

SATISFIABILITY OF ELASTIC DEMAND IN THE SMART GRID

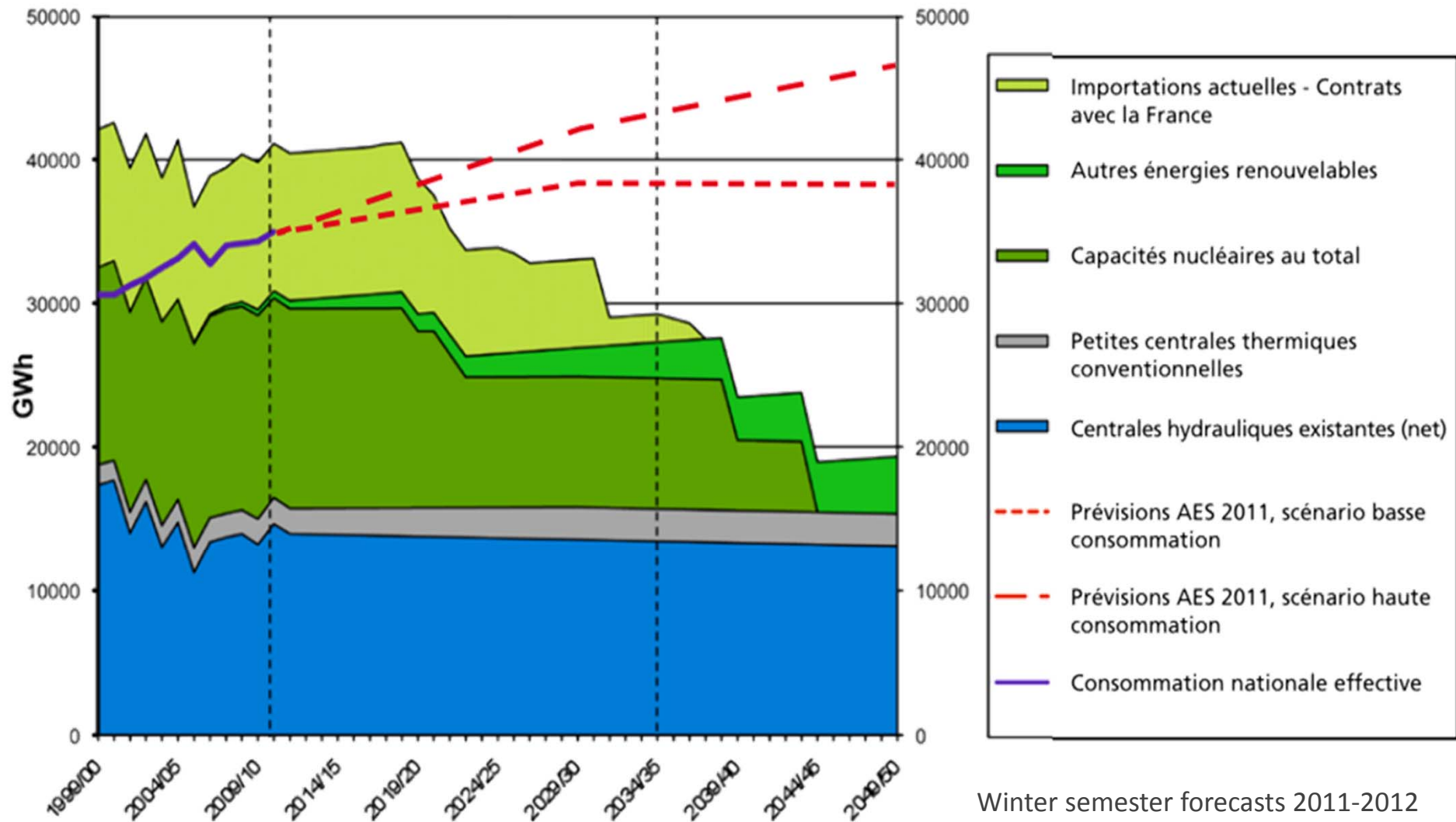
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EPFL

