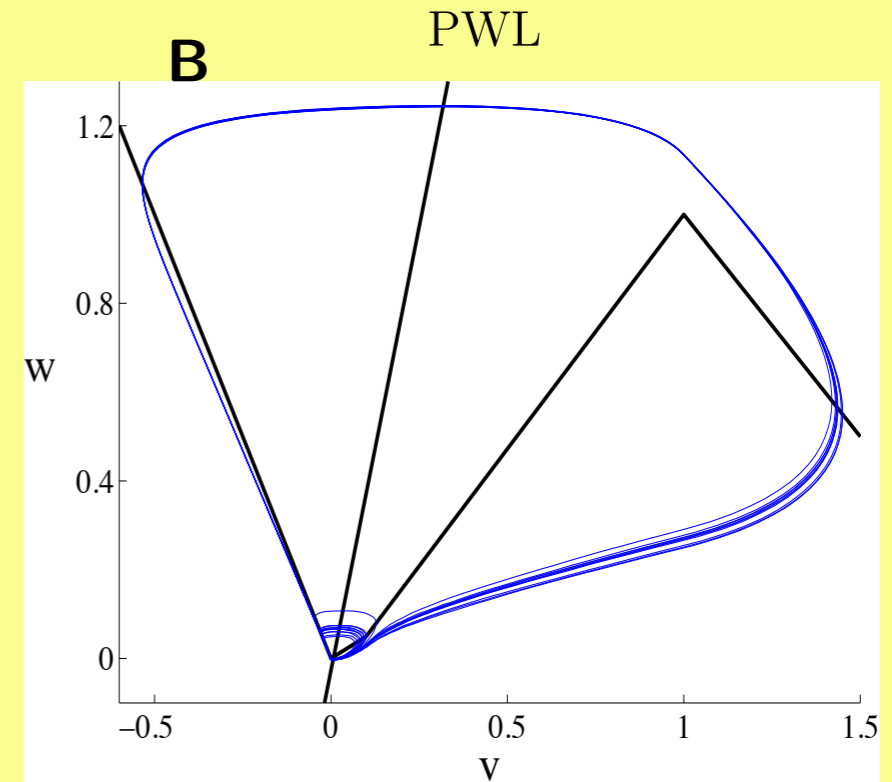
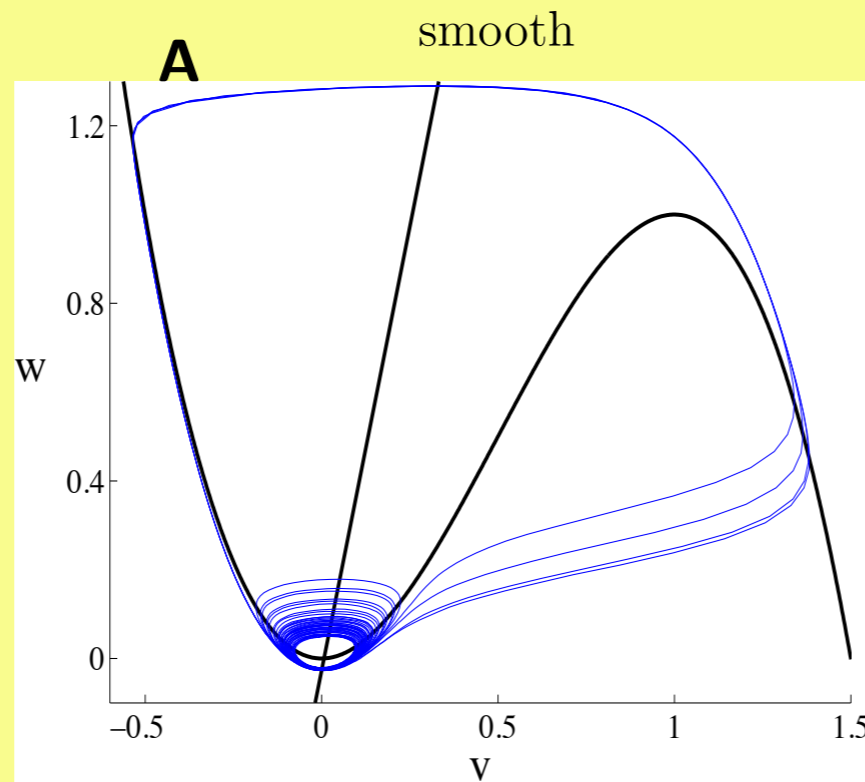
Neuro-dynamics: Typical structure

$$dv = (f(v) - w) dt ,$$

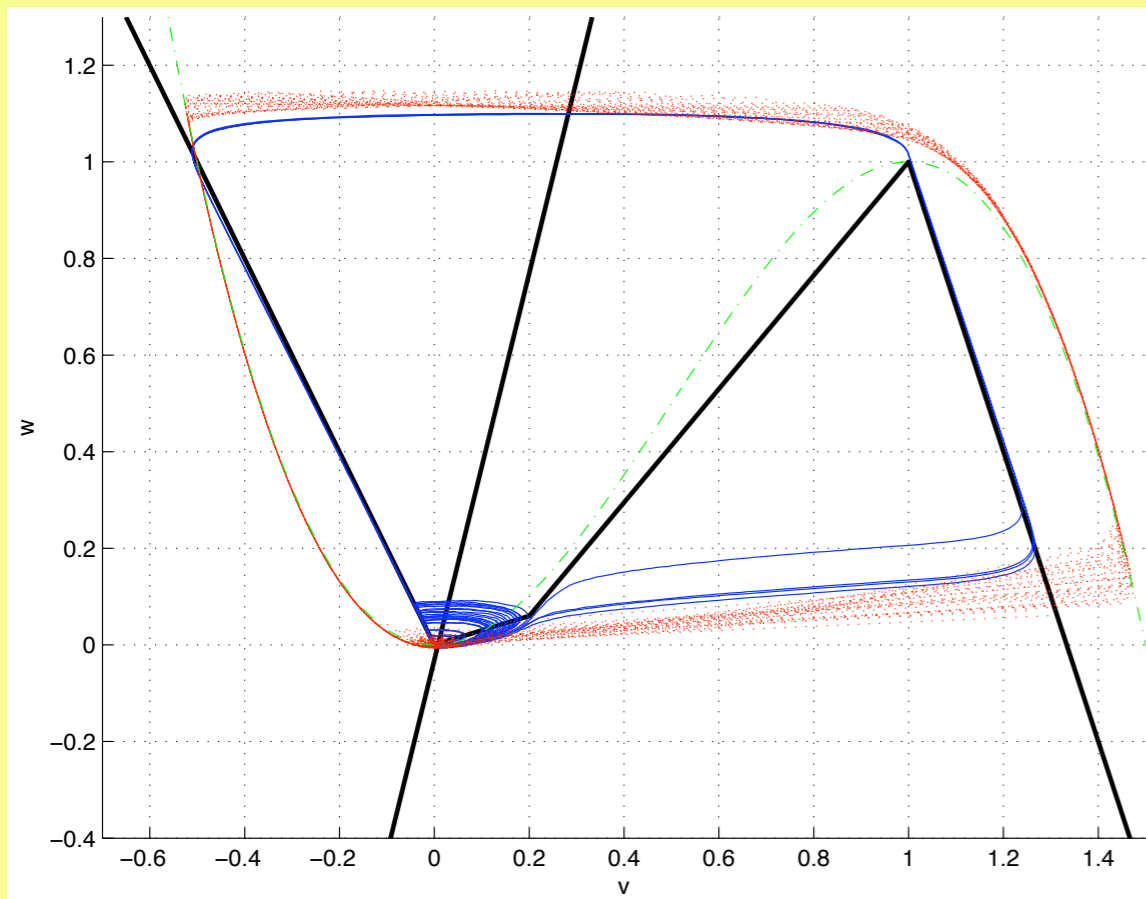
$$dw = \varepsilon(\alpha v - \sigma w - \lambda) dt + D dW$$

$$f(v) = \begin{cases} \eta_L v , & v \leq 0 \\ \eta_1 v , & 0 < v \leq v_1 \\ \eta_2(v - v_1) + w_1 , & v_1 < v \leq 1 \\ \eta_R(v - 1) + 1 , & v > 1 \end{cases}$$

