

RAPTOR: a lightweight transport model for open-loop optimization and real-time simulation

Federico Felici

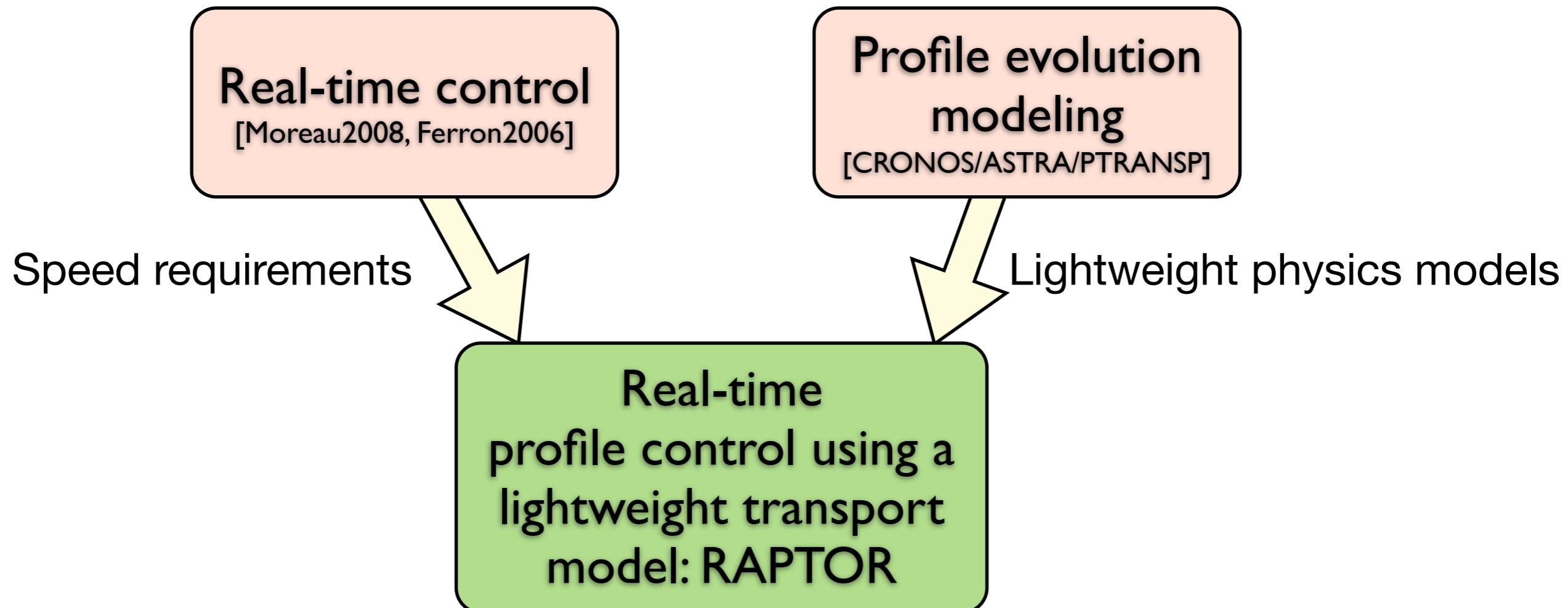
O. Sauter, S. Coda, B.P. Duval, T.P. Goodman J-M. Moret and the TCV team

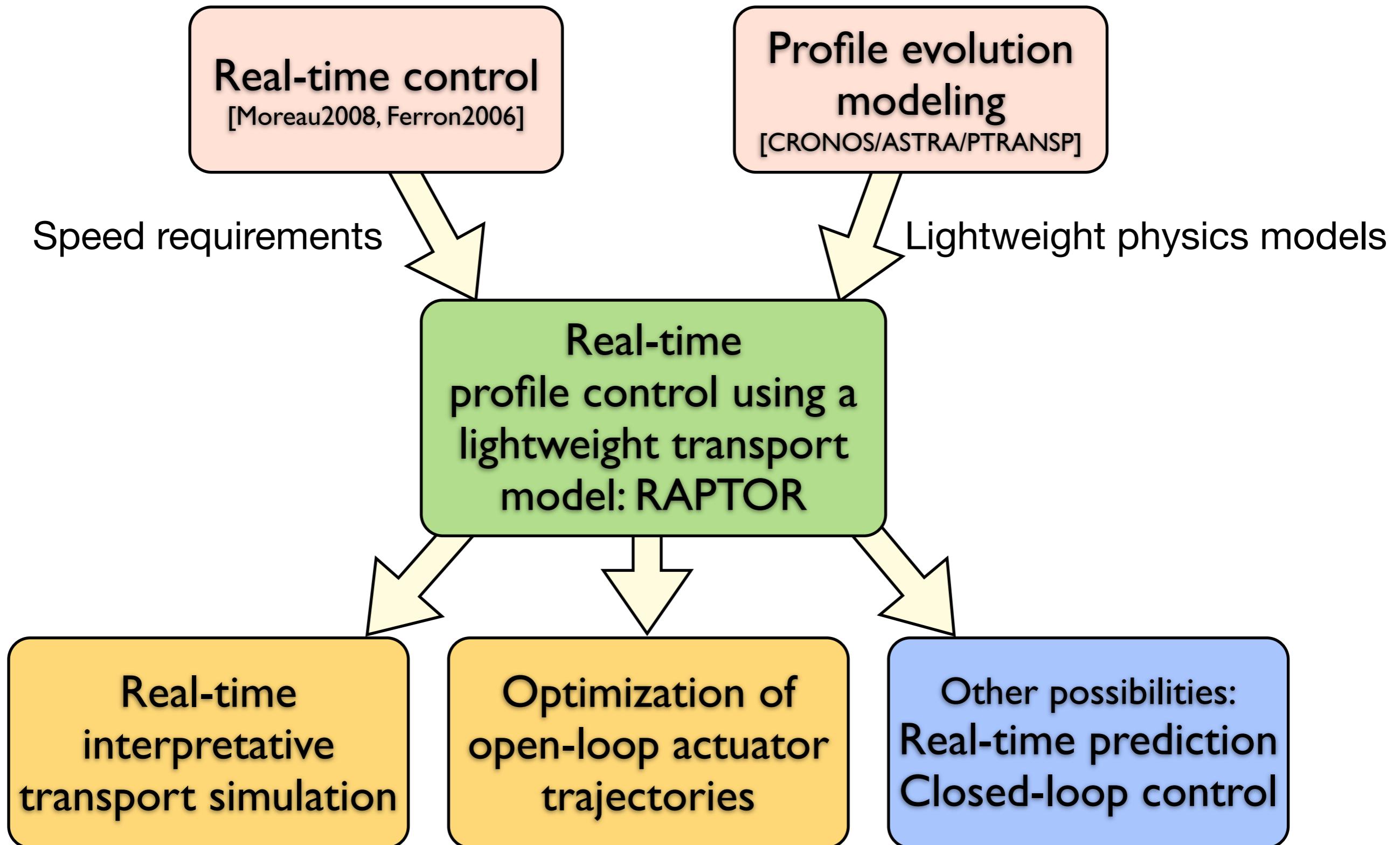
CRPP-EPFL, Association Euratom-Suisse, CH-1015 Lausanne Switzerland

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Real-time control
[Moreau2008, Ferron2006]

**Profile evolution
modeling**
[CRONOS/ASTRA/PTRANSP]





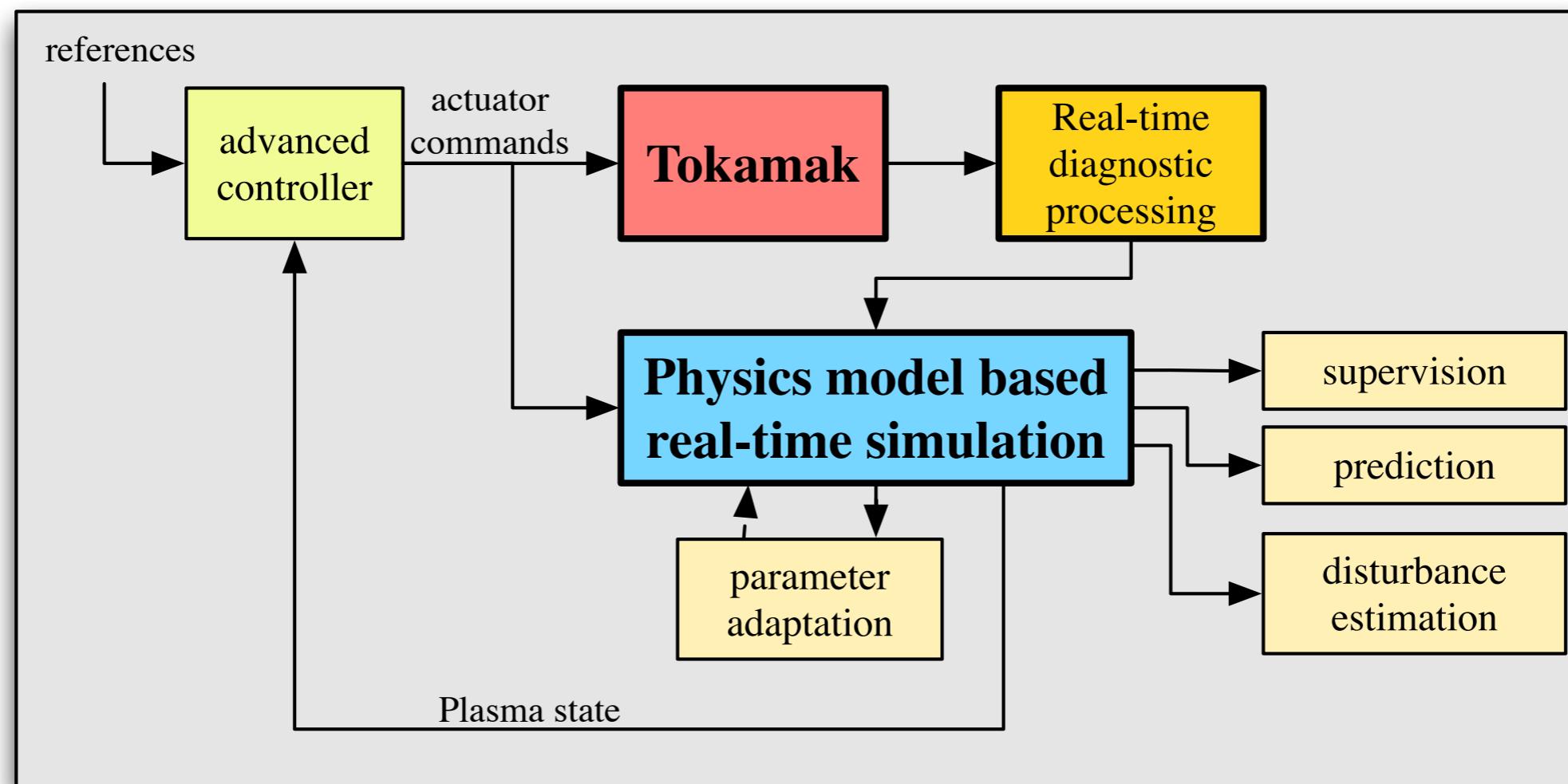
Need lightweight transport model Compatible with real-time execution

- **RAPTOR - RApid Plasma Transport simulOR**
- Fast 1-D transport code for real-time implementation and fast optimization
 - Evolves profiles of poloidal flux $\Psi(\rho,t)$, and electron temperature $T_e(\rho,t)$
 - Fixed flux surface shapes from pre-calculated MHD equilibrium
 - Neoclassical resistivity, bootstrap current [Sauter PoP 1999,2002]
 - q , shear profile dependent ad-hoc transport model χ_e (similar to [Polevoi2002,Garcia2010])
 - Parametrized heating / current drive sources
- Includes nonlinear profile coupling, crucial for hybrid/advanced scenarios
- Similar to [Witrant, PPCF 2007]
 - But additionally solves full T_e profile dynamics
 - Different numerics (Finite Elements, implicit solver)

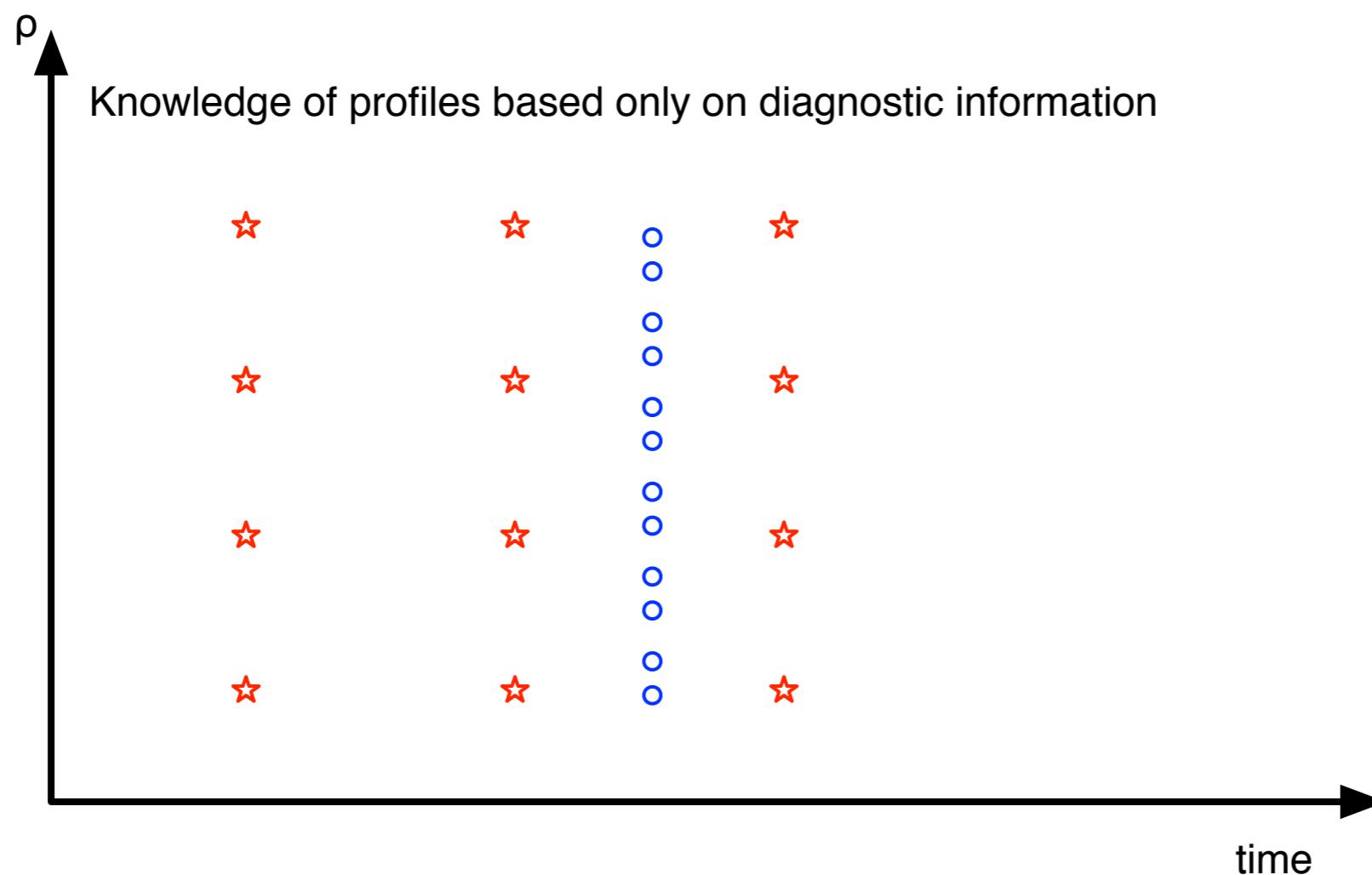
New paradigm: real-time simulation for physics-based plasma state reconstruction