Energy Systems Day
Isaac Newton Institute
Cambridge, 2012 March 12

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2. A Model of Elastic Demand
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Swiss Future Supply is Uncertain



Winter semester forecasts 2011-2012

■ Source: ASE (Association des Entreprises Electriques Suisses,)₃

Smart Grid Vision : Larger Network

Larger Networks

A shift towards larger and larger aggregated networks and new, major interconnections between large load and generation centres over long distances (including continents)

Source:
ELECTRICITY SUPPLY
SYSTEMS OF THE
FUTURE
White paper on behalf
of the CIGRE Technical
Committee
TC Chair: Klaus Froelich
2011



Smart Grid Vision : Smaller, More Autonomous Networks

Smaller Networks

Greater shifts to distributed generation and localised solutions, a slowing and reversal of greater interconnections between grids and parts of grids, more self sufficiency and reduced reliance on generation sources large distances from load centres

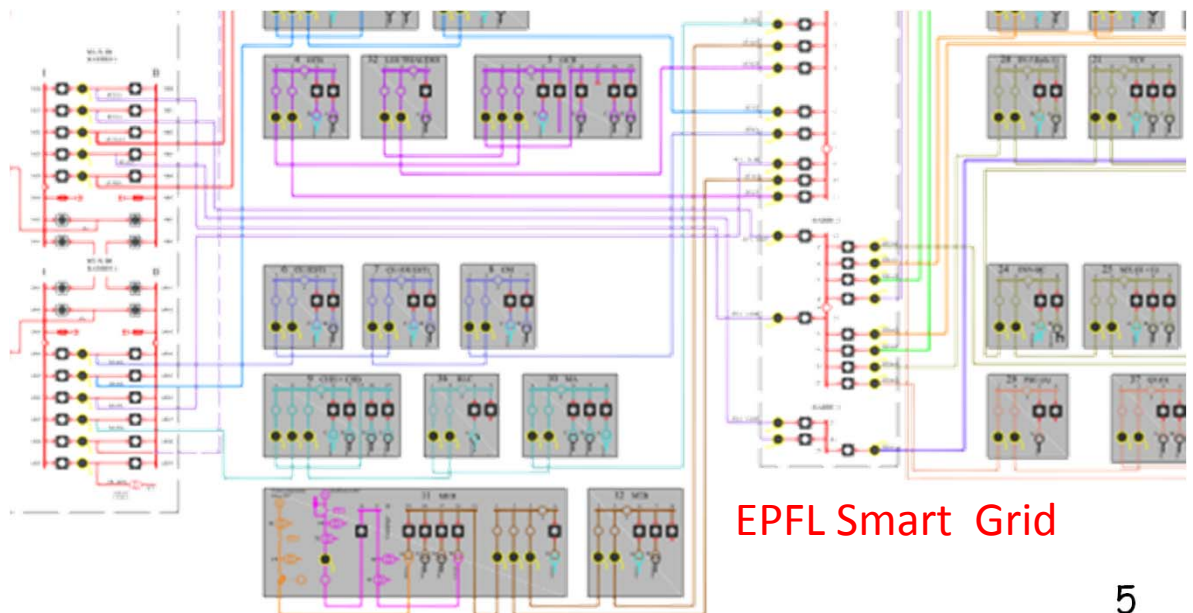
- Flexible services as alternative to blackout, including across slack bus
 - ▶ Islanded operation possible

Source:

ELECTRICITY SUPPLY
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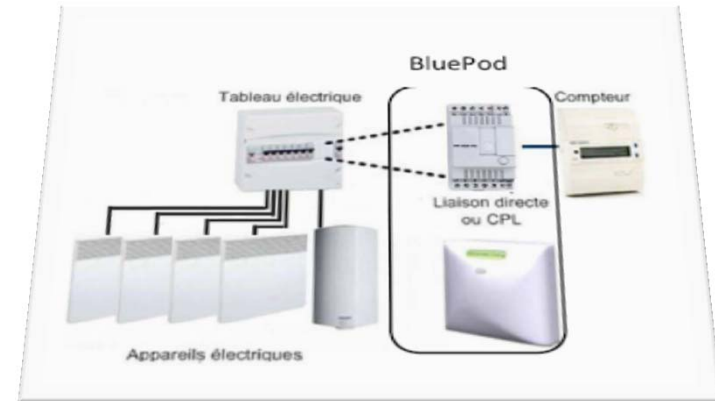
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EPFL Smart Grid