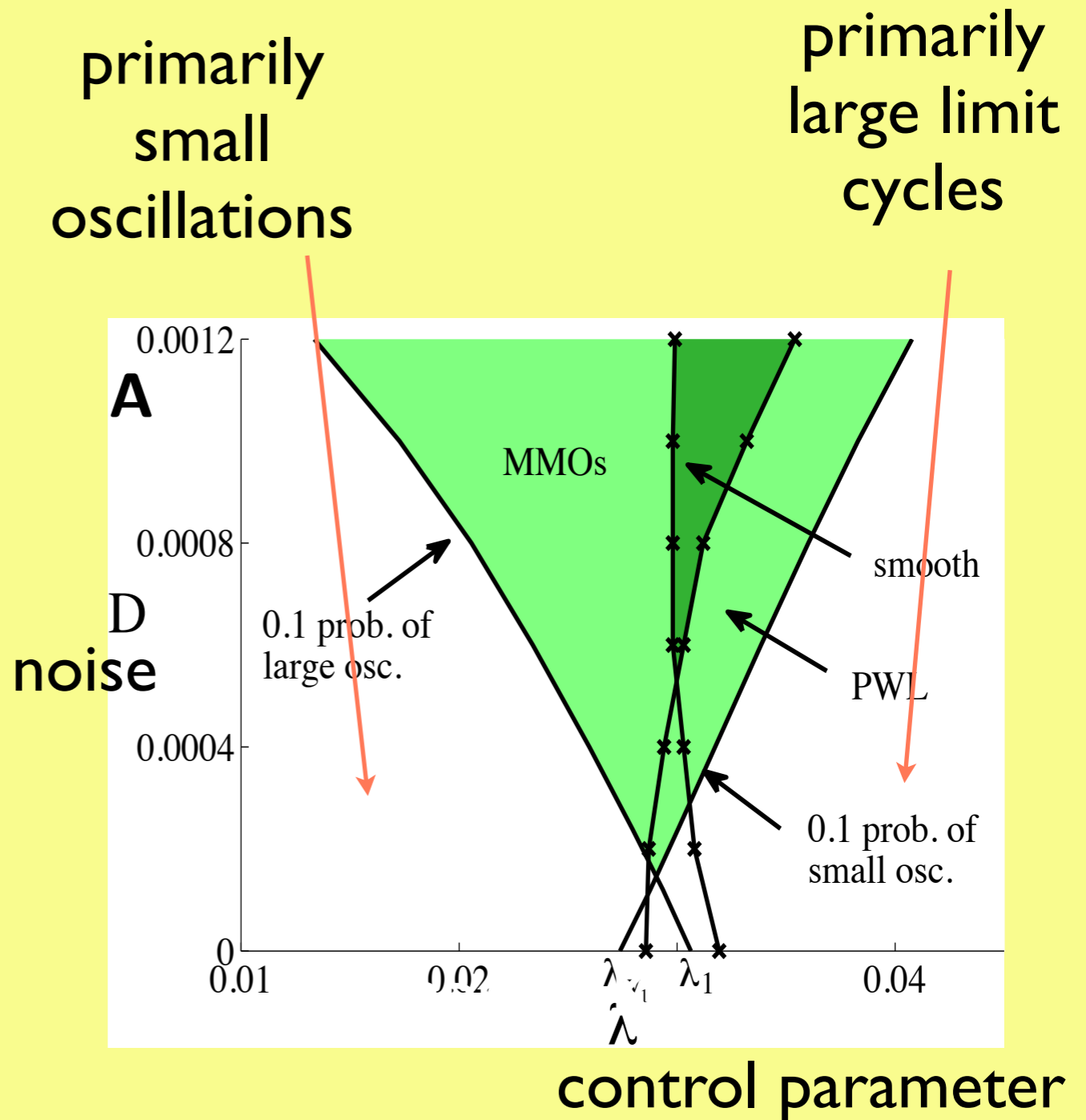
$$f(v) = 3v^2 - 2v^3$$



MMO's in PWL models w/ canard structure



MMO's robust over a larger range of control parameter for PWL



Simpson, K. Physica D, 2011

Analysis of PWL FHN type models, w/ noise

- Linear analysis on subregions
- Time dependent probabilities

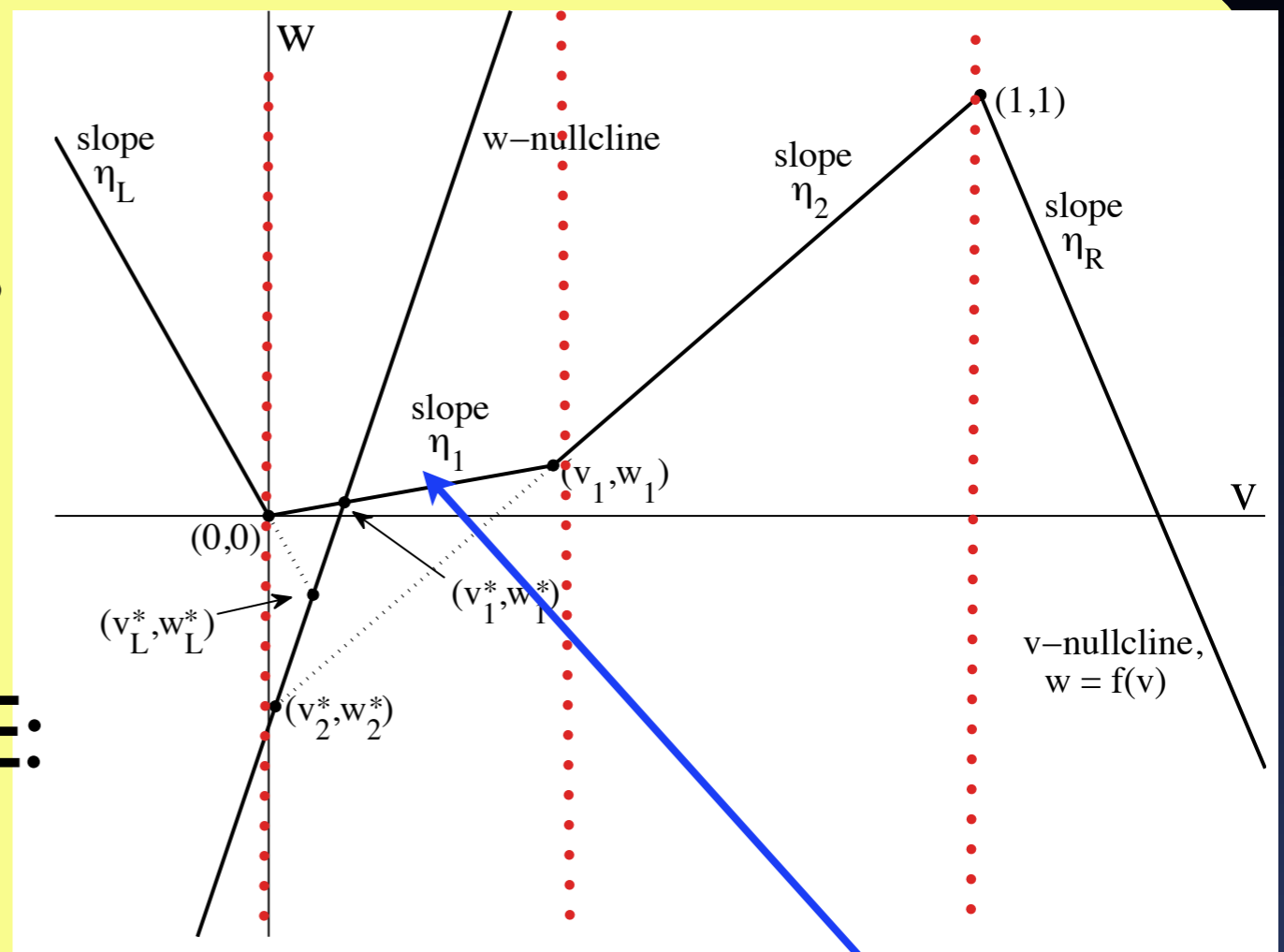
In each for four regions, SDE:

$$dv = (f(v) - w) dt ,$$

$$dw = \varepsilon(\alpha v - \sigma w - \lambda) dt + D dW$$

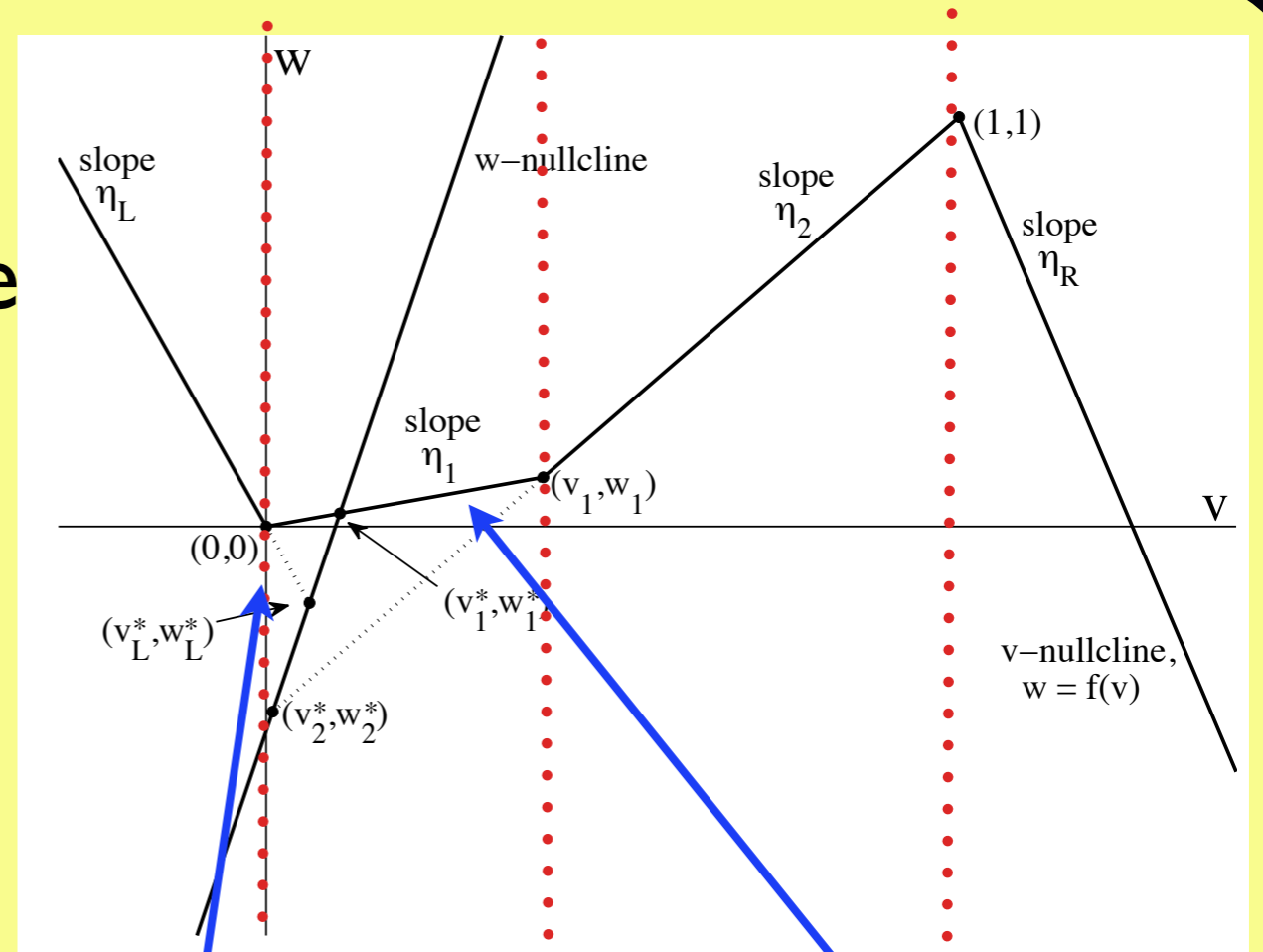
Time dependent probability density with $f(v)$ linear
Solve PDE (Fokker Planck equation) with linear
coefficients: Gaussian probabilities

Time dependent Ornstein-Uhlenbeck processes

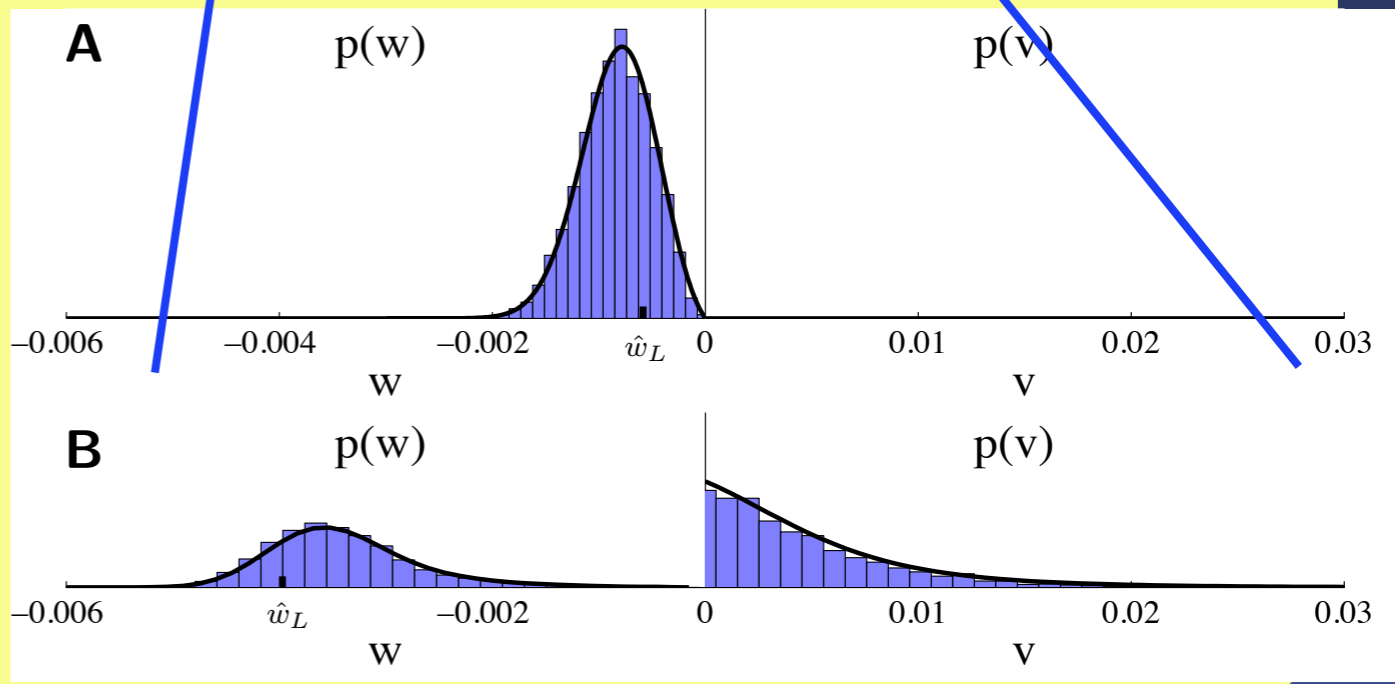
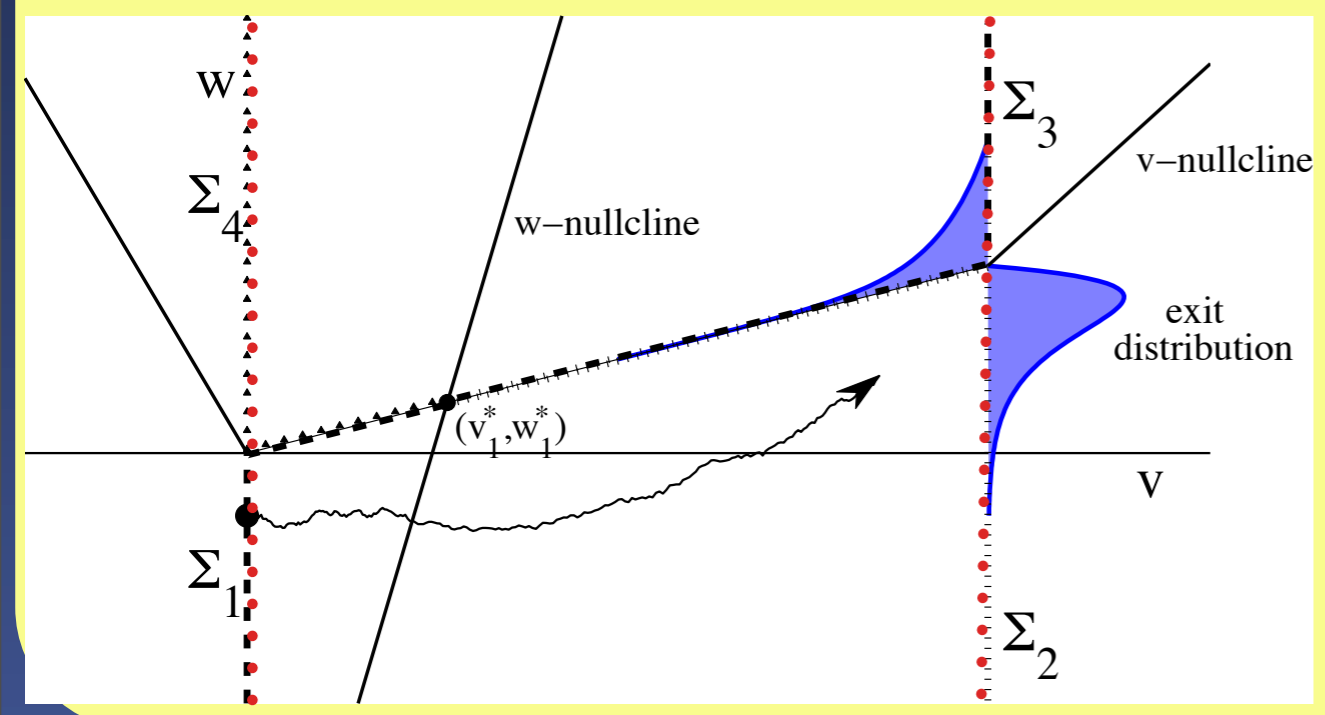


Time dependent densities: transition distributions

- Time dependent probabilities
- Semi-analytical approach: iterate on transitions between regions
- Additional local analysis at crossings: first vs. last crossing time
- Parametric dependence