Real-time simulation

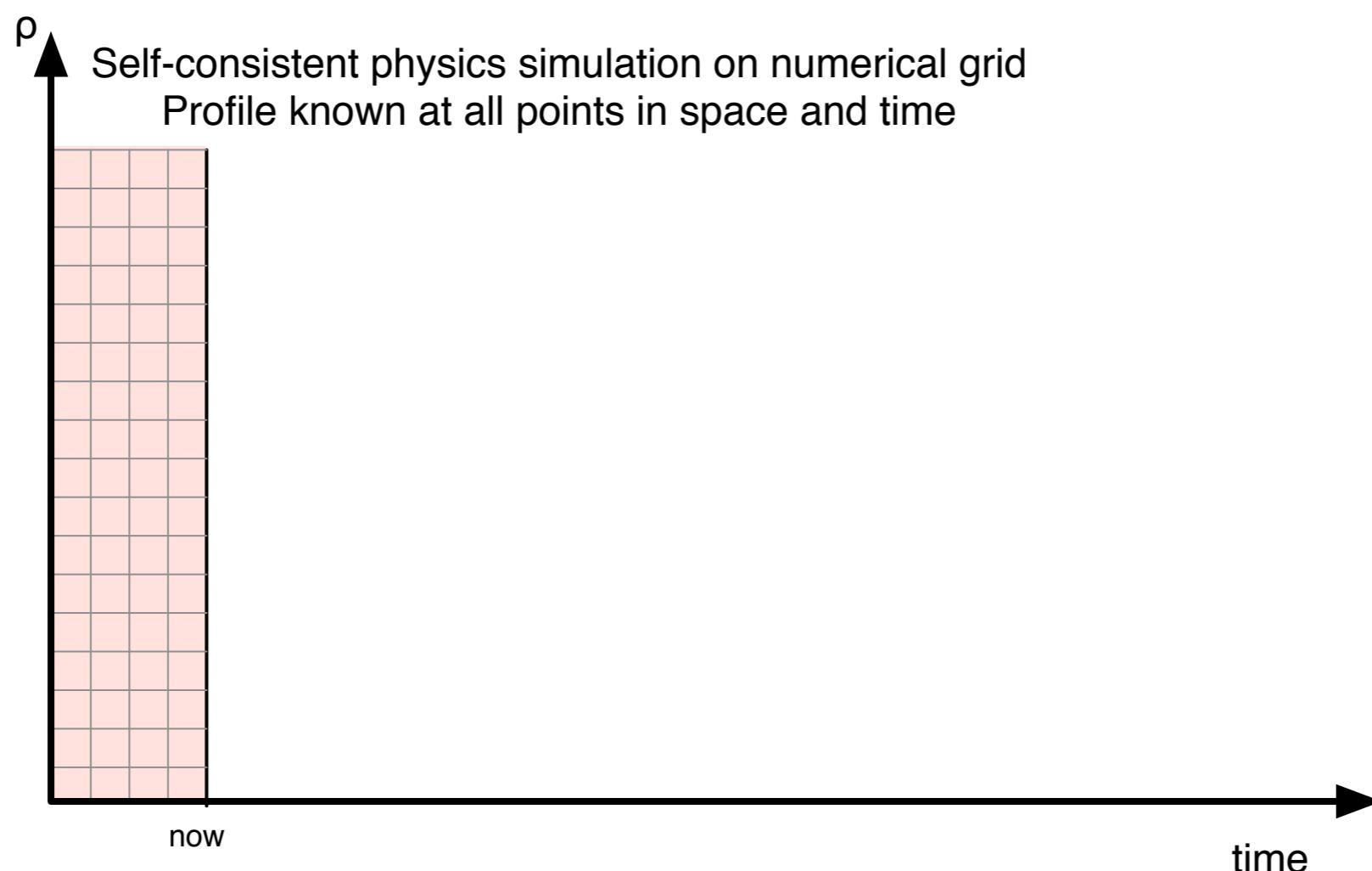
- Evolve plasma numerically, while it is physically evolving in the tokamak
- Use *available* diagnostics as constraints
- In control engineering terms: a *Nonlinear, dynamic model-based state observer*
- Model-reality mismatch: disturbance estimation or parameter adaptation



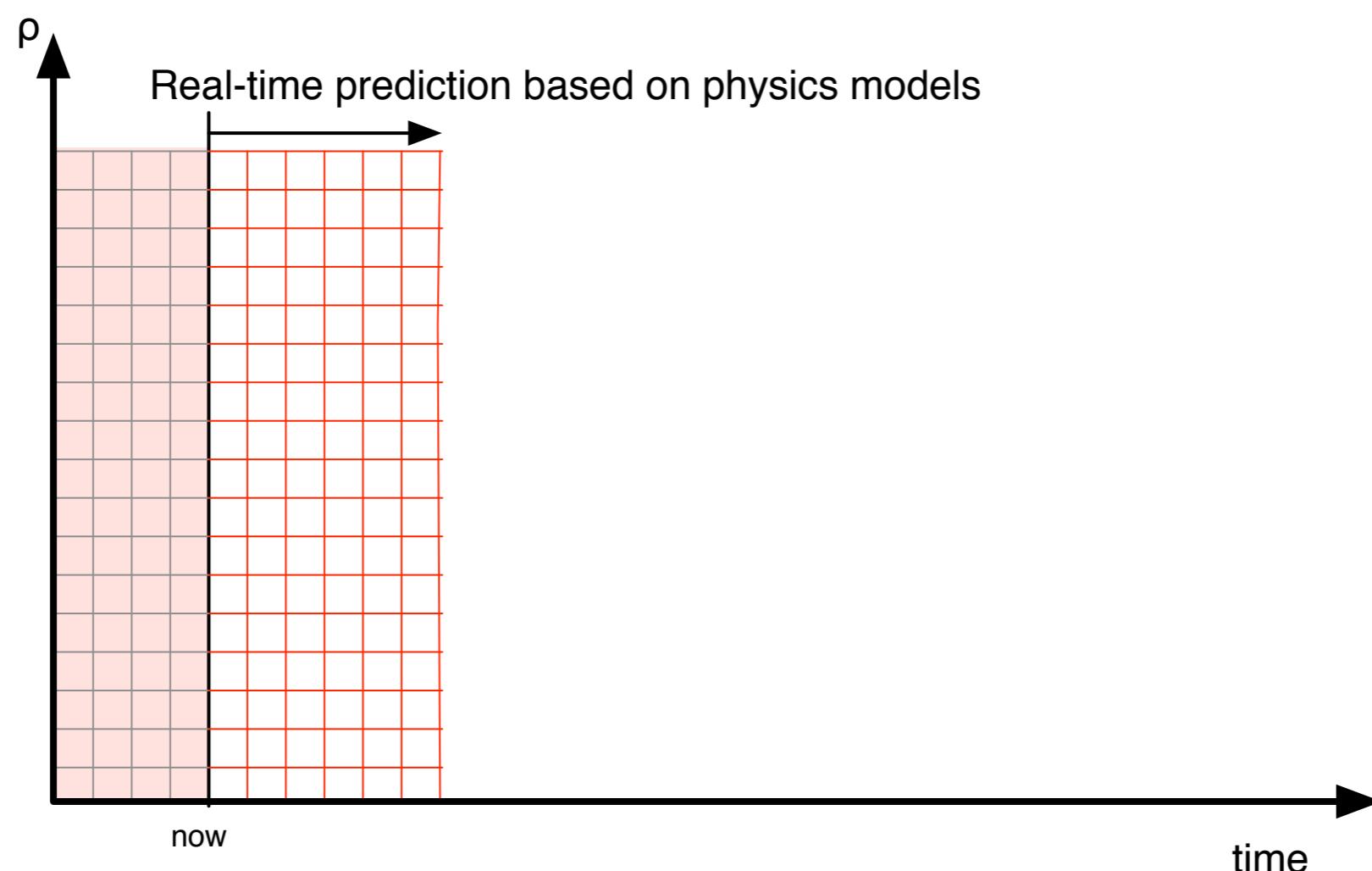
- Continuous function on (ρ, t)
- Different diagnostics give information at different spatial and temporal points
 - Today: profile information (e.g. for control) based exclusively on these points



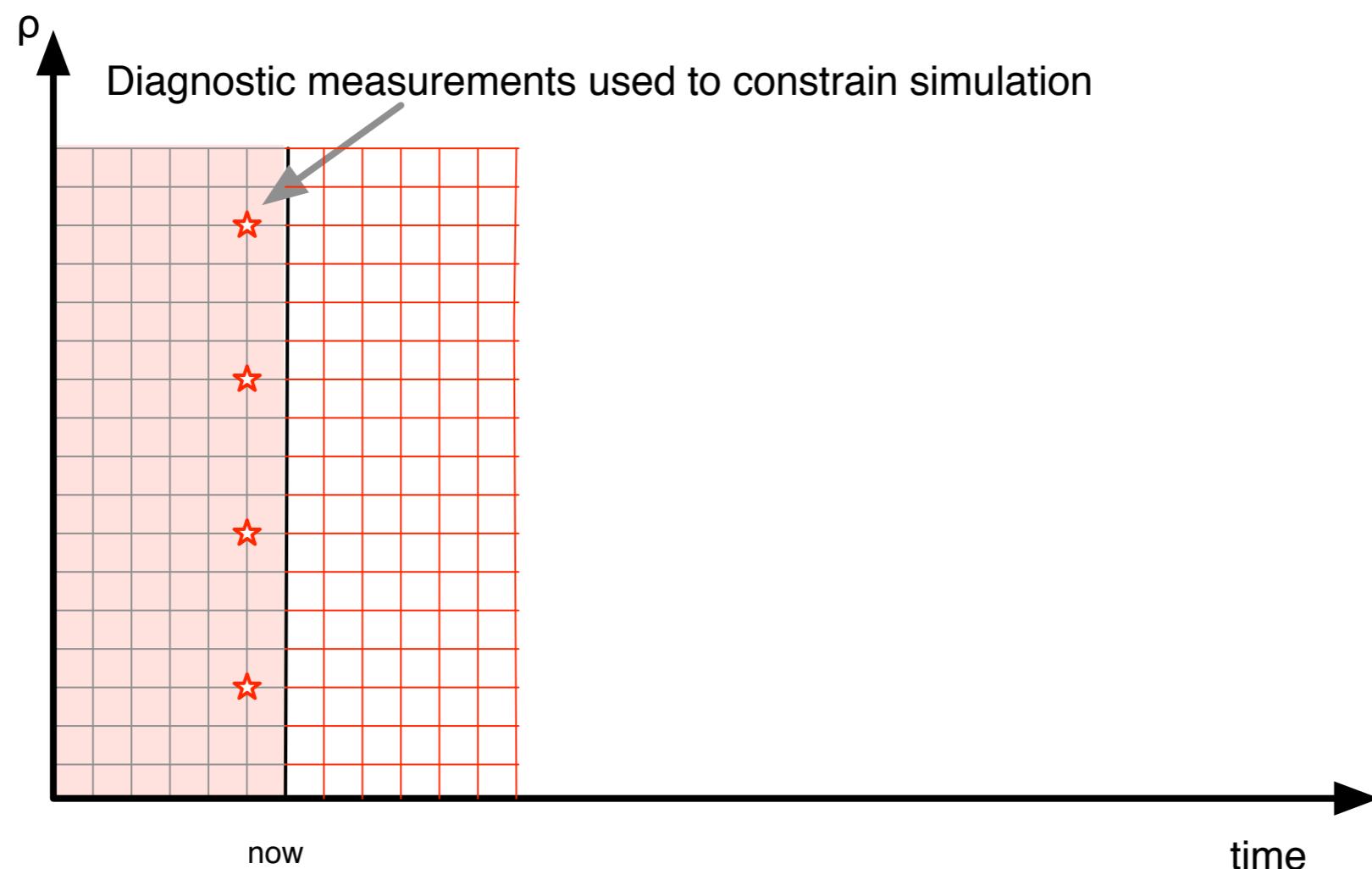
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 - Grid size determined by underlying physics, available CPU power



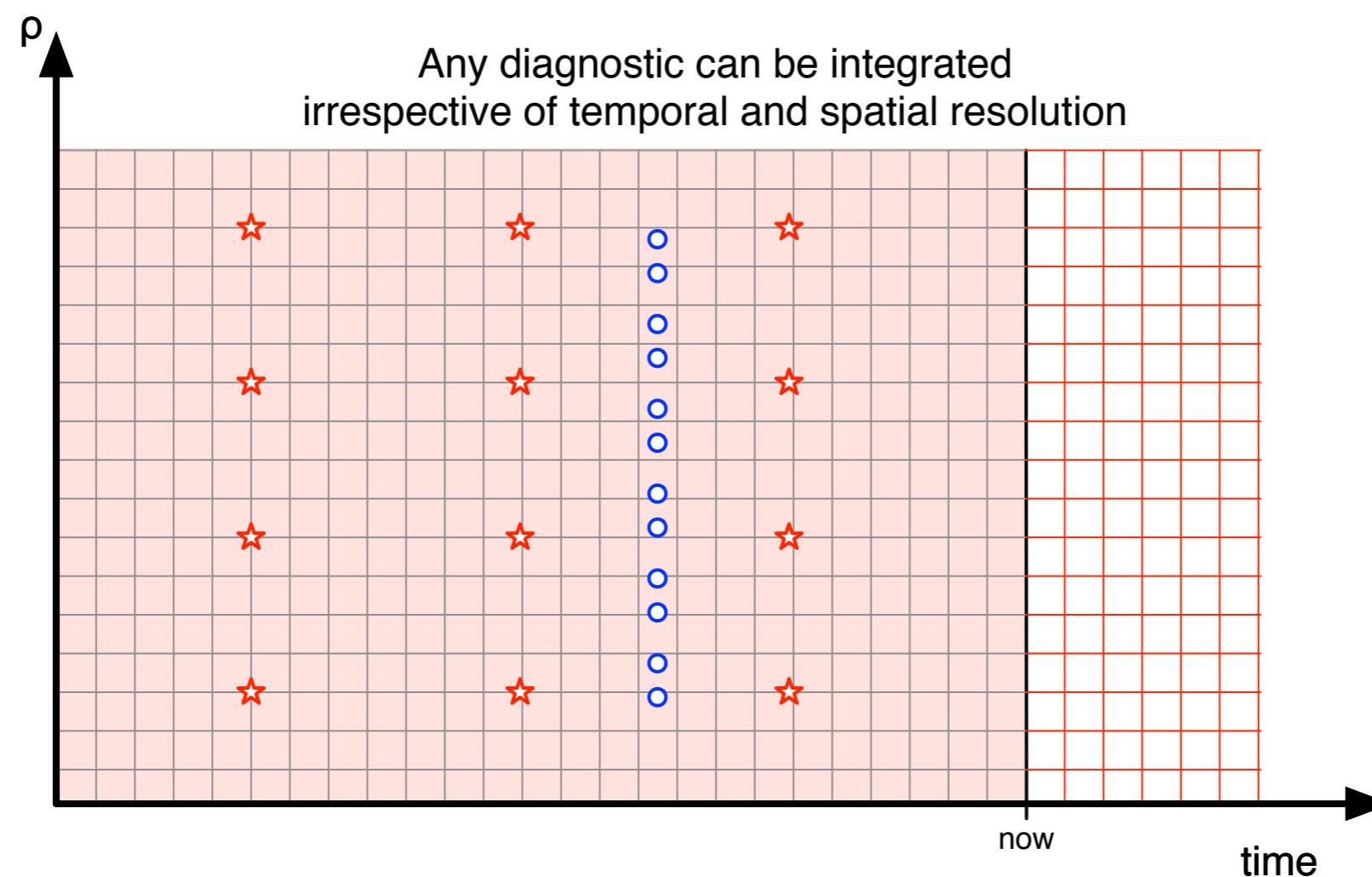
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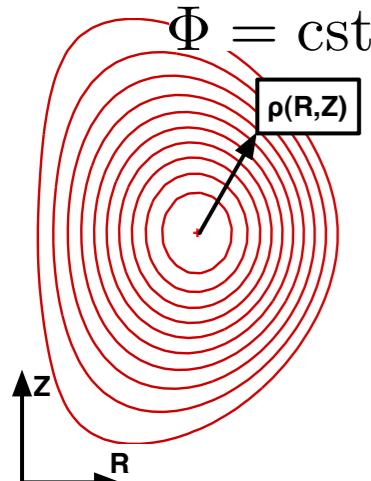