Flexible Services

- Flexible services = distribution network operator may interrupt / modulate power
- elastic loads support graceful degradation
- Thermal load (Votalis), washing machines (Romande Energie«commande centralisée») e-cars,



Voltalis Bluepod switches off thermal load for 60 mn



Our Problem Statement

- Do elastic services work ?

- ▶ Delays
- ▶ Returning load

- **Problem Statement**

Is there a control mechanism that can stabilize demand ?

- A very coarse (but fundamental) first step

- We leave out for now the details of signals and algorithms

2.

A MODEL OF ELASTIC DEMAND

Macroscopic Model of Cho and Meyn [1], non elastic demand, mapped to discrete time

Step 1: Day-ahead market

- Forecast demand:

$$D^f(t)$$

- Forecast supply:

$$G^f(t) = D^f(t) + r_0$$

Step 2: Real-time market

- Actual demand

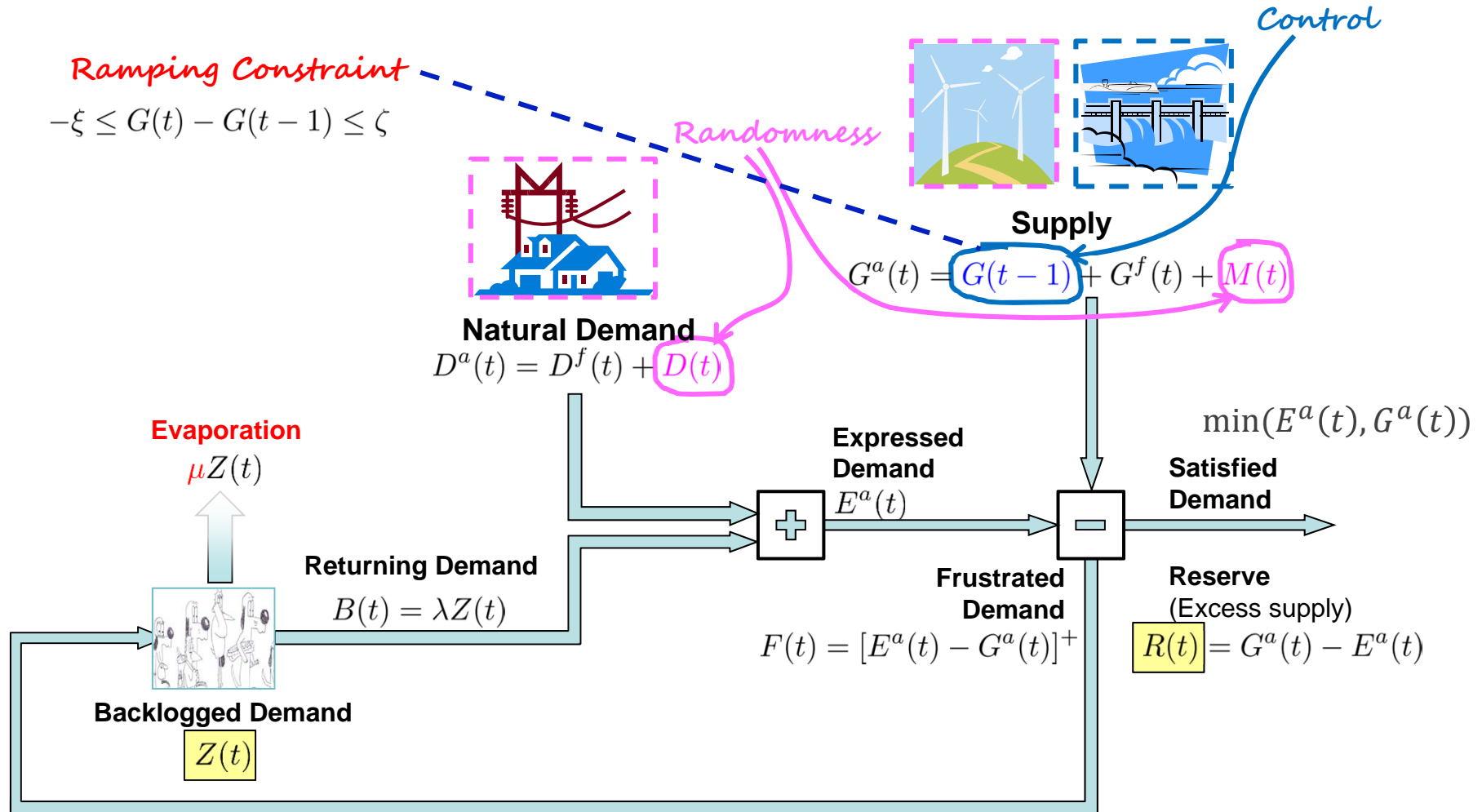
$$D^a(t) = D(t) \leftarrow + D^f(t) \rightarrow \text{random}$$