Variability with λ and critical values v_1, w_1



Sliding dynamics: Relay control

$$\dot{\mathbf{x}} = A\mathbf{x} + B\mathbf{u}$$

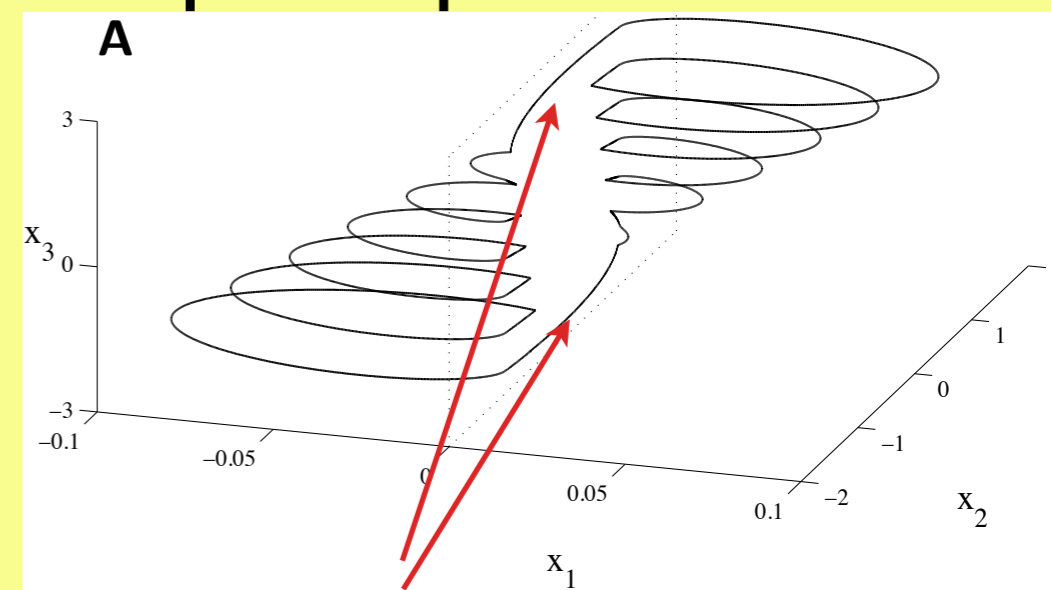
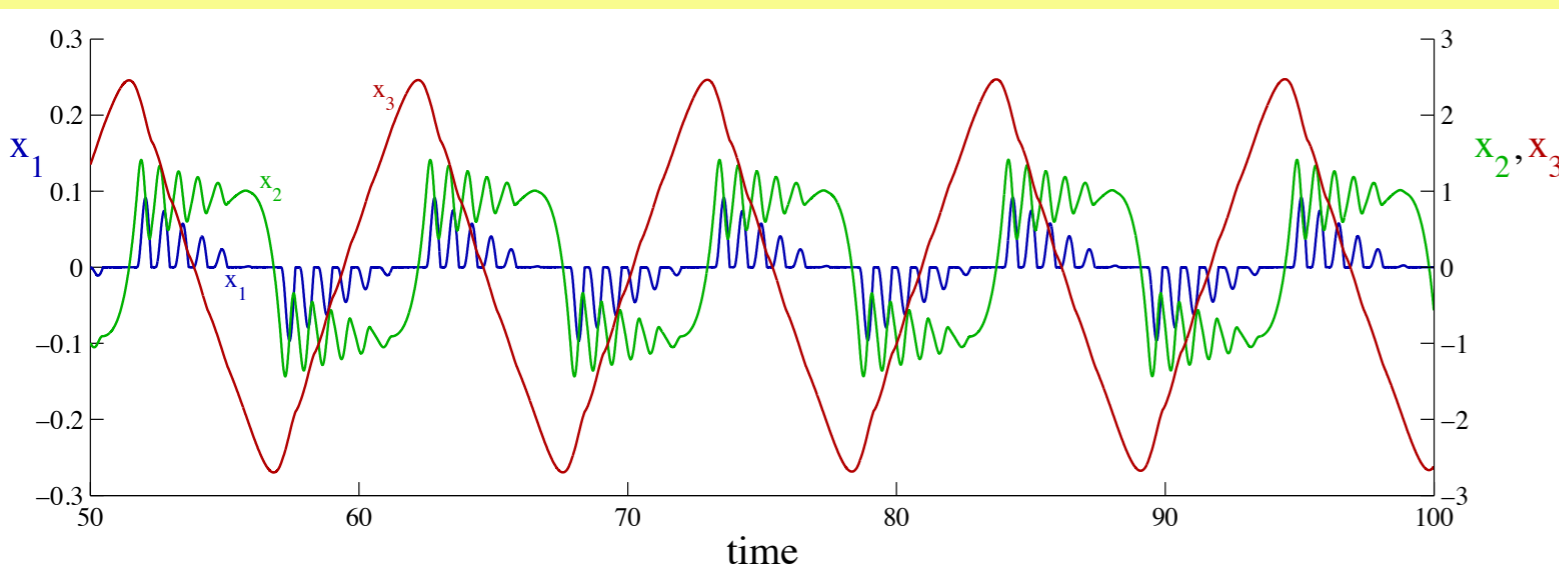
$$\varphi = C^T \mathbf{x},$$

$$\mathbf{u} = -\text{sgn}(\varphi)$$

Control (\mathbf{u}) depends on state \mathbf{x}

no noise

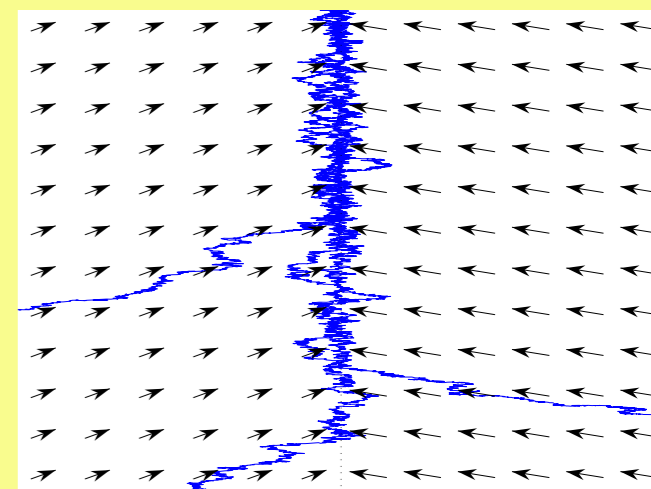
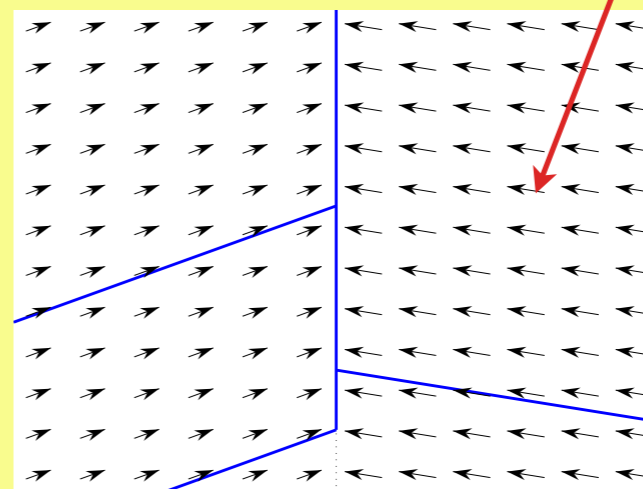
phase plane