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Simplified physics model for real-time poloidal flux transport simulations

- Poloidal flux diffusion equation (1-D)
 - Solved assuming fixed flux surface distribution on (R,Z)



$$\sigma_{||} \frac{\partial \psi}{\partial t} = \frac{R_0 J^2}{\mu_0 \rho} \frac{\partial}{\partial \rho} \left(\frac{G_2}{J} \frac{\partial \psi}{\partial \rho} \right) - \frac{V'}{2\pi \rho} (j_{BS} + j_{CD})$$

$$\rho = \sqrt{\frac{\Phi}{\pi B_0}}, \quad J = \frac{RB_\phi}{R_0 B_0}, \quad V' = \frac{\partial V}{\partial \rho}, \quad G_2 = \frac{V'}{4\pi^2} \left\langle \frac{(\nabla \rho)^2}{R^2} \right\rangle$$

[Hinton&Hazeltine Rev. Mod. Phys 1976], [Pereverzev IPP rep 1991]

- Sources

$$j_{BS} = -\frac{2\pi J(\psi)}{B_0 R_{pe}} \frac{\partial \rho}{\partial \psi} \left[\mathcal{L}_{31} \frac{\partial n_e}{\partial \rho} T_e + (\mathcal{L}_{31} + R_{pe} \mathcal{L}_{32} + (1 - R_{pe}) \alpha \mathcal{L}_{34}) \frac{\partial T_e}{\partial \rho} n_e \right]$$

Bootstrap current:
Neoclassical physics
[Sauter PoP 1999]

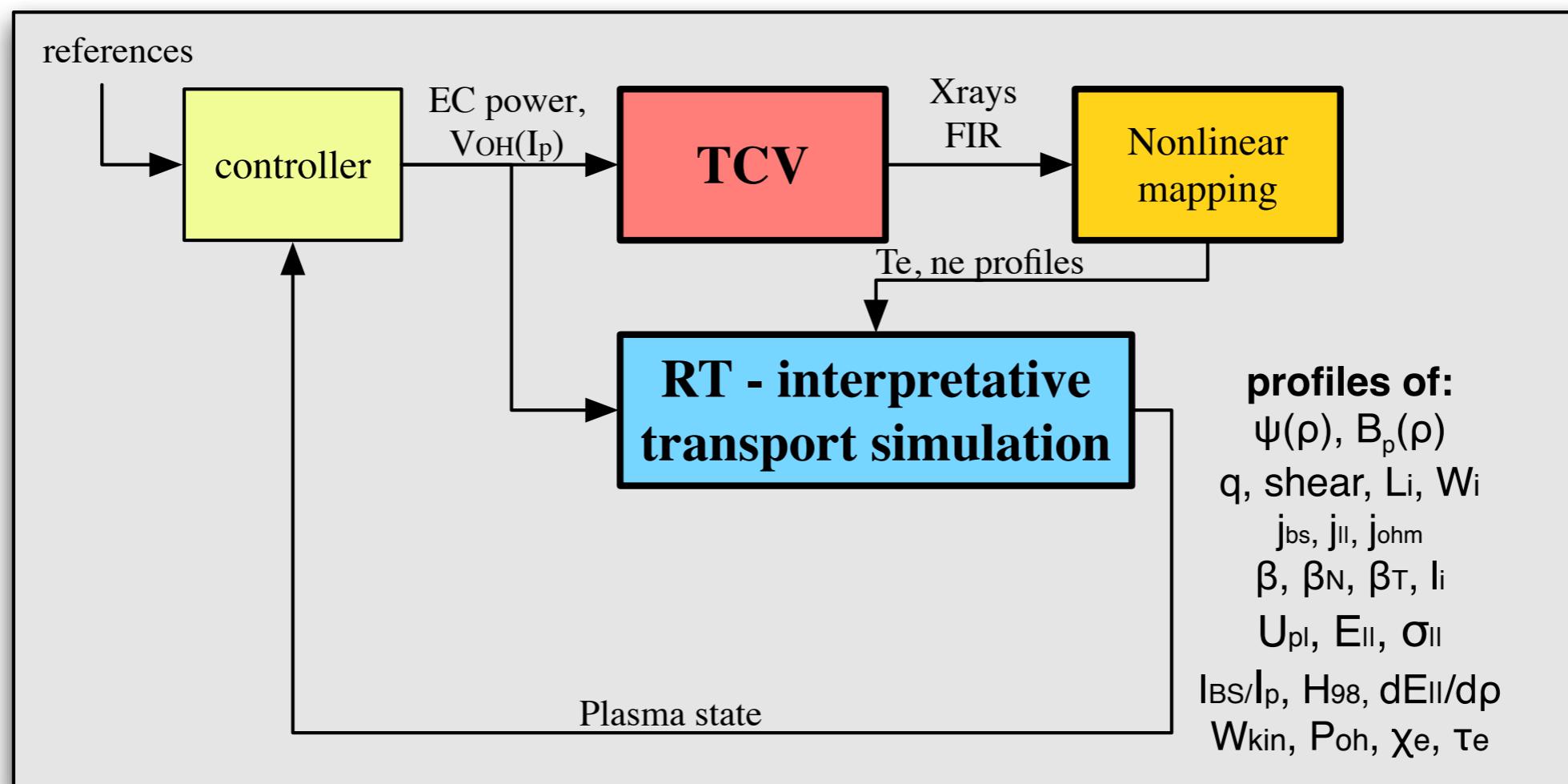
$$j_{ECCD}(\rho, t) = \underbrace{c_{exp} e^{-\rho^2/0.5^2}}_{\eta_{EC}} \frac{T_e}{n_e} \exp \left\{ \frac{(\rho - \rho_{dep})^2}{w_{cd}^2} \right\} P_{gyro}(t)$$

Gaussian shape for
EC current deposition

- Need inputs: I_p , $T_e(\rho)$, $n_e(\rho)$ at each time step: **from RT diagnostics**

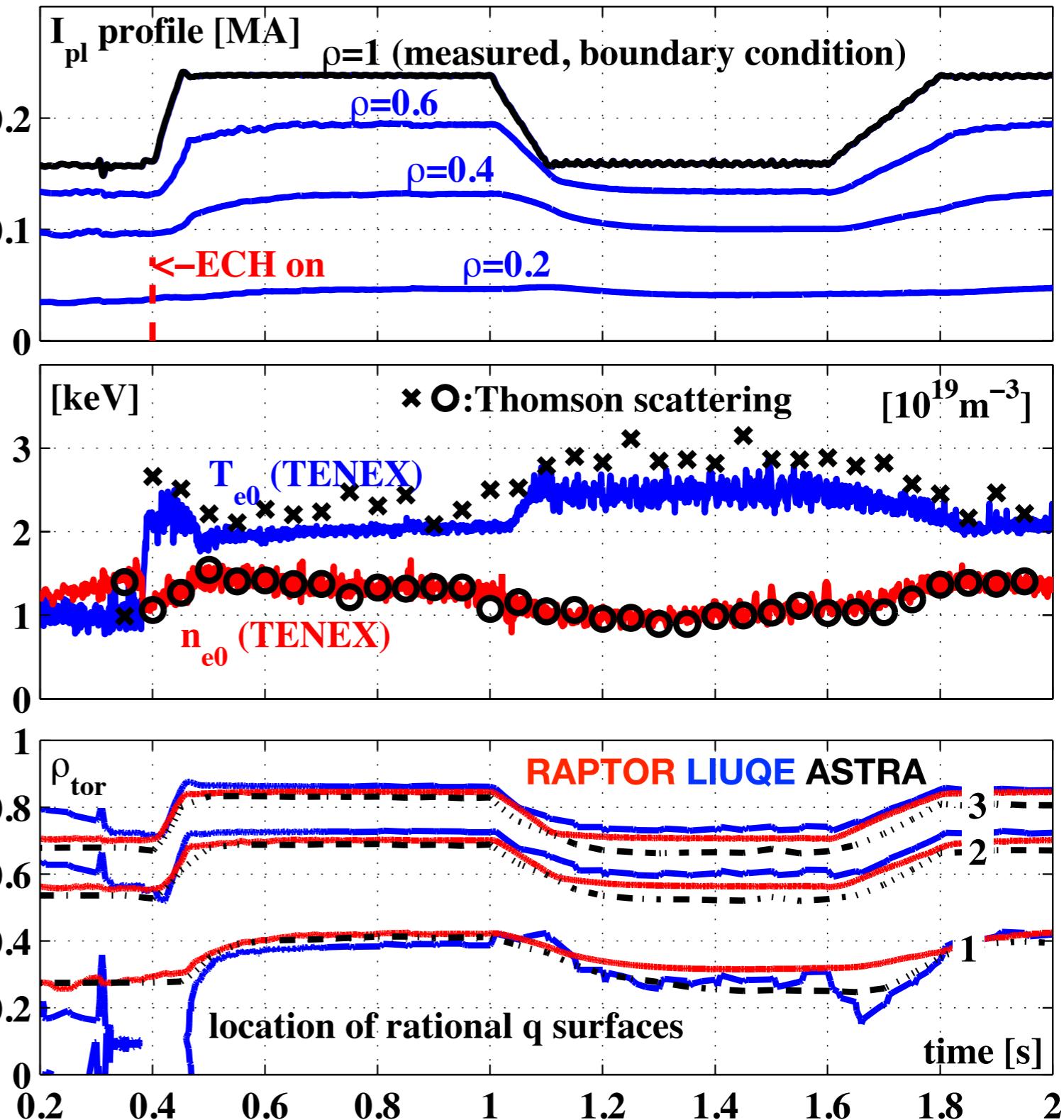
Pilot implementation on TCV demonstrates that real-time simulation is certainly feasible on larger tokamaks

- Current density profile: hard to measure, physics well understood
 - Solve flux diffusion equation with kinetic profiles from real-time diagnostics
 - Flux profile simulated on TCV every 1ms (<150ms current redistribution time)
 - Results comparable to off-line interpretative modeling (ASTRA)



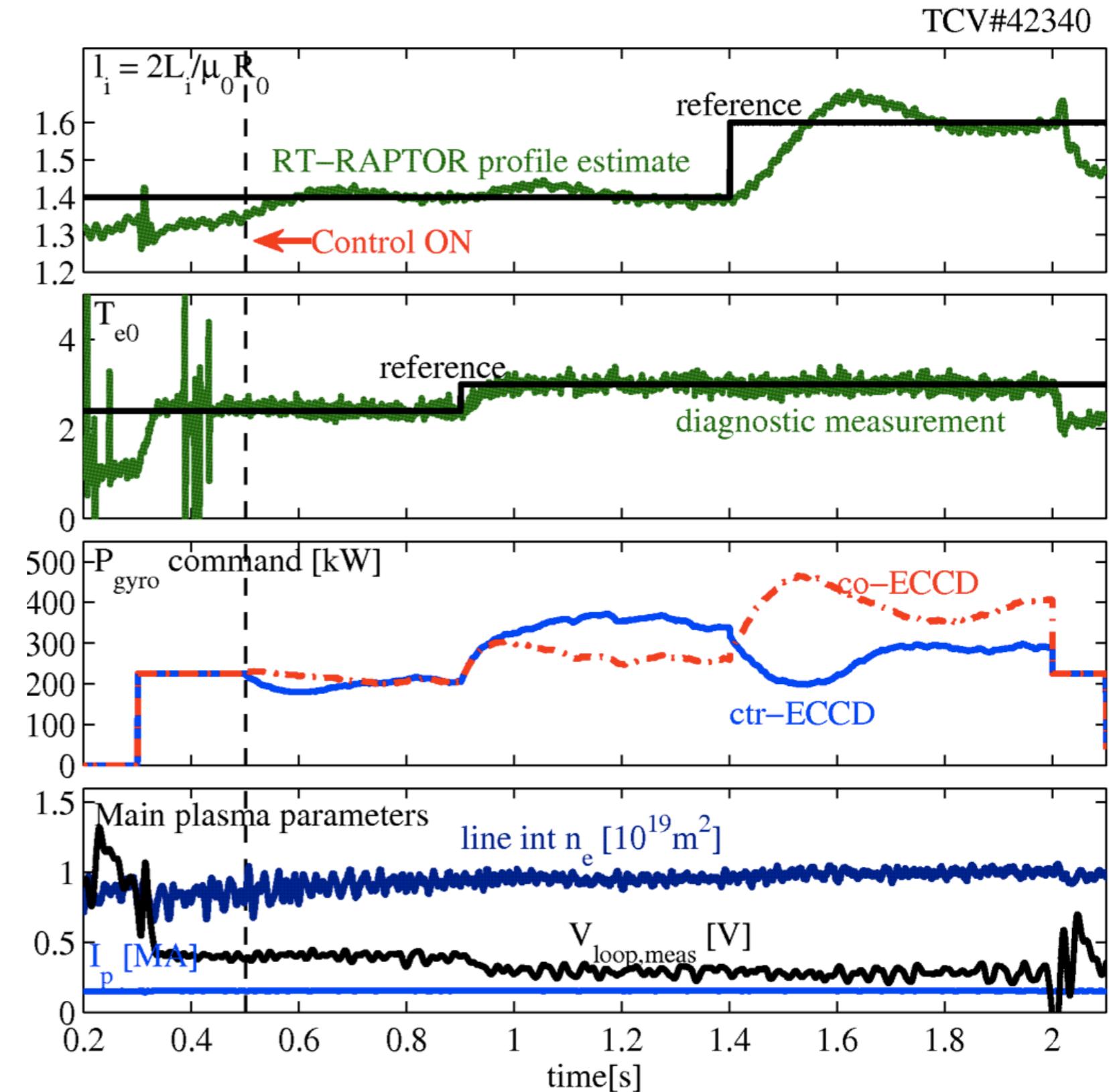
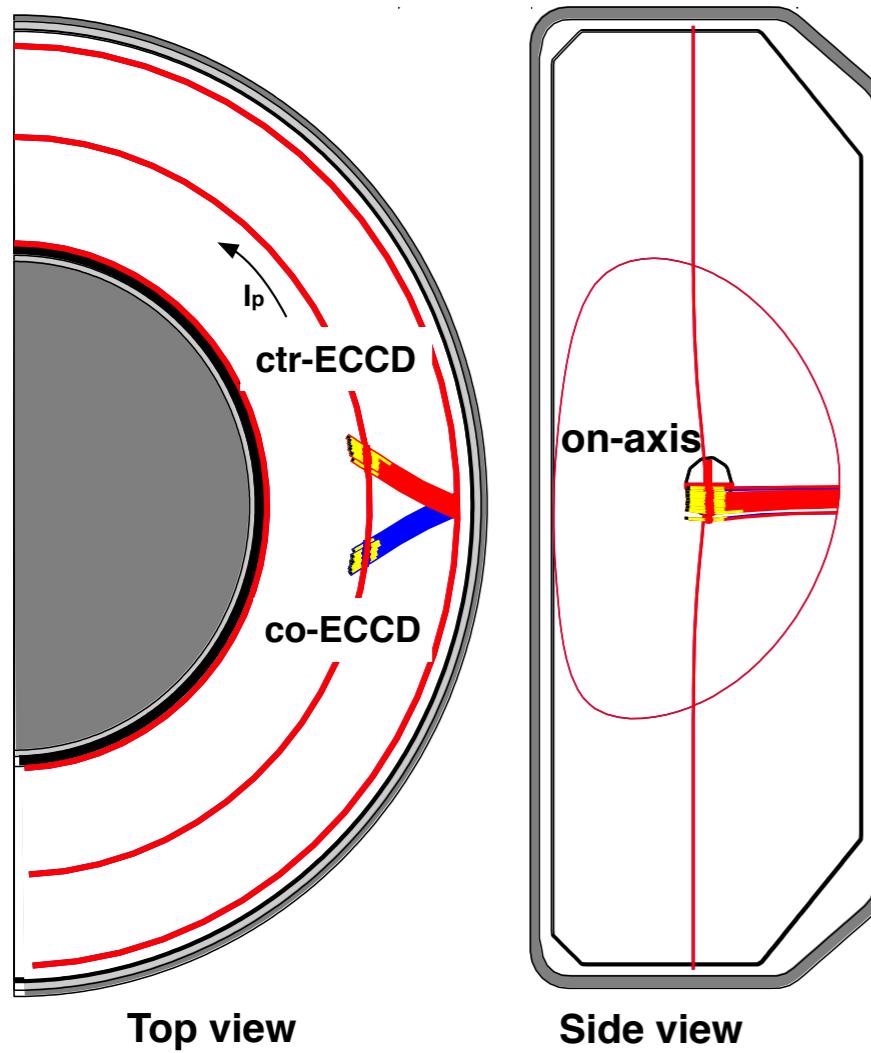
Experiments confirm that RT-RAPTOR gives results similar to off-line estimates