- Actual supply $G^a(t) =$

$$G(t-1) \leftarrow + G^f(t) + M(t) \rightarrow$$

- We now add the effect of elastic demand / flexible service
Some demand can be «frustrated» (delayed)

Our Macroscopic Model with Elastic Demand

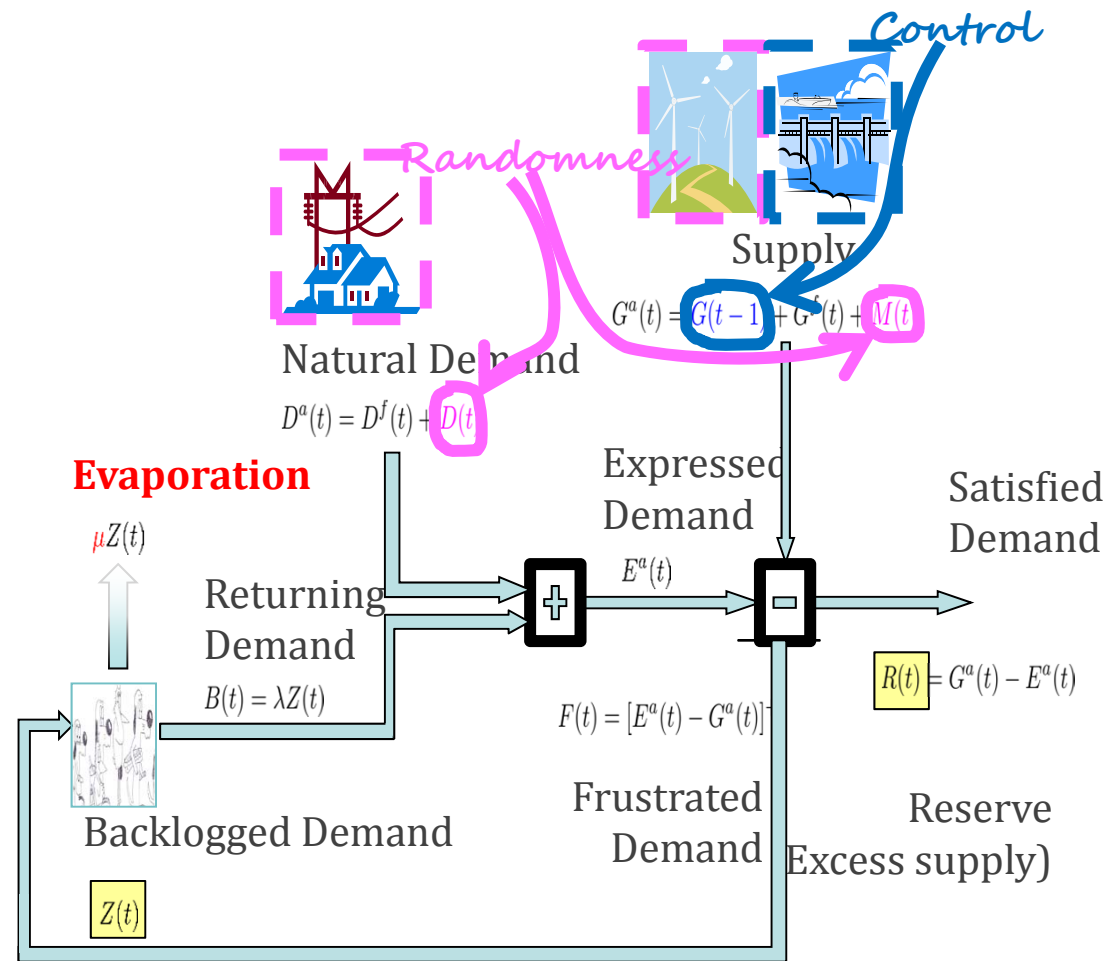


$$R(t) = G(t-1) - \lambda Z(t) + M(t) - D(t) + r_0$$

$$Z(t) = Z(t-1) - \lambda Z(t) - \mu Z(t) + \mathbb{1}_{\{R(t) < 0\}} |R(t)|$$

Backlogged Demand

- We assume backlogged demand is subject to two processes: update and re-submit
- Update term (evaporation): $\mu Z dt$ with $\mu > 0$ or $\mu < 0$
 μ is the evaporation rate (proportion lost per time slot)
- Re-submission term $\lambda Z dt$
 $1/\lambda$ (time slots) is the average delay