sliding

w/noise

vector field is discontinuous along the switching (sliding) manifold



Potential contributions to $O(I)$ change to average period

$$\dot{\mathbf{x}} = A\mathbf{x} + B\mathbf{u}$$

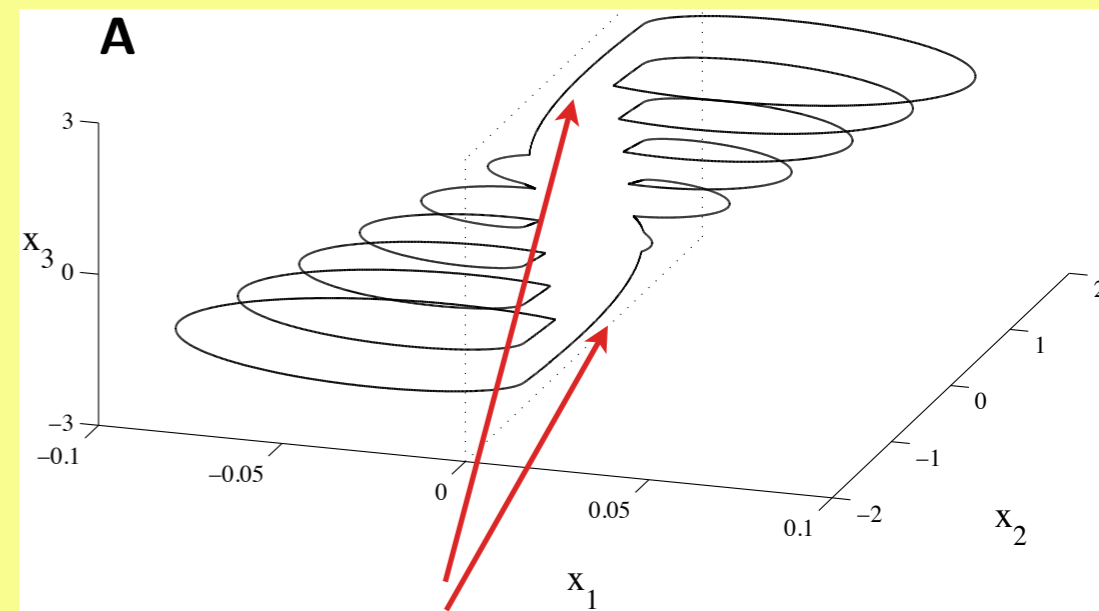
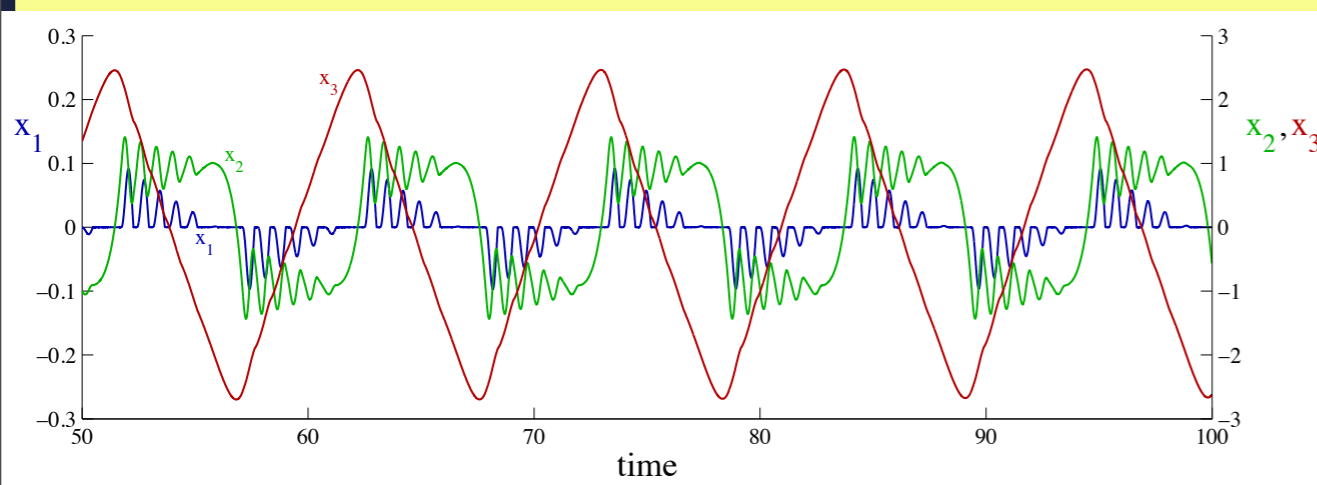
$$\varphi = C^T \mathbf{x},$$

$$\mathbf{u} = -\text{sgn}(\varphi)$$

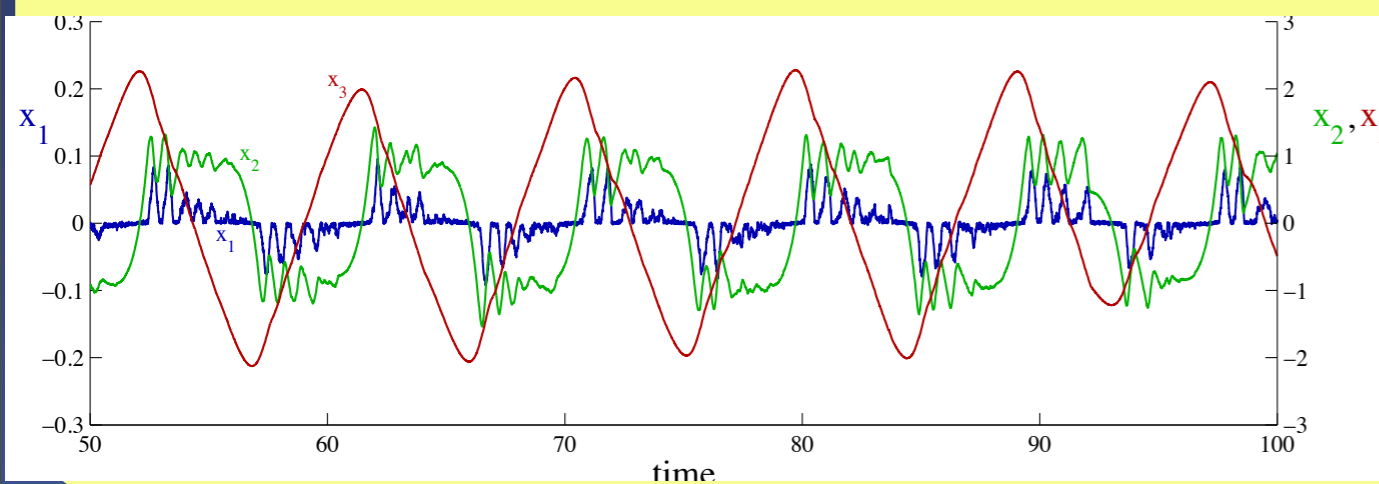
Control (\mathbf{u}) depends on state \mathbf{x}

phase plane

no noise

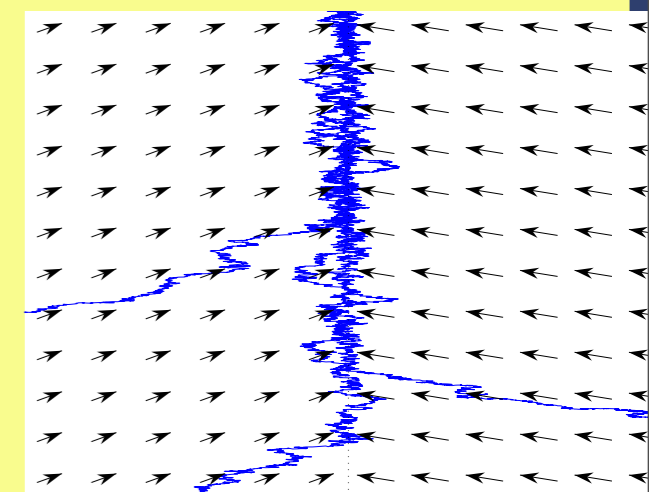
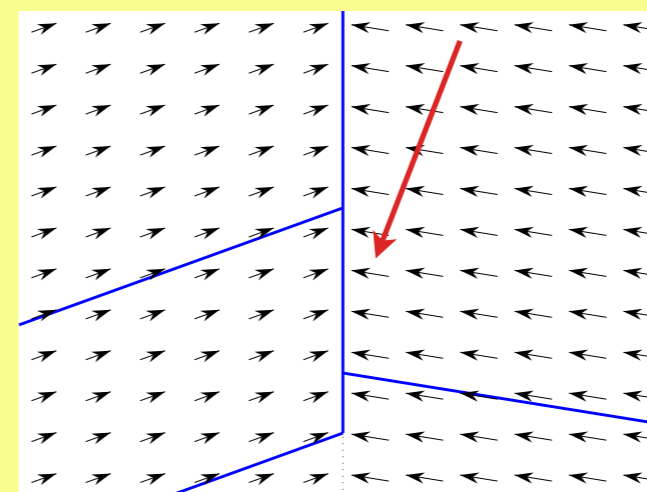