- Experiment involving I_p ramps
 - Compare RAPTOR with
 - LIUQE (offline GS solver)
 - ASTRA (offline interpretative transport modeling)



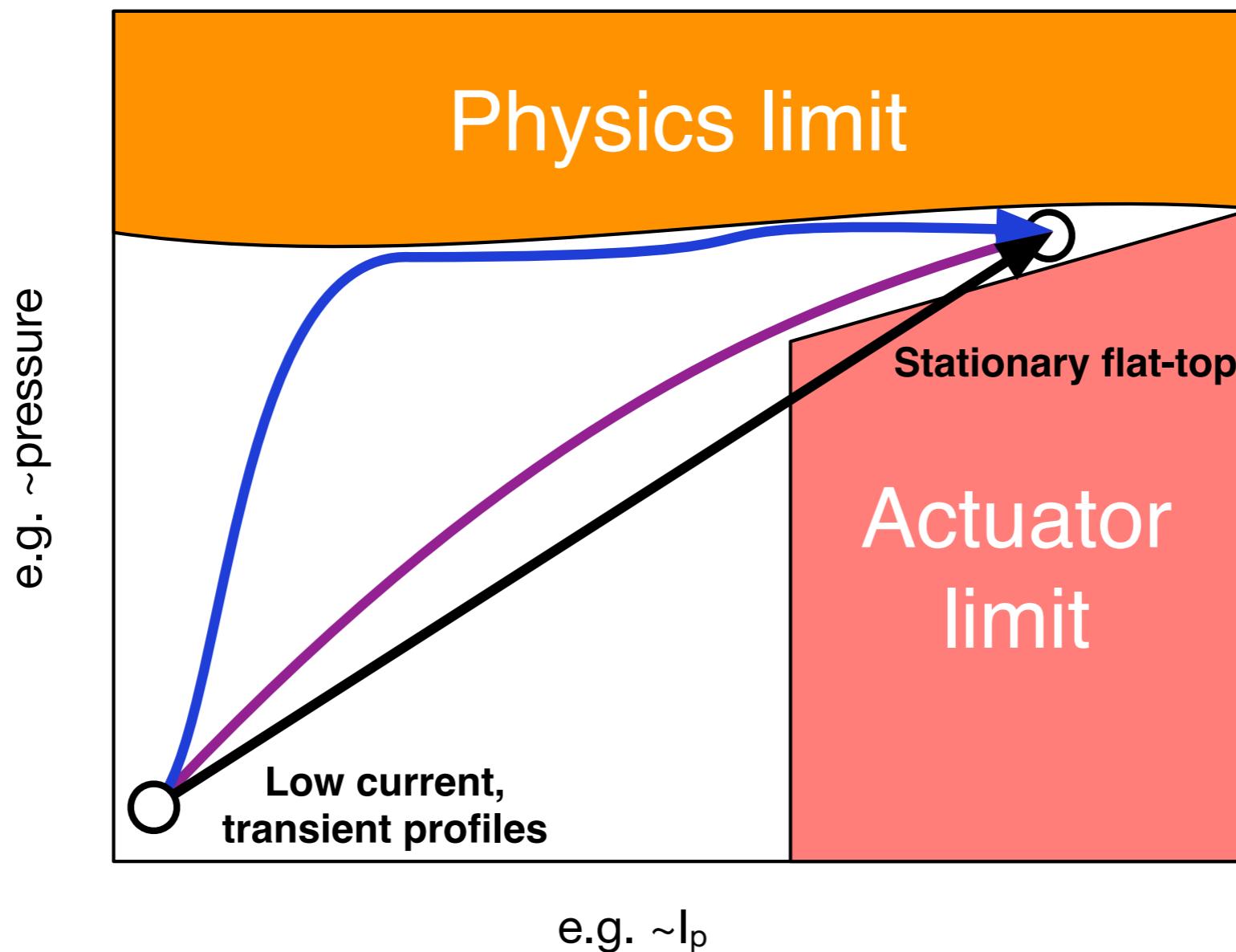
Closing the loop: Simultaneous feedback control of I_i and T_{e0}

- 1 co-ECCD gyrotron
- 1 counter ECCD gyrotron
 - Sum of powers $\rightarrow T_e$
 - Difference of powers $\rightarrow L_i$



Optimization of open-loop trajectories

Tokamak operational space
Which route to take?



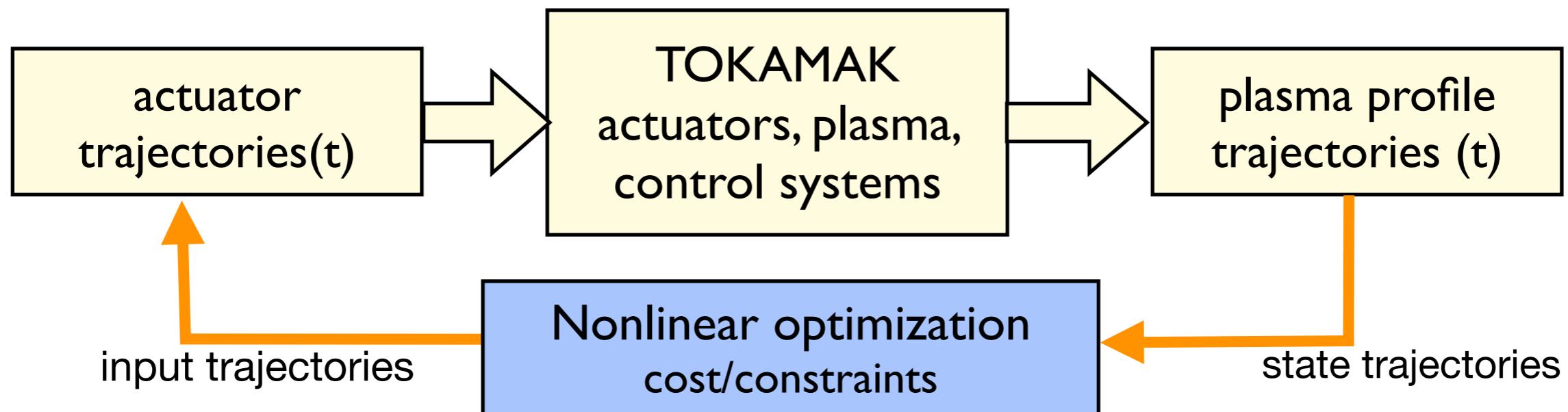
Determine open-loop actuator trajectories: a nonlinear optimization problem

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- Given the initial plasma profiles, what input trajectories should I use to:
 - Minimize a **cost function** depending on the *final* plasma state
 - While satisfying **constraints on the state**
 - And satisfying **constraints on the actuators**

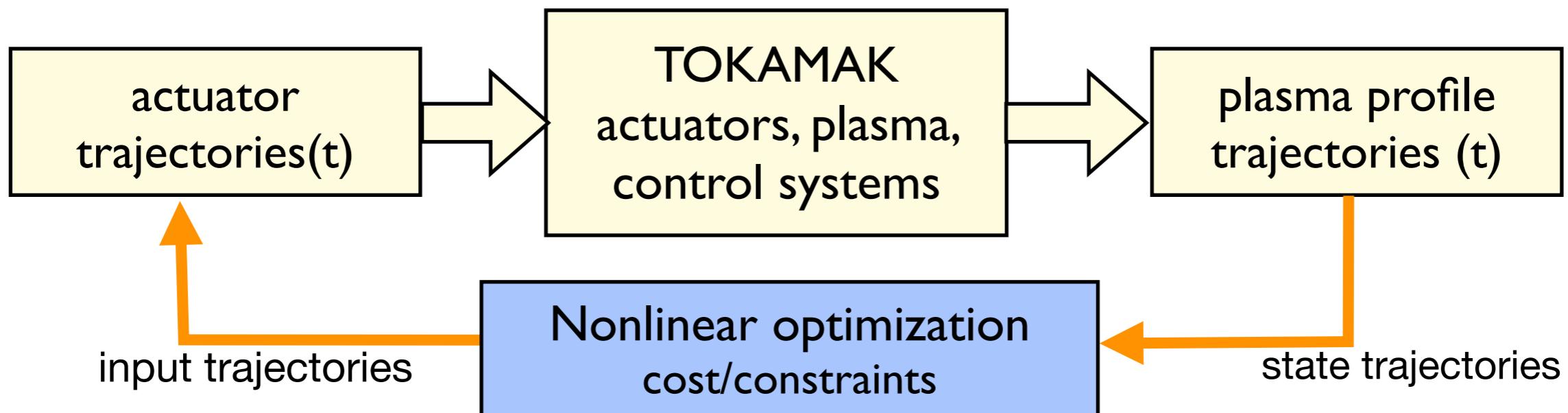
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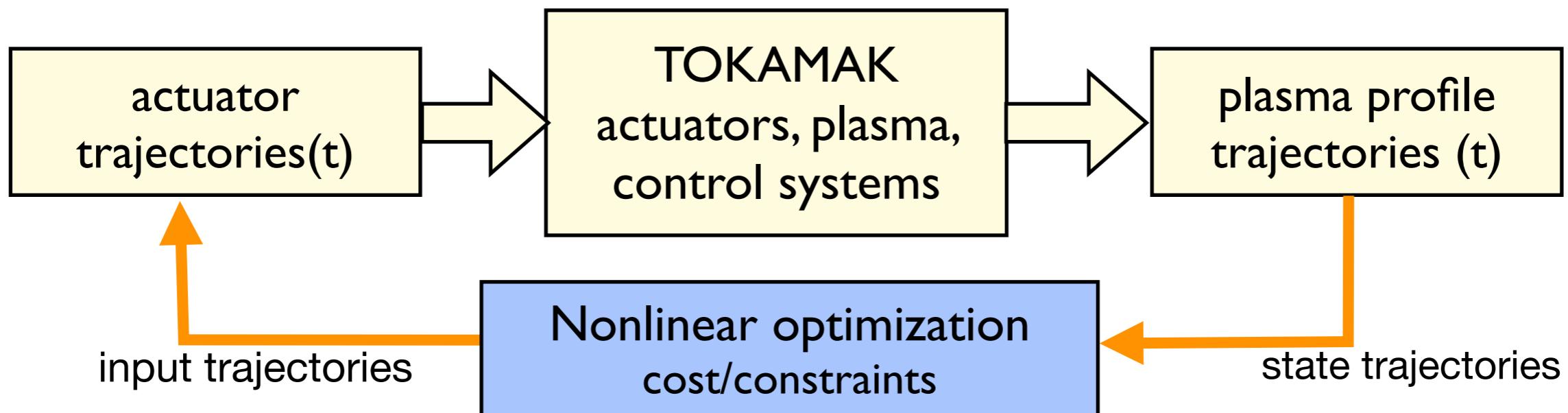
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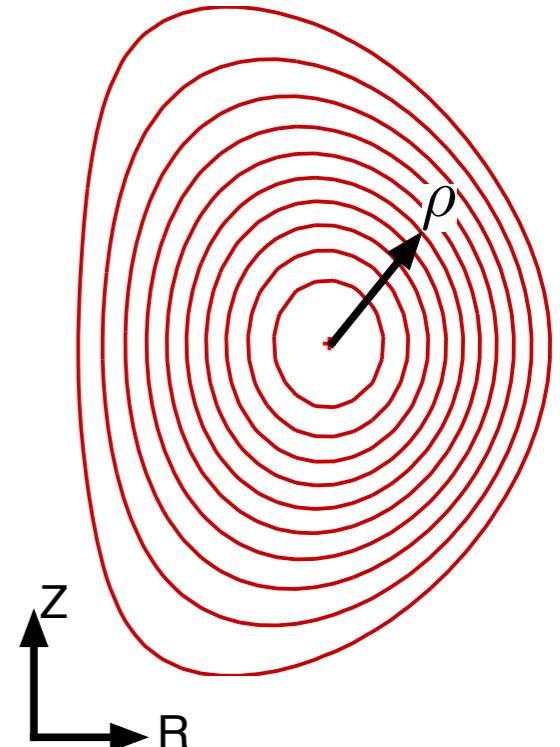
- Need to do many simulations
- Predictive-RAPTOR transport code
 - Returns all gradients of state trajectories w.r.t input trajectory parameters
 - Very fast (one time step: ~10ms, one simulation: ~1 second, full optimization ~1 minute)

Predictive-RAPTOR equations

- Noncircular, axisymmetric, fixed toroidal flux surface shape
- 1D, (flux surface averaged) diffusion of poloidal flux

$$\sigma_{||} \frac{\partial \psi}{\partial t} = \frac{R_0 J^2}{\mu_0 \rho} \frac{\partial}{\partial \rho} \left(\frac{G_2}{J} \frac{\partial \psi}{\partial \rho} \right) - \frac{V'}{2\pi\rho} (j_{BS} + j_{ext})$$

- Neoclassical conductivity $\sim T_e^{3/2}$
- Bootstrap current $\sim \nabla T_e$
- Current drive sources as sums of gaussians
- Boundary condition through total I_p



- Flux surface averaged electron temperature diffusion

$$V' \frac{\partial}{\partial t} [n_e T_e] = \frac{\partial}{\partial \rho} n_e \chi_e \frac{\partial T_e}{\partial \rho} + V' P_e$$

- Fixed ion and density profile
- Heat sources as sums of gaussians
- Ad-hoc model for thermal diffusivity

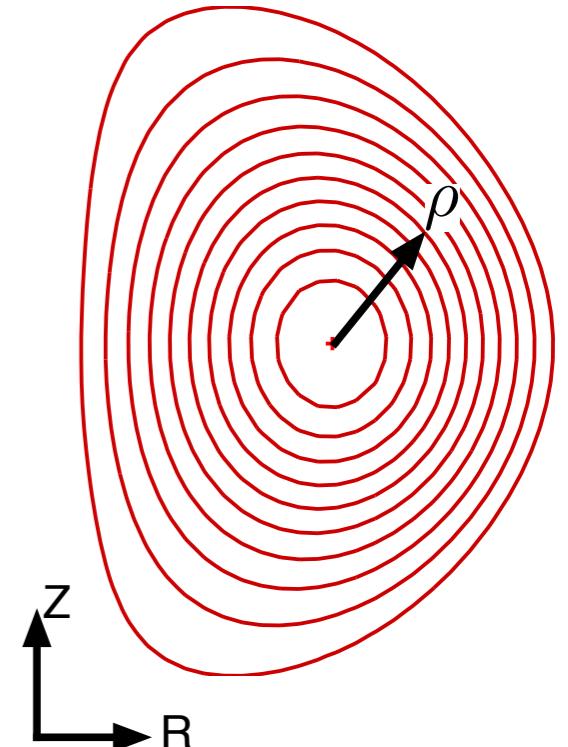
$$\chi_e = \chi_{neo} + c_{ano} \rho q F(s) + \chi_{central} e^{-\rho^2/0.1^2}$$

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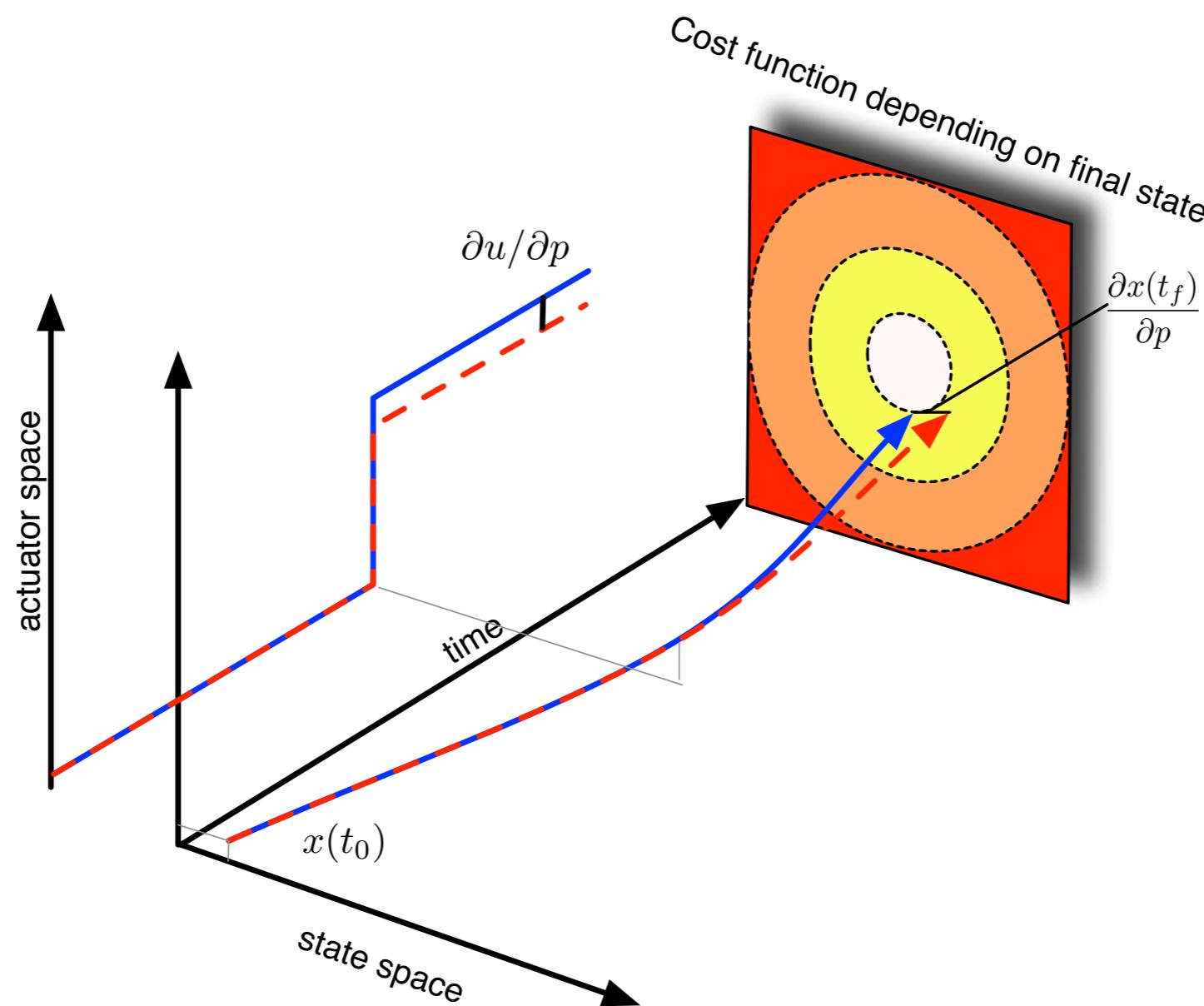
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Compared to other transport models (e.g. CRONOS/ASTRA):

- No self-consistent equilibrium, fixed bnd
- No consistent ray tracing/NBI modules
- No ion or density simulation
- No complex transport models (eg GLF23)

RAPTOR returns gradients of profile evolution w.r.t. input trajectory parameters

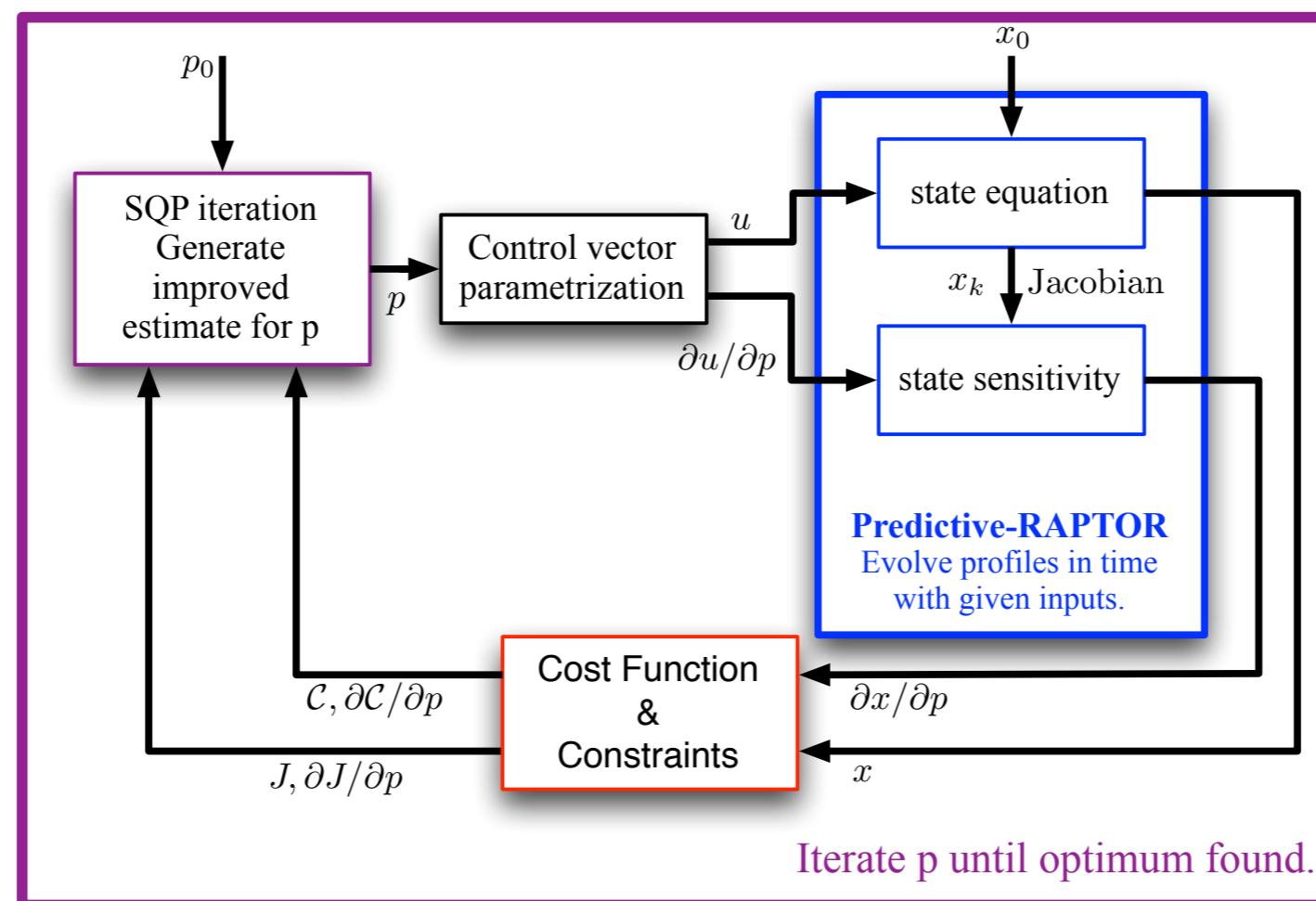
- Gradients computed using forwards sensitivity method
 - State sensitivities: dx/dp at all times.
 - Used to quickly evaluate cost function gradient $dJ/dp = dJ/dx_f \ dx_f/dp$
 - Linearization of the profile dynamics around the profile trajectory
 - Useful for control



Solution of optimal control problem by SQP nonlinear programming algorithm

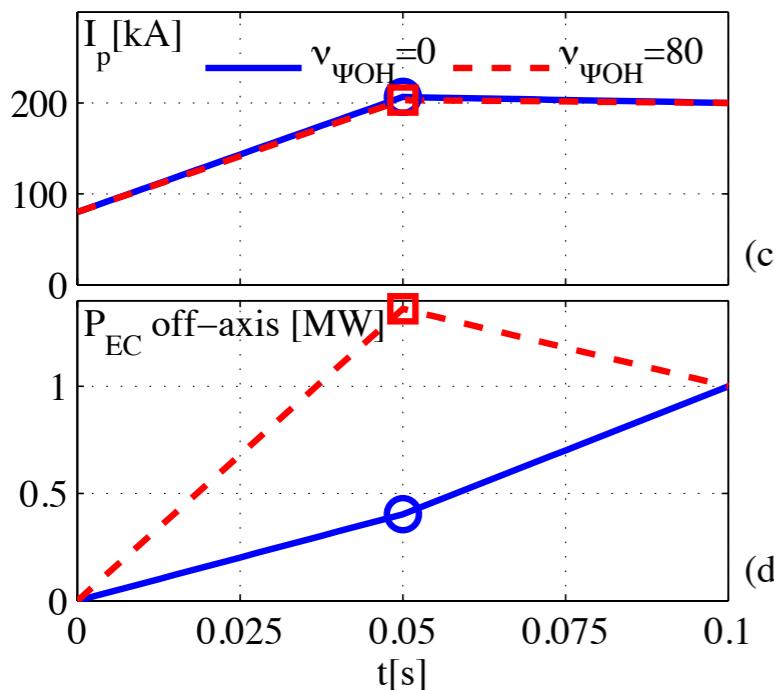
○ Solution: Sequential Quadratic Programming (SQP)

- Iteratively solve local approximation to nonlinear optimization problem:
 - Quadratic cost function + linear constraints
 - Gradients dJ/dp and dC/dp , are computed from **state sensitivities**
 - Quasi-newton method for Hessian
 - **Avoid finite-difference evaluation of gradients (expensive!)**
 - Use version implemented in Matlab, called via fmincon



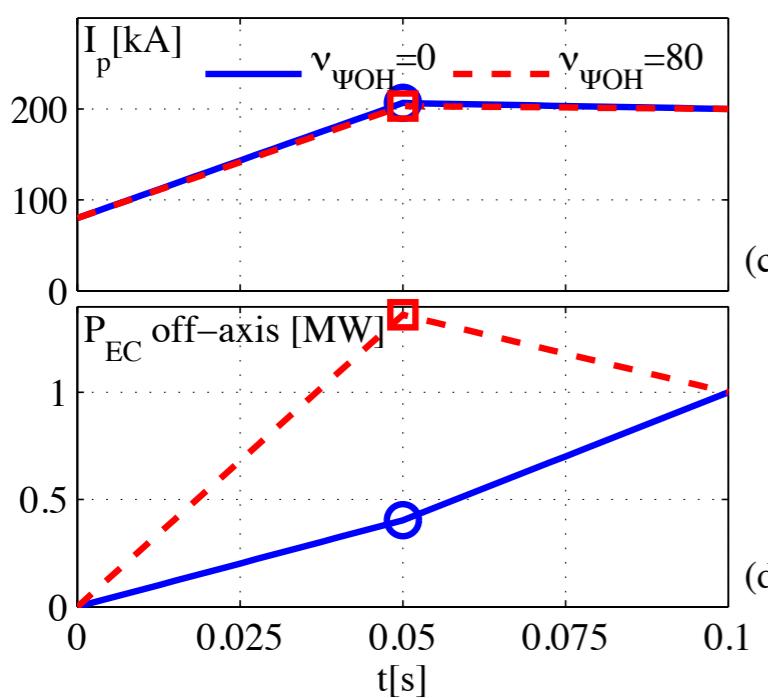
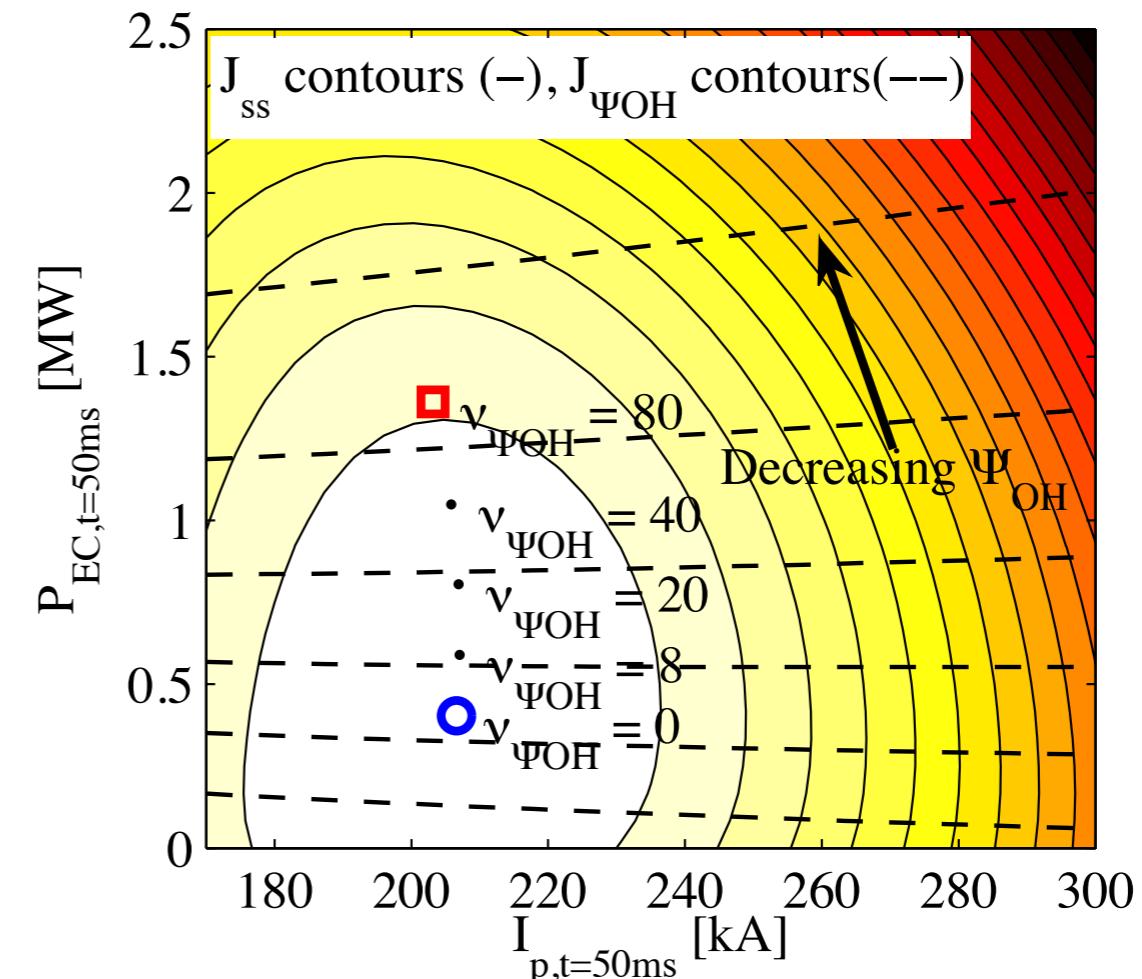
A first example: 2 parameters