w/ noise



sliding

w/noise

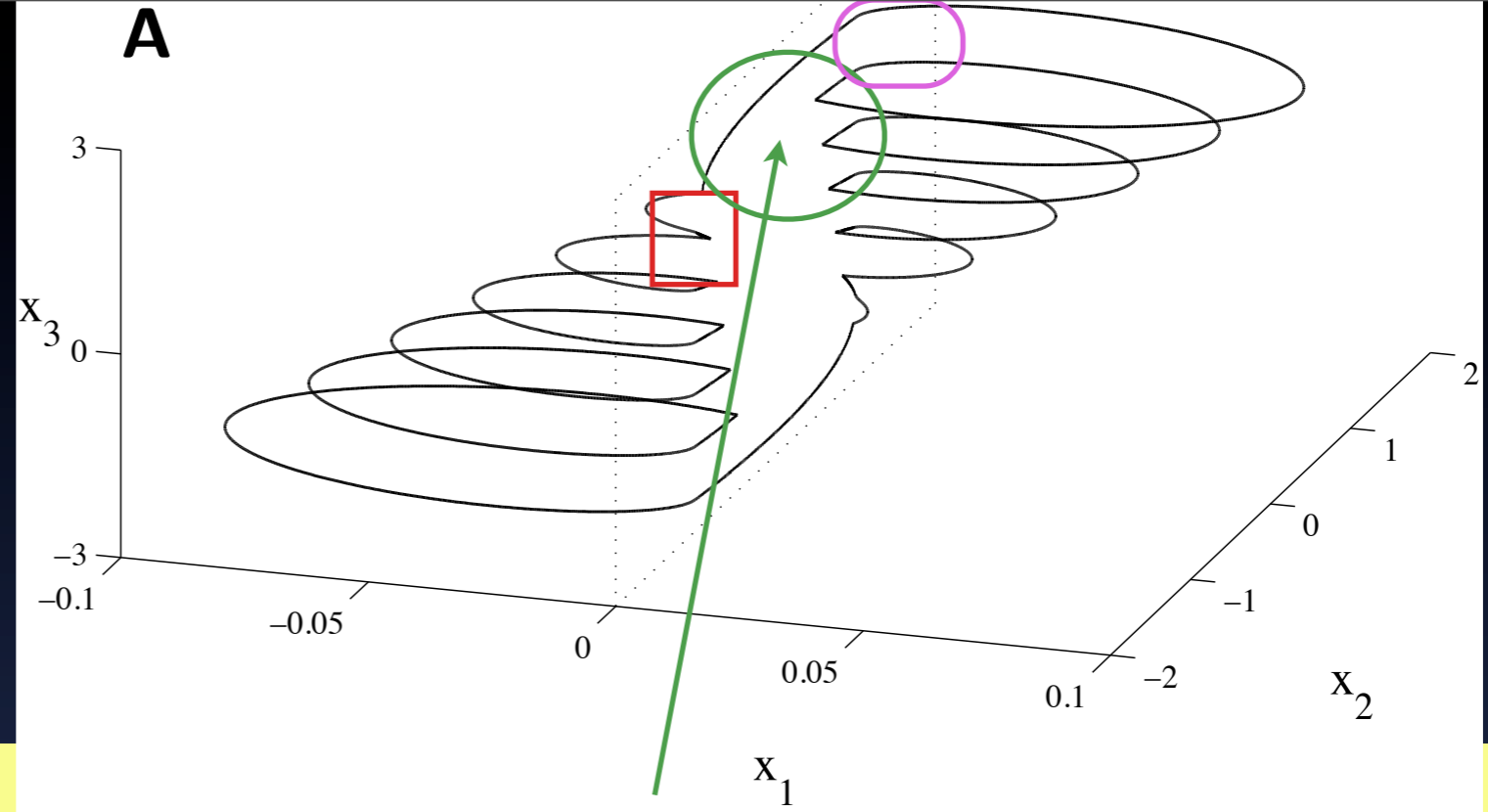


3D model

$$\dot{\mathbf{x}} = A\mathbf{x} + Bu$$

$$\varphi = C^T \mathbf{x},$$

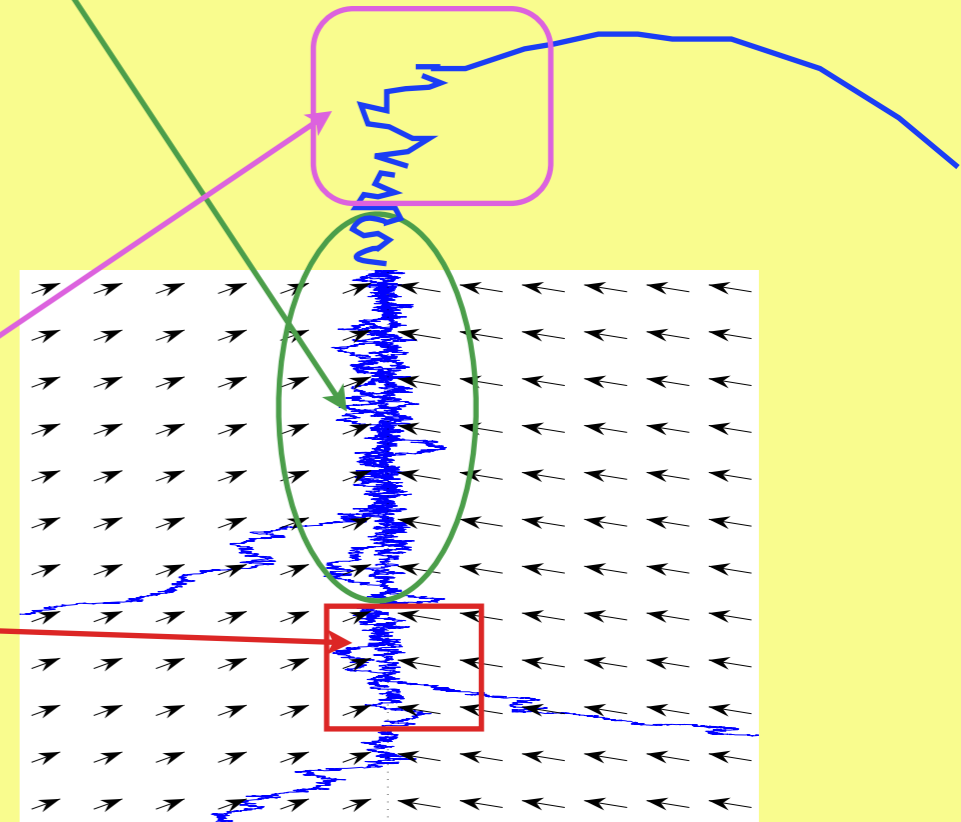
$$u = -\text{sgn}(\varphi)$$



Results: influence of noise in deviation from sliding

Other regions: connection of sliding with other dynamics (exit/entrance), use of stochastic averaging, asymptotic analysis of FPE

sliding



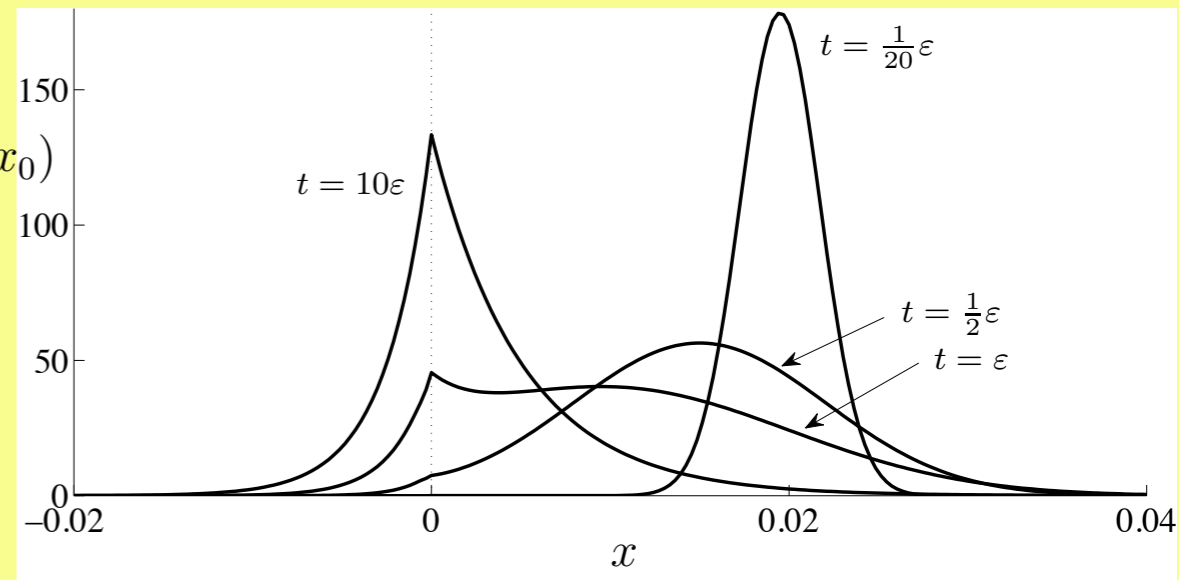
Distributions, moments

Constant drift case:

$$\text{Var}(y(t)) = \varepsilon \kappa t + \frac{(b_L - b_R)^2}{(a_L + a_R)^2} \varepsilon t + O(\varepsilon^2)$$

$p_\varepsilon(x, t; x_0)$

Time dependent density:
 Needed to compute correlations
 (Karatzas, Shreve)