- Ramp up from 80kA to 200kA in 100ms
 - Off axis ECCD at $\rho=0.3$
- Cost function terms:
 - J_{ss} : Stationary profiles at final time (flat V_{loop})
 - $J_{\Psi OH}$: Flux consumption
 - Total $J = J_{ss} + v_{\Psi OH} J_{\Psi OH}$
- 2 parameters: I_p and P_{EC} at $t=50ms$



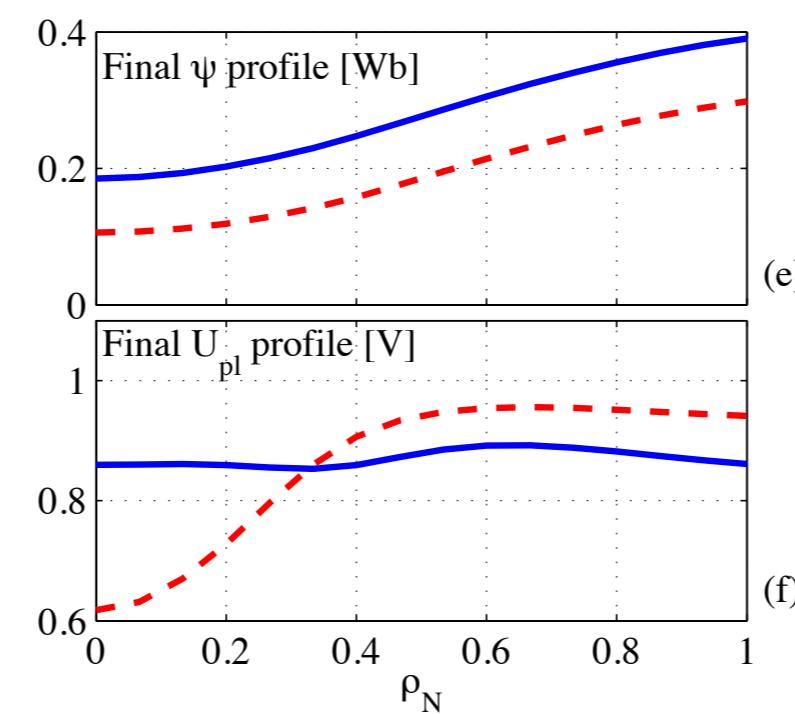
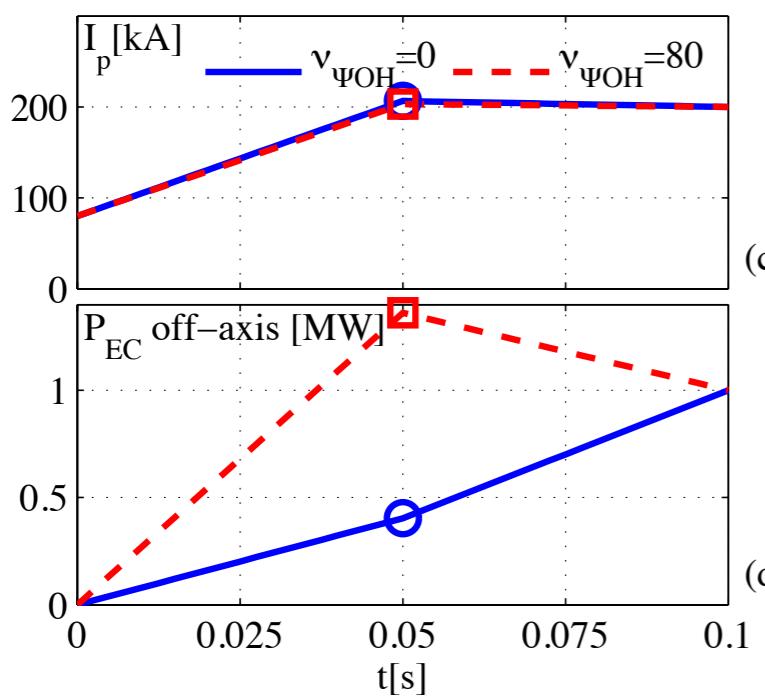
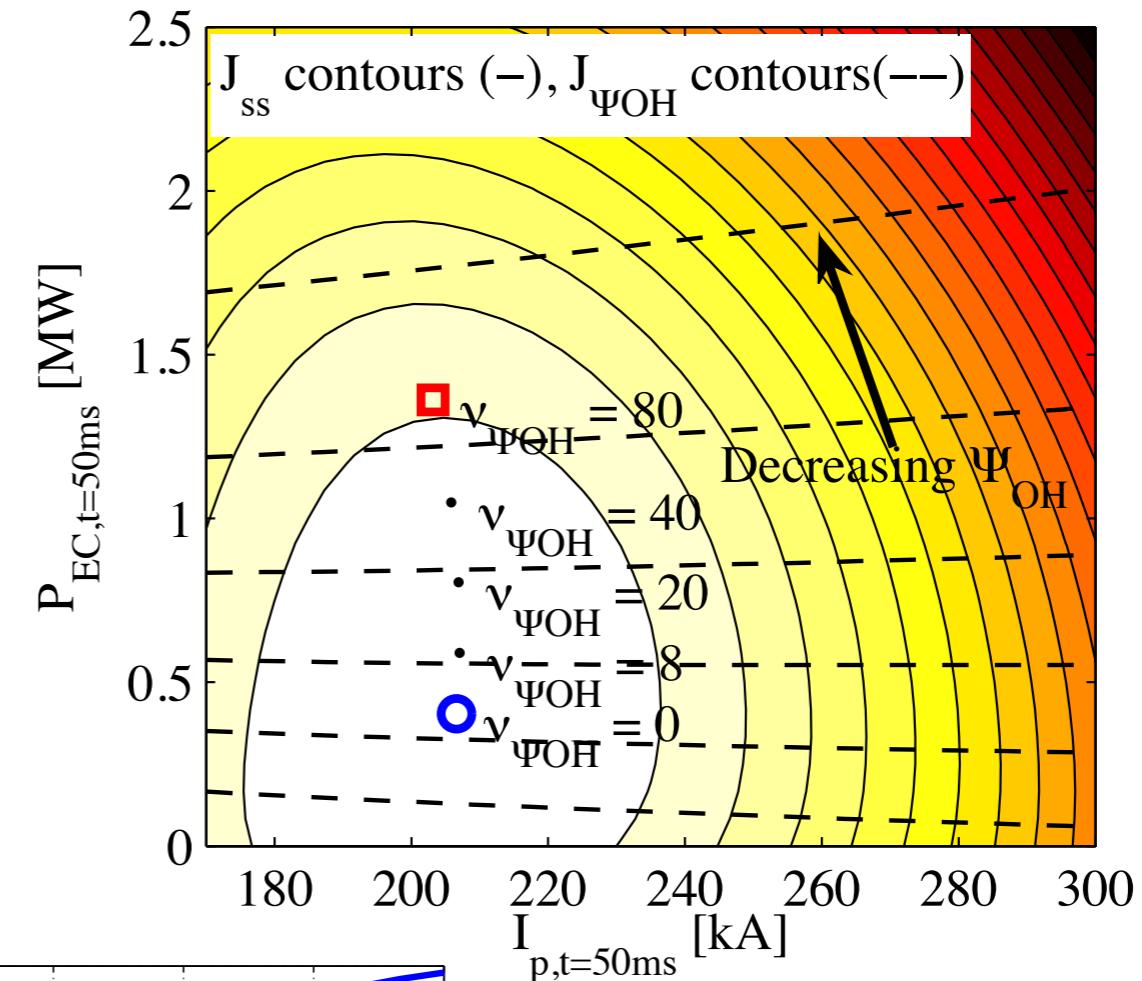
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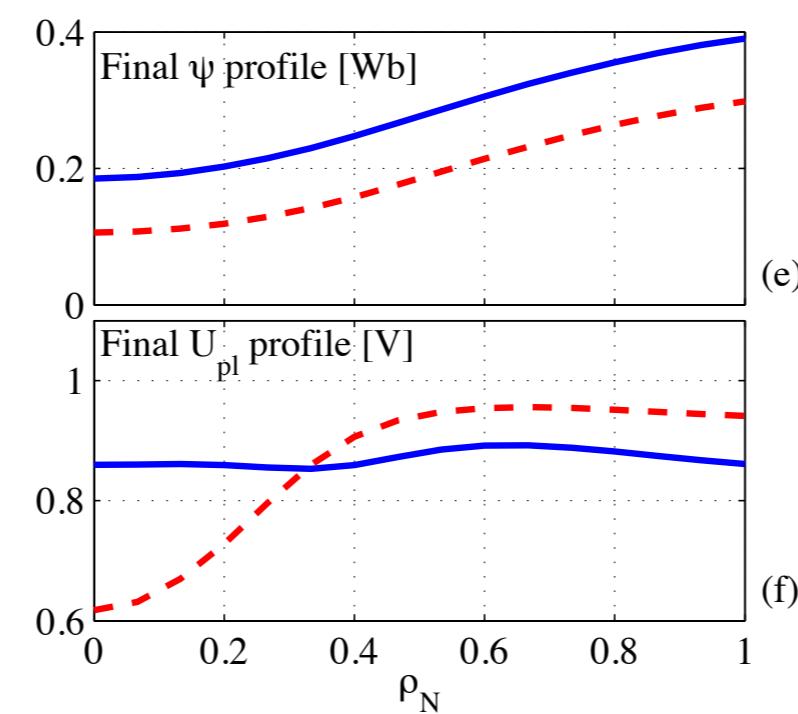
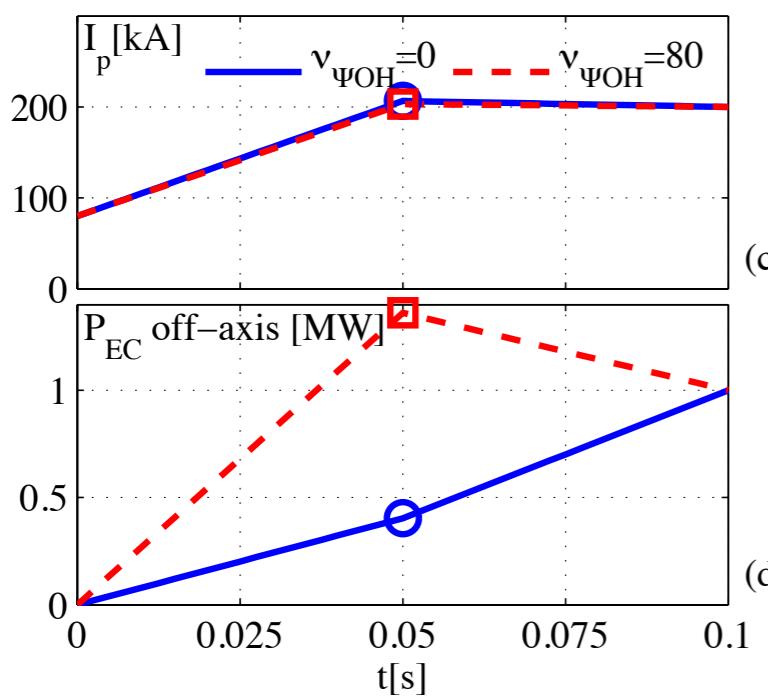
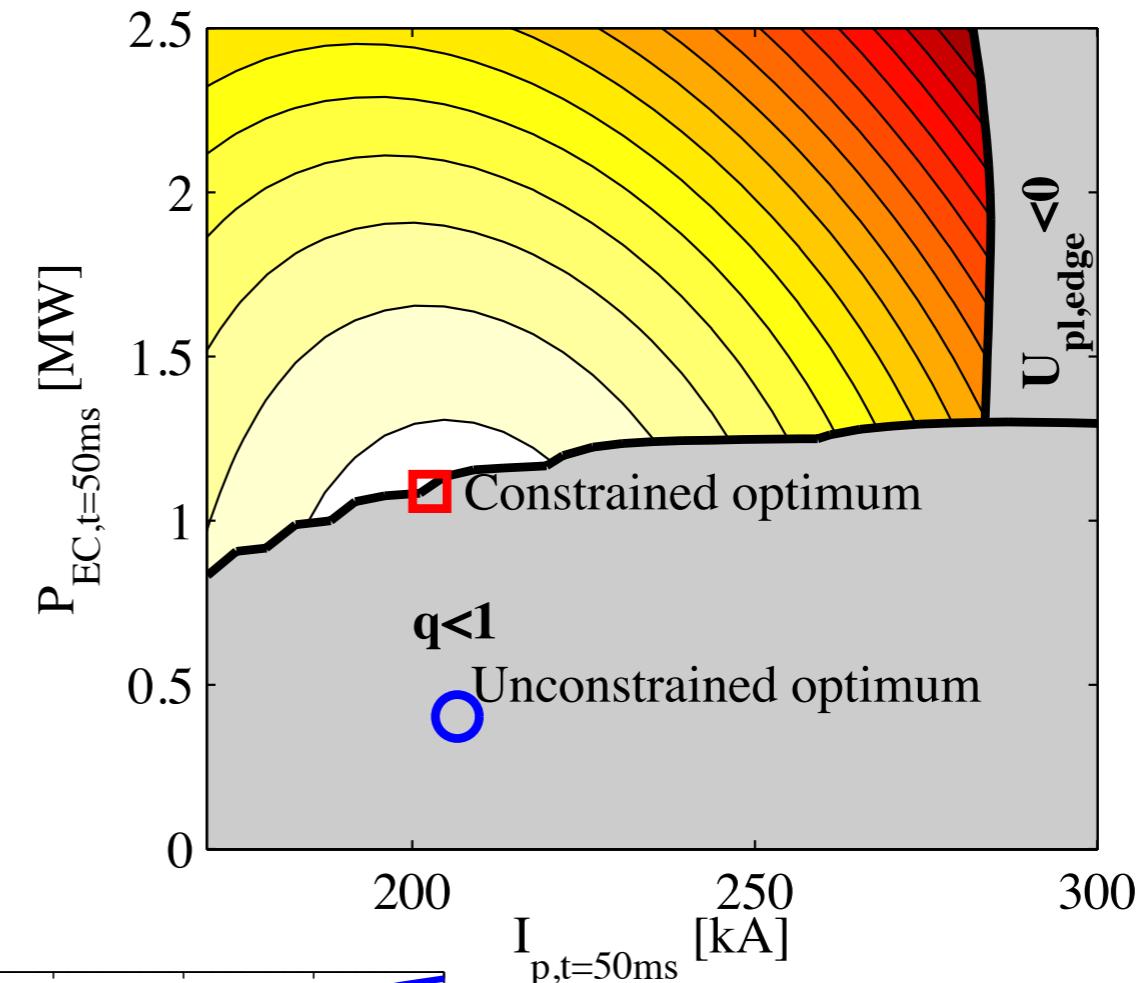
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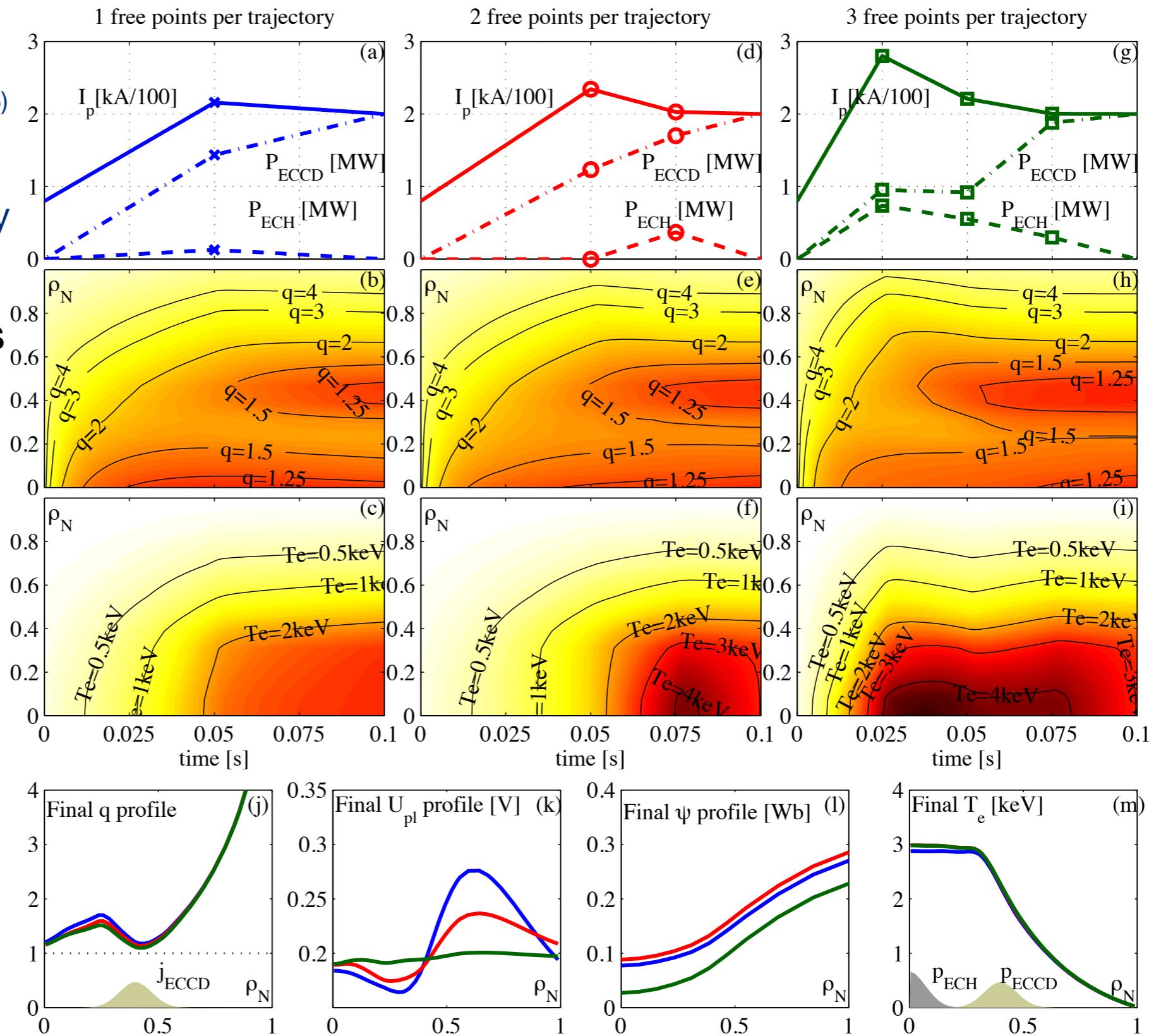
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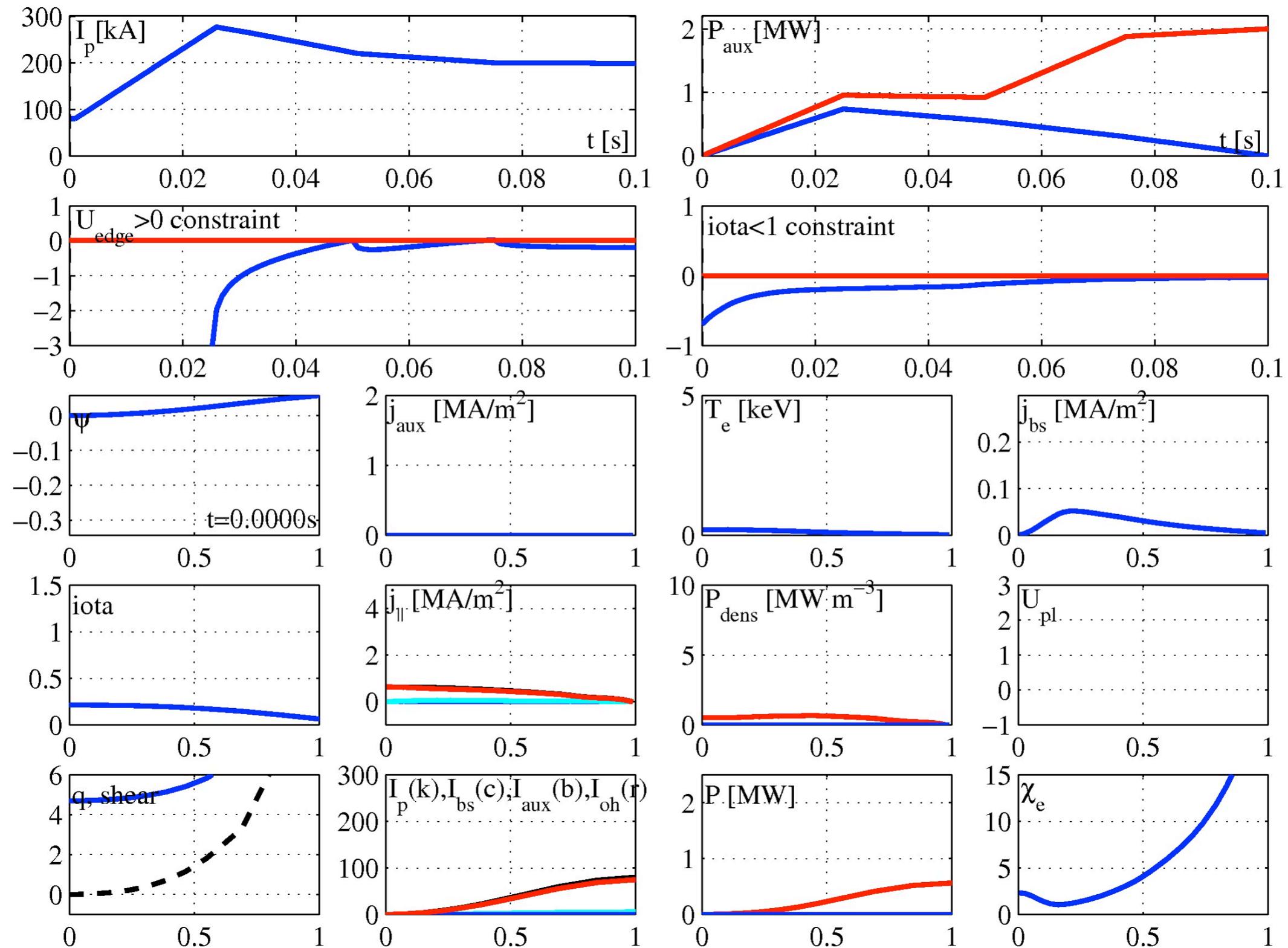