$p(x, t | x_0) =$

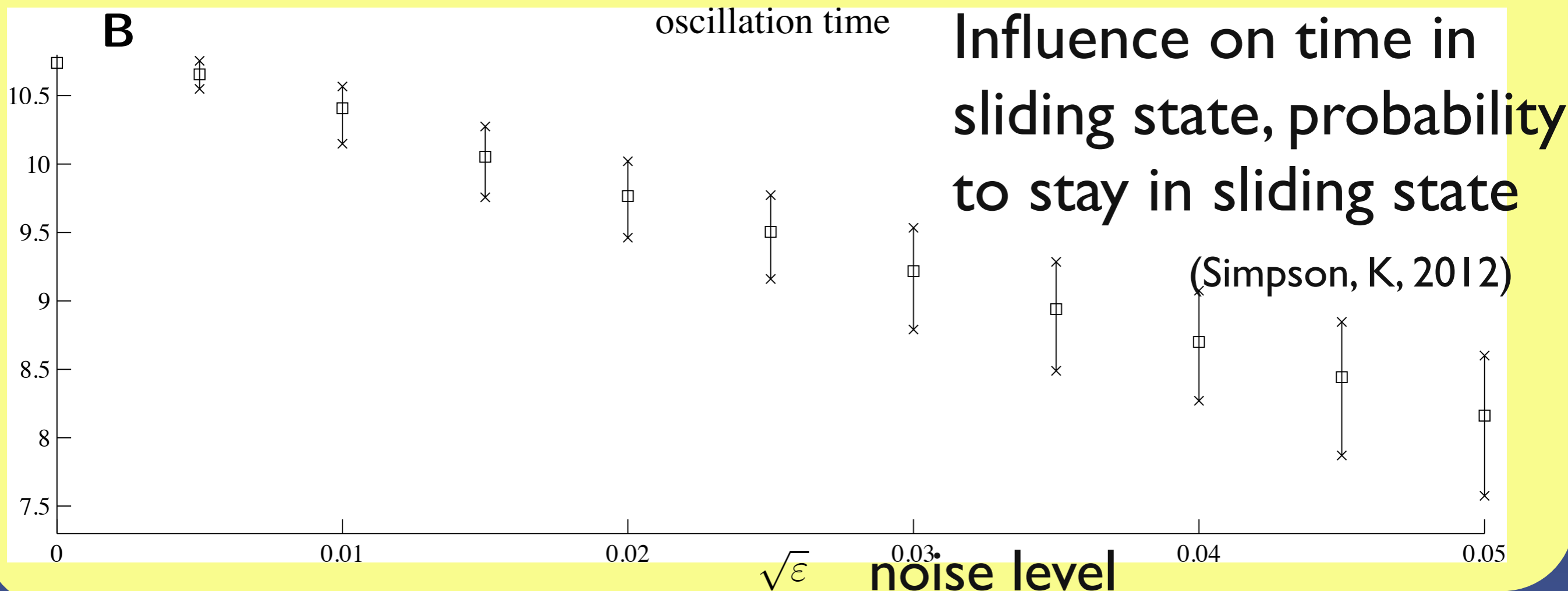
$$\begin{cases} \frac{2}{\varepsilon} e^{\frac{2a_L x}{\varepsilon}} \int_0^\infty h_\varepsilon(t, b, a_R) * h_\varepsilon(t, b - x - x_0, a_L) db + G_{\text{absorb}, \varepsilon}(x, t, a_L | x_0), & x_0 \leq 0, x \leq 0 \\ \frac{2}{\varepsilon} e^{\frac{-2a_R x}{\varepsilon}} \int_0^\infty h_\varepsilon(t, b + x, a_R) * h_\varepsilon(t, b - x_0, a_L) db, & x_0 \leq 0, x \geq 0 \\ \frac{2}{\varepsilon} e^{\frac{2a_L x}{\varepsilon}} \int_0^\infty h_\varepsilon(t, b + x_0, a_R) * h_\varepsilon(t, b - x, a_L) db, & x_0 \geq 0, x \leq 0 \\ \frac{2}{\varepsilon} e^{\frac{-2a_R x}{\varepsilon}} \int_0^\infty h_\varepsilon(t, b + x + x_0, a_R) * h_\varepsilon(t, b, a_L) db + G_{\text{absorb}, \varepsilon}(x, t, -a_R | x_0), & x_0 \geq 0, x \geq 0 \end{cases}$$

Implications for (near) sliding dynamics

$$\langle y(t) \rangle = y_{\text{slide}}(t) + \frac{(a_L^2 d_R - a_R^2 d_L)(a_L + a_R) - (a_L^2 c_R - a_R^2 c_L)(b_L - b_R)}{2a_L a_R (a_L + a_R)^2} \varepsilon t + o(\varepsilon)$$

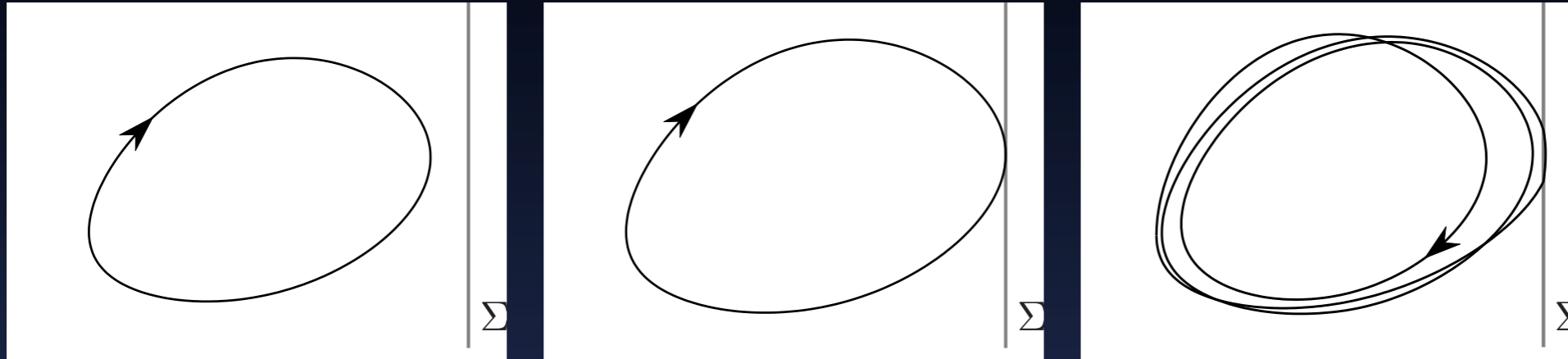
$$\text{Var}(y(t)) = \varepsilon \kappa t + \frac{(b_L - b_R)^2}{(a_L + a_R)^2} \varepsilon t + O(\varepsilon^2)$$

Standard deviation larger than mean shift from sliding
Large coefficients can shift average oscillation time

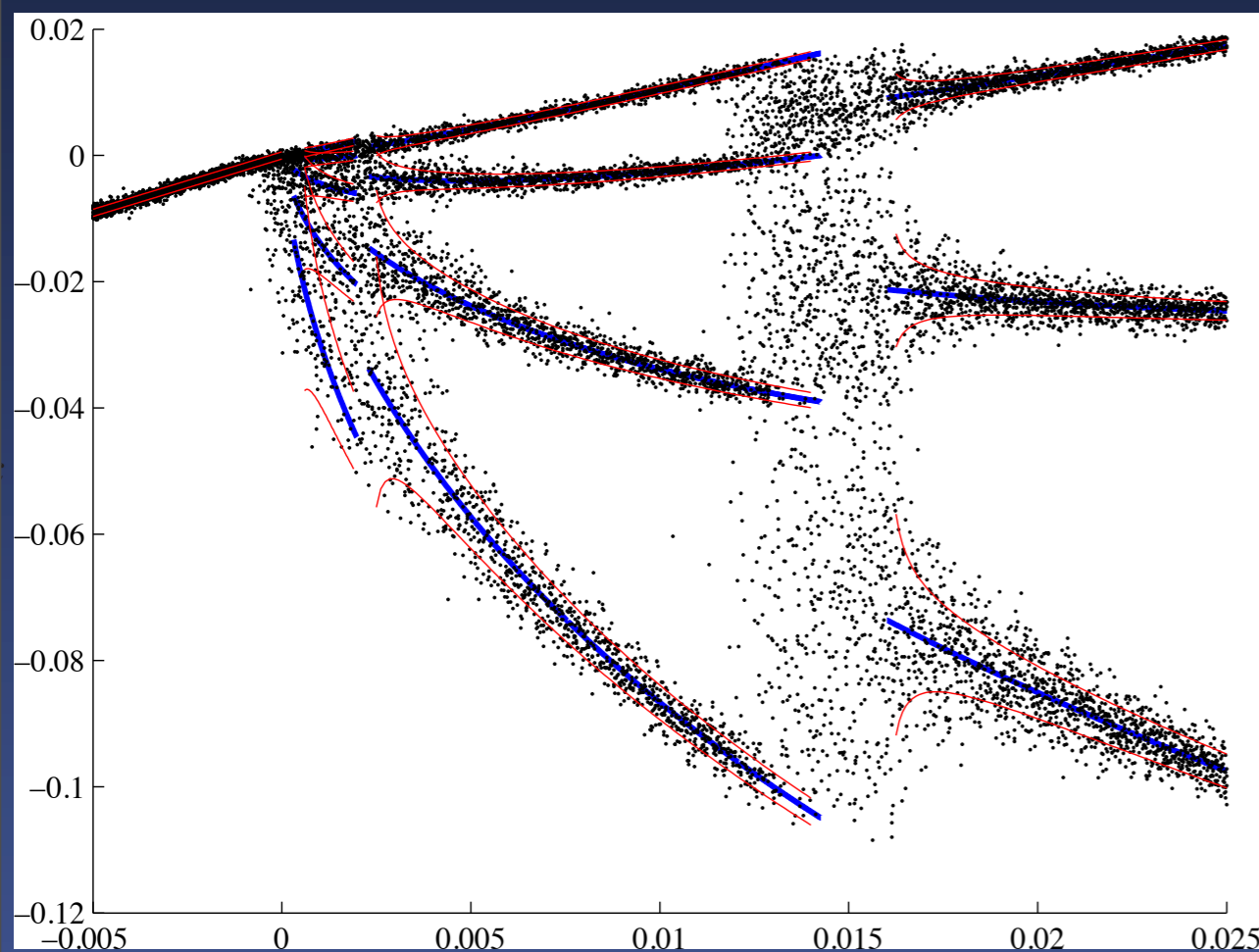


Noise sensitivity: grazing

Grazing: vibro-impacts, friction, AFM, stick-slip



Poincare map: a discontinuity map captures impact



Grazing normal form:
Nordmark map
Square root behavior

$$\begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{cases} A \begin{bmatrix} x \\ y \end{bmatrix} + \begin{bmatrix} 0 \\ 1 \end{bmatrix} \mu, & x \leq 0 \\ A \begin{bmatrix} x \\ y - \chi\sqrt{x} \end{bmatrix} + \begin{bmatrix} 0 \\ 1 \end{bmatrix} \mu, & x \geq 0 \end{cases}$$