Results for ramp up to 'hybrid' q profile

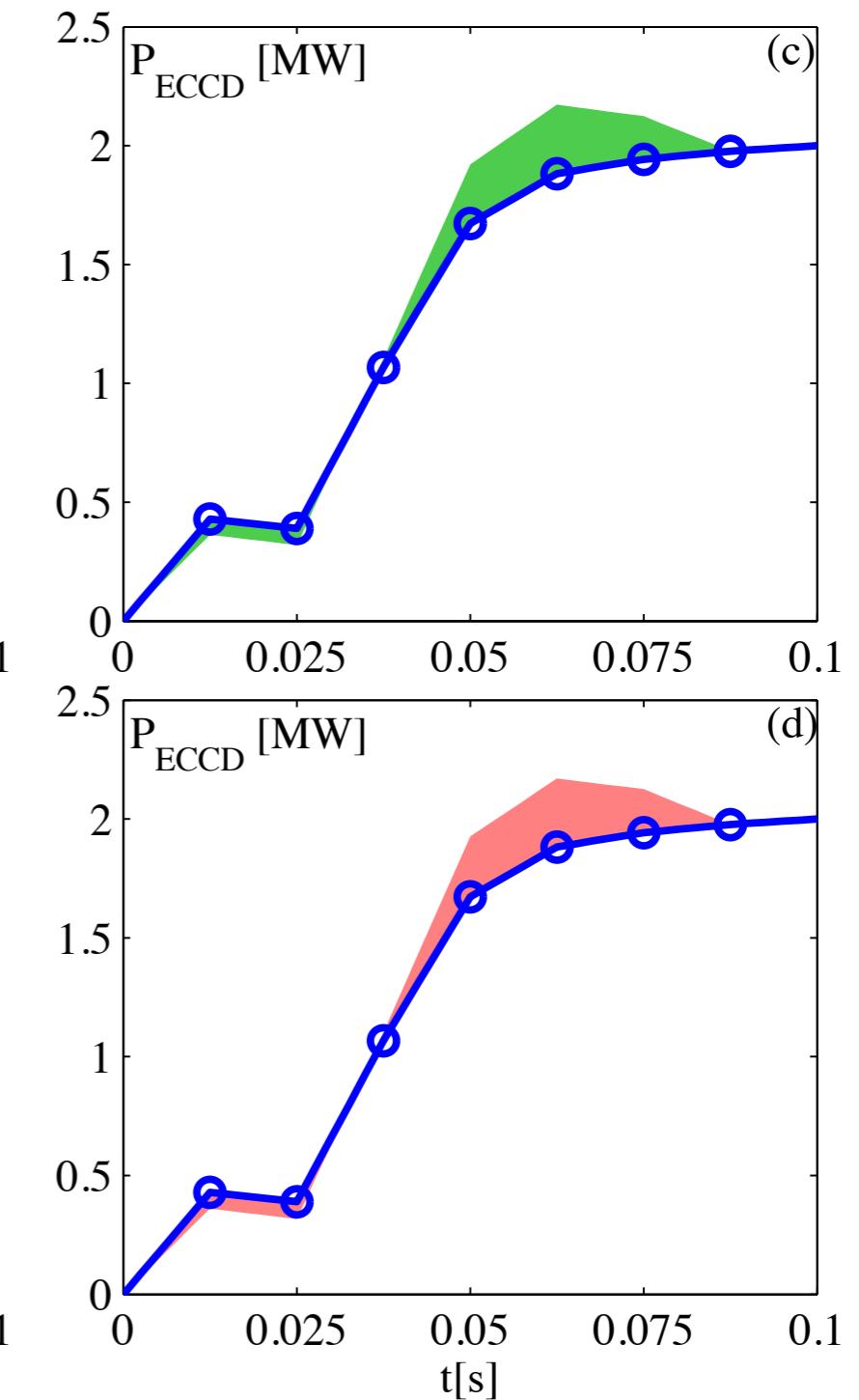
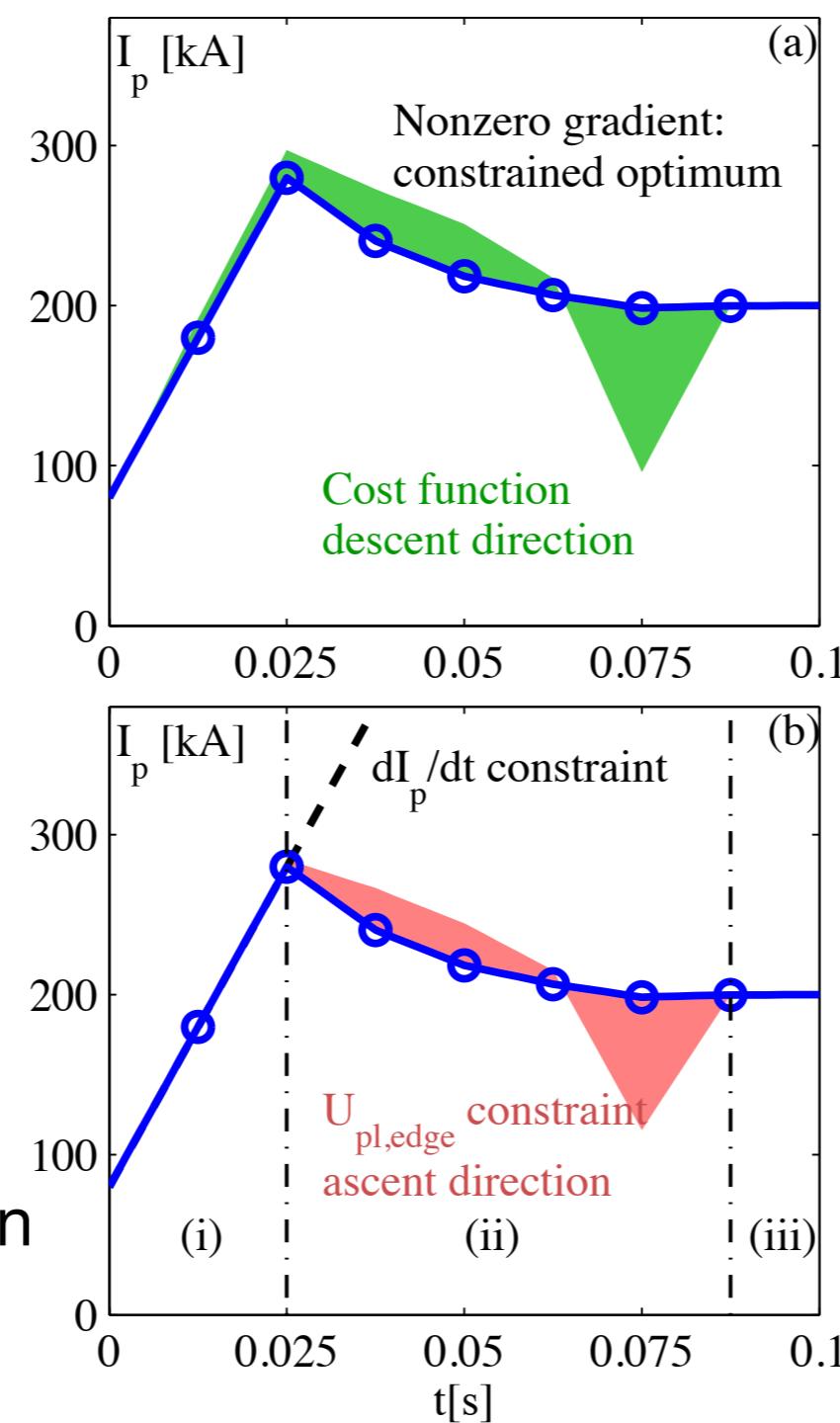
- Three actuators
 - I_p , $P_{EC}(\rho=0)$, $P_{ECCD}(\rho=0.3)$
- Two cost terms
 - Maximize stationarity
 - Minimize flux cons.
- Multiple constraints
 - $q > 1.1$, $V_{loop} > 0$
 - $dI_p/dt = 8 \text{ MA/s}$
 - $P_{max} = 2 \text{ MW}$, $P_{min} = 0$
- Optimal strategy
 - I_p overshoot
 - Early broad heating
 - Not too much early CD



Time evolution with optimal trajectories shows effect of actuator time evolution in flattening final v_{loop} profile



- Similar scenario, only $U_{pl,edge} > 0$ constraint
- Cost function gradient
 - Move in this direction to decrease cost
- Constraint gradient
 - Move in this direction to violate constraint
- Input arc classification
 - (i) Input constrained
 - (ii) State constrained
 - (iii) Unconstrained
- Consequences for feedback control design



Conclusions & Outlook

- New lightweight transport code RAPTOR for physics-based profile control
- Real-time simulation of q profiles demonstrated on TCV
 - q profiles every 1ms, without internal diagnostics!
 - Used for feedback control of T_e and L_i
 - Outlook
 - Closed loop control of q profile in advanced scenarios
 - Integrate with RT current density diagnostics
 - Couple to RT-equilibrium solver
- Optimization of actuator trajectories
 - Actuator trajectories tailored to get stationary profiles at start of flat-top.
 - Outlook
 - Further scenario optimization studies (advanced scenarios, ramp-down)
 - Add more physics (T_i , density, alpha profiles)
 - Real-time prediction
- Application to other tokamaks envisaged - collaborations are welcome

Thank you

Backup slides

Advanced scenario experiments are known to benefit from early I_p overshoot

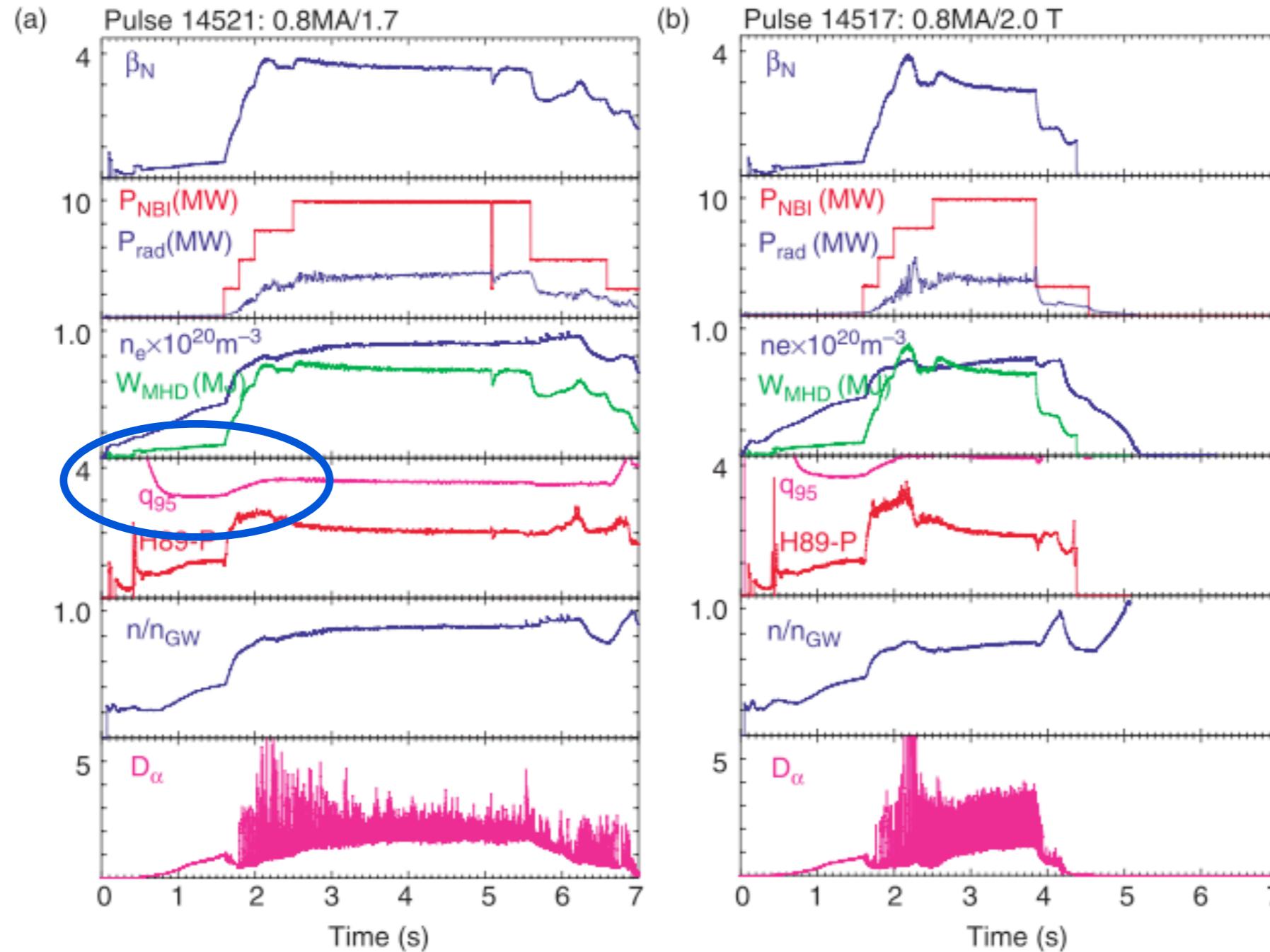
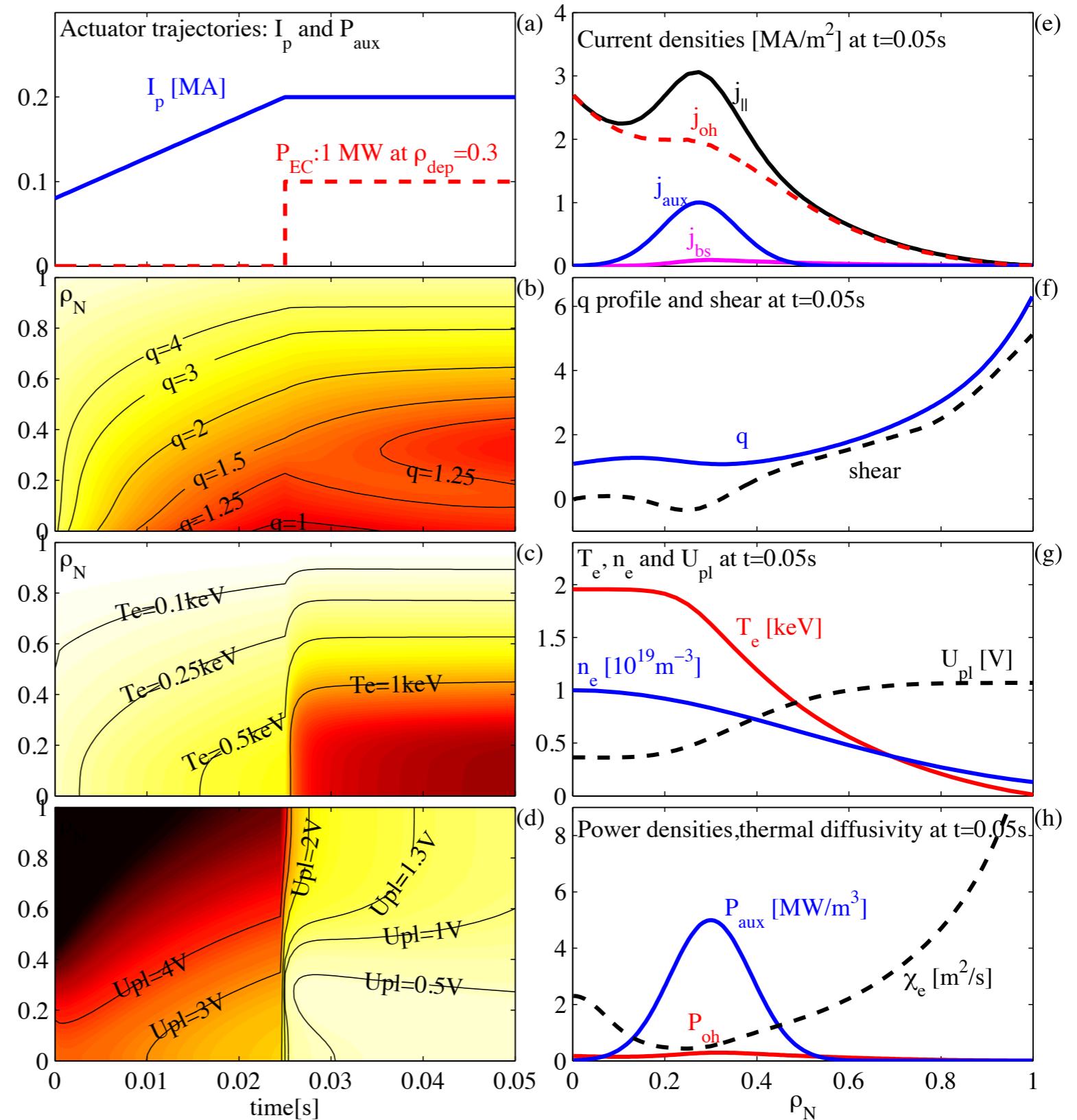


Figure 2. Two discharges that reach high β in ASDEX. (a) Pulse 14521 and (b) pulse 14517.

[Sips et al, *Progress towards steady-state advanced scenarios in ASDEX Upgrade*, PPCF 2002]

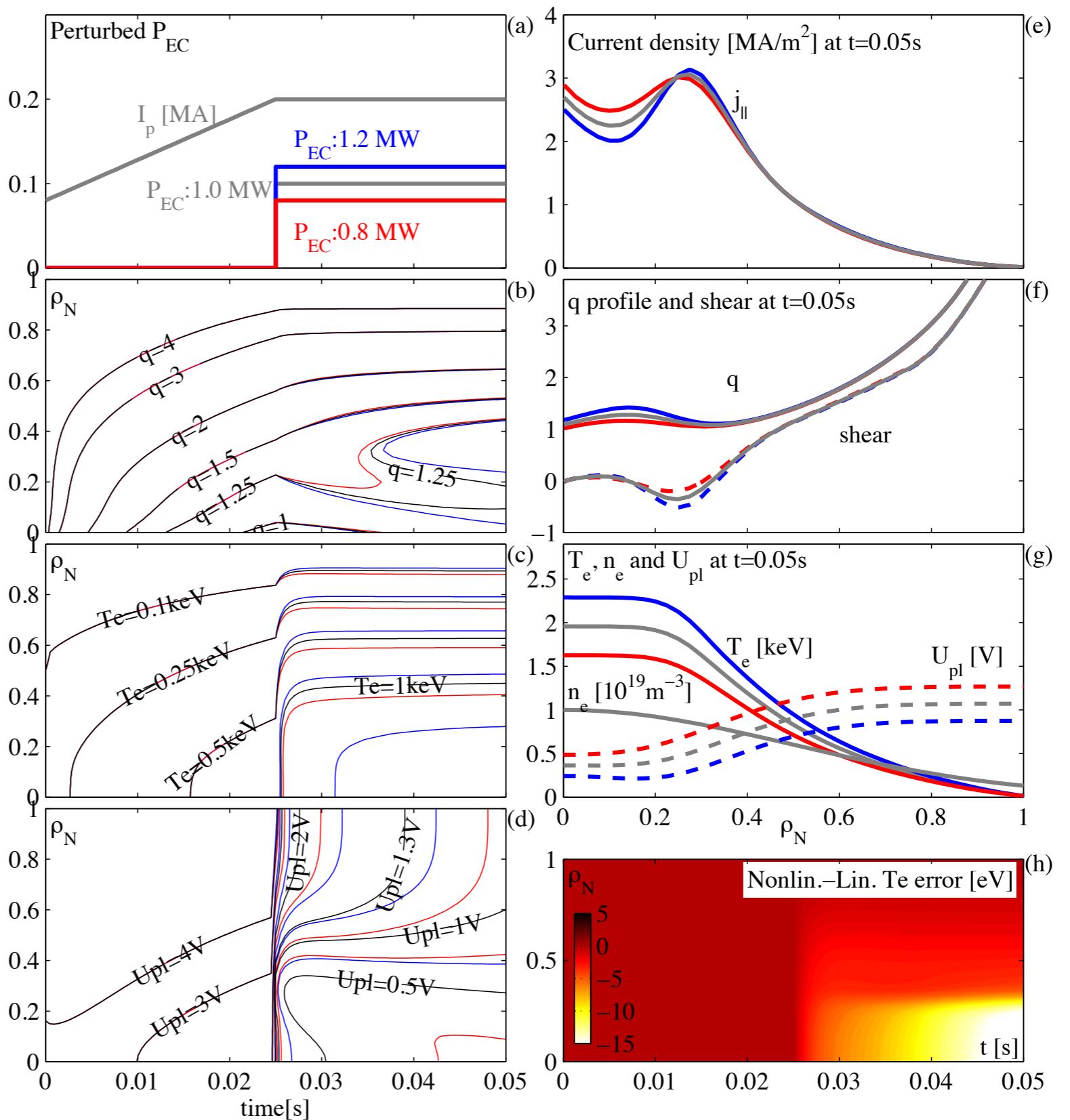
Example of TCV plasma simulation

- Simulation parameters
 - Equilibrium: existing shot
 - Transport model parameters: hand-picked to get reasonable profiles
- Ramp-up scenario
 - 80 to 200kA in 25ms
 - Sudden P_{EC} switch-on
- Some features of the profile evolution:
 - ~zero central shear profile
 - Transient, non-flat U_{pl} profile
 - Improved confinement at low magnetic shear
 - Low j_{BS} contribution
 - Back EMF at j_{CD} location



Evolution of perturbed trajectories

- Parameter: P_{EC} after 0.25
 - 1.2MW
 - 1MW (nominal)
 - 0.8MW
- Perturbed trajectories computed *without running new simulation*
 - Small error w.r.t. nonlinear case



Optimization problem: ingredients

- Cost function: reflects desired properties of **final profiles**

- Weighted sum of several profile terms
 - 1/safety factor $\|1/q(t_f) - 1/q_{ref}\|^2$ (e.g. for ITBs)
 - Loop voltage (e.g. for non-inductive scenarios) $\|U_{pl}(t_f) - U_{pl,ref}\|^2$
 - Loop voltage derivative (for steady-state) $\|dU_{pl}/dp\|^2$
 - Flux consumption (for longer pulse) $\|\Delta\Psi_{OH}\|^2$
 - Temperature (e.g. for high beta) $\|T_e(t_f) - T_{e,ref}\|^2$

$$J = \nu_\iota J_\iota + \nu_{U_{pl}} J_{U_{pl}} + \nu_{ss} J_{ss} + \nu_{OH} J_{OH} + \nu_{T_e} J_{T_e}$$

- Constraints: impose limitations on actuator and plasma evolution

- Constrain current ramp rate, maximum/minimum auxiliary power...
- Constrain minimum q: $q > q_{min}$ to avoid (e.g.) sawteeth
- Constrain edge loop voltage: $V_{loop} > 0$ to avoid negative edge currents
- Other constraints possible: shear, j_0 , ...

- Parametrize $u(t)$ with a finite number of parameters p using basis functions $P(t)$

$$u_i(t) = \sum_j^{n_i} P_{ij}(t) p_{i,j}$$

- Given state x_k , inputs u_k at time step k ,
 - PDE(ρ, t) -> discretize -> Nonlinear ODE at each time step:

$$\tilde{f}(x_{k+1}, x_k, u_k) = \tilde{f}_k = 0 \quad \forall k$$

- Take steps in Newton descent direction d
 - Need Jacobian
 - Obtained from analytical expression for all the derivatives (copious application of chain rule)
 - Iterate until residual $f_k < \text{tolerance}$
 - Go to next time step
- Store Jacobians at each time step

$$\mathcal{J}_{k+1}^k d = \tilde{f}_k,$$

$$\mathcal{J}_{k+1}^k = \frac{\partial \tilde{f}_k}{\partial x_{k+1}}$$

- Time evolution depends on mode parameters

- One example: a transport model parameter
- Another example: a parameter defining the input trajectory

$$\tilde{f}(x_{k+1}, x_k, u_k) = \tilde{f}_k = 0 \quad \forall k$$

- Differentiating with respect to parameter p , we get the *sensitivity equation*

$$0 = \frac{d\tilde{f}_k}{dp} = \frac{\partial \tilde{f}_k}{\partial x_{k+1}} \frac{\partial x_{k+1}}{\partial p} + \frac{\partial \tilde{f}_k}{\partial x_k} \frac{\partial x_k}{\partial p} + \frac{\partial \tilde{f}_k}{\partial u_k} \frac{\partial u_k}{\partial p} + \frac{\partial \tilde{f}_k}{\partial p}$$

- Linear ODE for dx_k/dp , solve while evolving nonlinear PDE: **Forward sensitivity analysis**
 - Jacobians df_k/dx_k , df_k/dx_{k+1} are known from Newton iterations
 - Computational cost proportional to p
-
- dx_k/dp gives the **linearization** of the state trajectories in the parameter space

$$T_e(\rho, t)|_{p=p_0+\delta p} \approx T_e(\rho, t)_{p_0} + \frac{\partial T_e}{\partial x} \frac{\partial x}{\partial p} \delta p$$

End