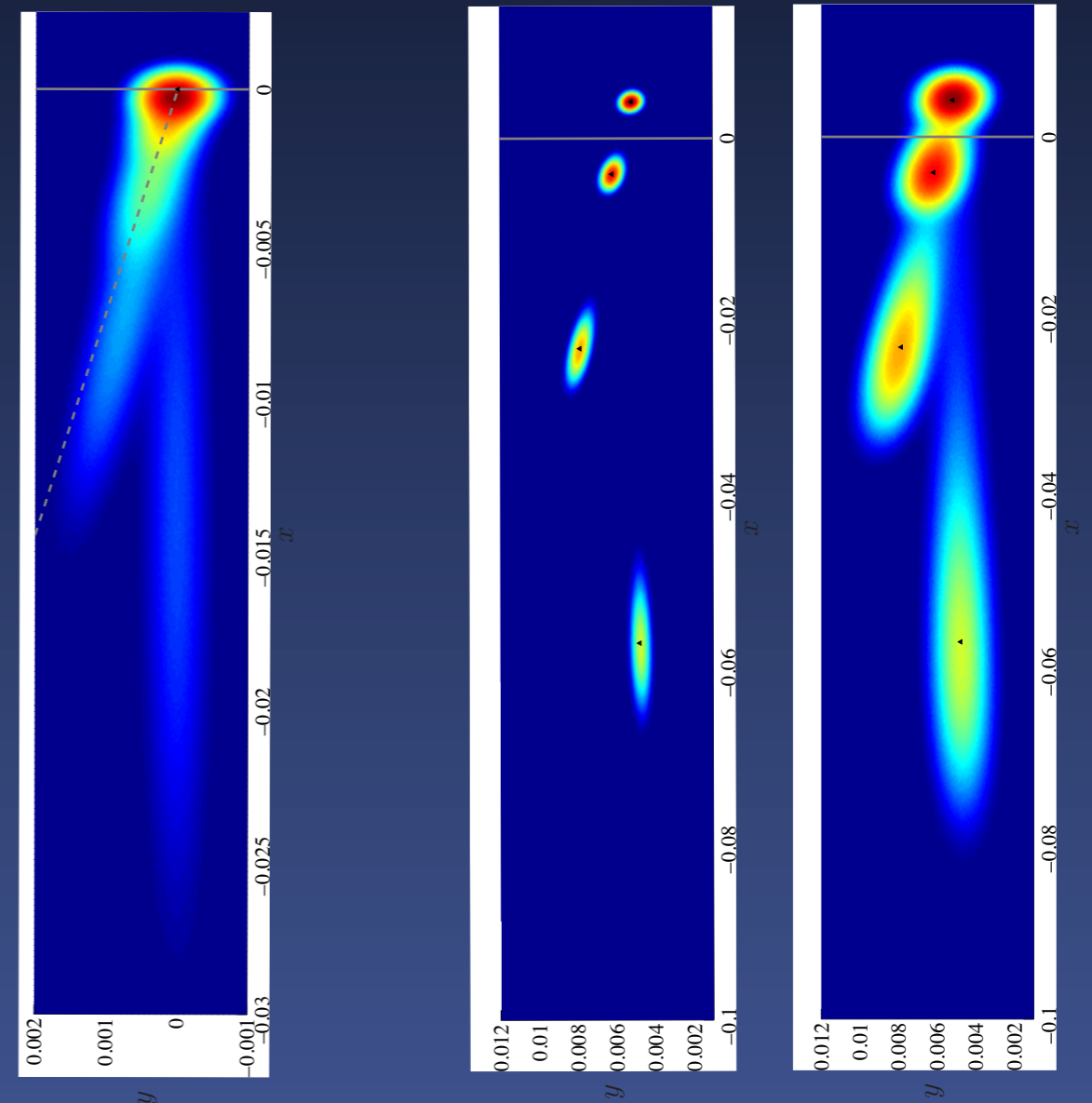
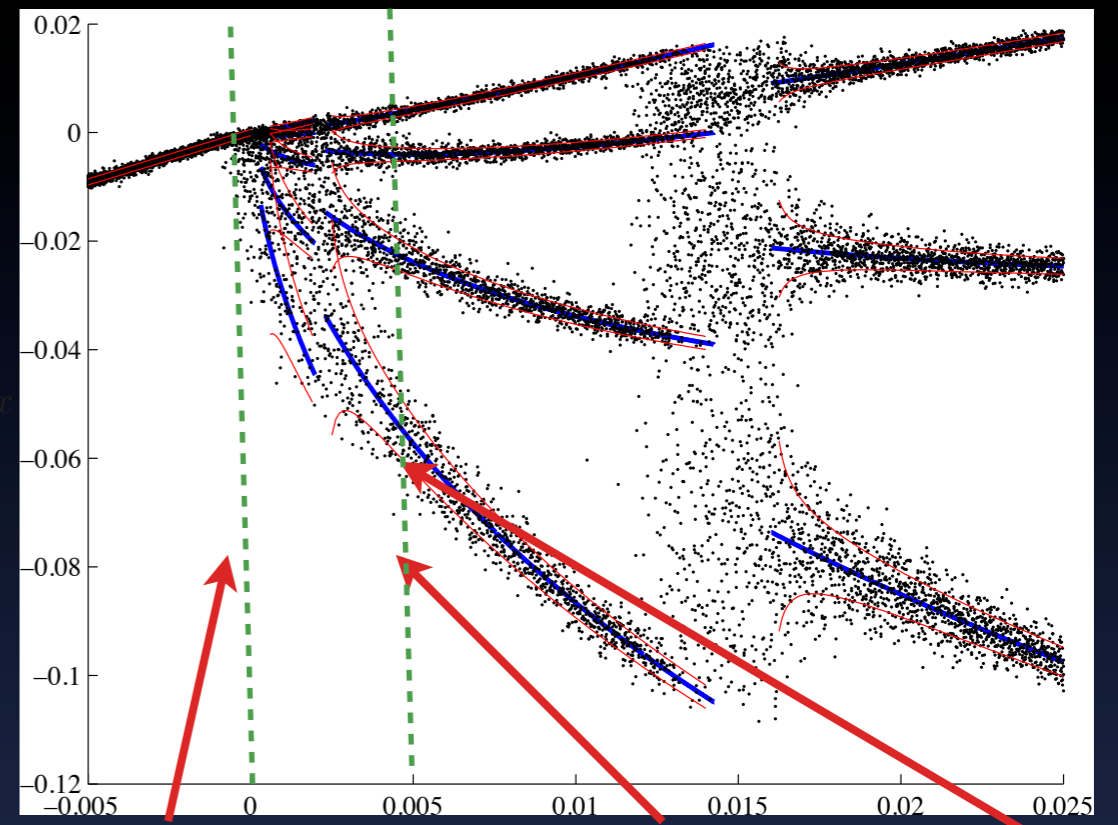
Noise sensitivity: grazing

Stochastic Poincare map
derived from cts model
(vs. Poincare map + noise)

Gaussian densities:
well separated branches

Non- Gaussian:
branches overlap, square root
“stretching” follows iterates
near switching

Simpson, Hogan, K. SIADS, to appear



Different types of stochastic discontinuous dynamics: need a variety of ideas

- Mixed mode oscillations: Semi-analytical iterations of time dependent probability density functions
- Discontinuity induced bifurcations - underlying sources of noise-sensitivity
- Positive occupation times: sliding
- Boundary layers and non-standard scaling limits: sliding transitions
- Grazing: Stochastic Poincare maps

Lots of mathematical and modeling challenges:

- Nonlinear models with delay: complex behaviors
- Piecewise continuous nonlinear systems: recently receiving more attention
- Stochastic modeling for systems with delay and discontinuities: open problems analytically and computationally, new approaches needed
- Robustness of different on/off control strategies
- Recent work: extension of results for continuous cases to discontinuous drift cases with “nice” noise (Mohammed, et al, 2013 in progress)