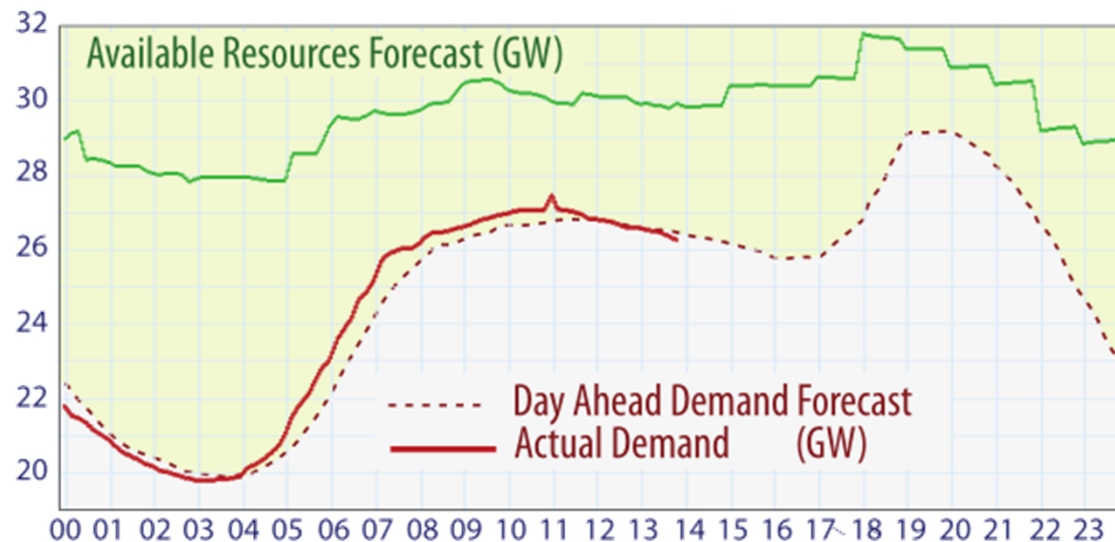


Macroscopic Model, continued

■ Assumption : $(M - D) = \text{ARIMA}(0, 1, 0)$

typical for deviation from forecast

$$(M(t+1) - D(t+1) - M(t) - D(t)) := N(t+1) \sim N(0, \sigma^2)$$



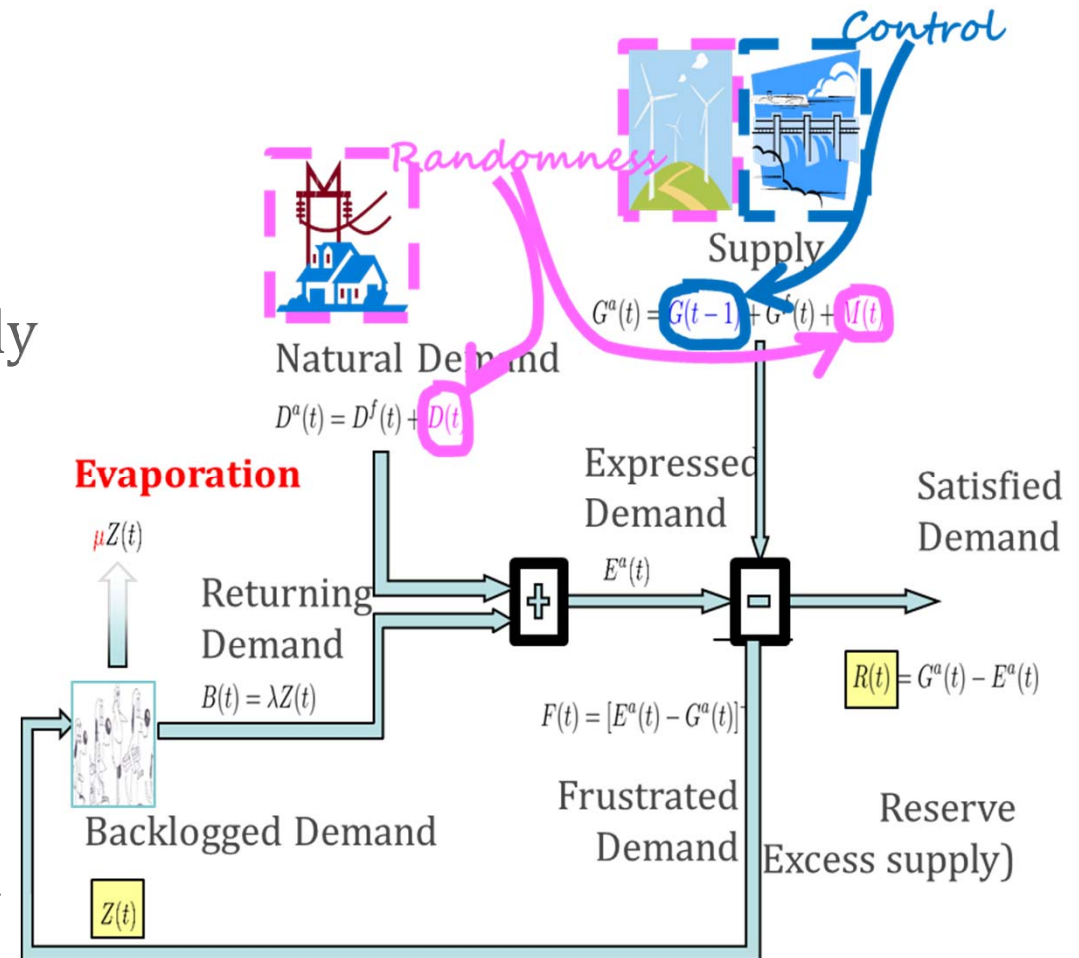
S. Meyn
“Dynamic Models and Dynamic Markets
for Electric Power Markets”

■ 2-d Markov chain on continuous state space

$$\begin{aligned} R(t+1) &= R(t) + \Delta G(t) + N(t+1) - \lambda[Z(t+1) - Z(t)] \\ Z(t+1) &= (1 - \lambda - \mu)Z(t) + \mathbb{1}_{\{R(t) < 0\}} R(t) \end{aligned}$$

The Control Problem

- **Control variable:**
 $G(t - 1)$
 production bought one
 time slot ago in real time
 market
- Controller sees only supply
 $G^a(t)$ and expressed
 demand $E^a(t)$
- **Our Problem:**
 keep backlog $Z(t)$ stable
- Ramp-up and ramp-down
 constraints
 $\xi \leq G(t) - G(t - 1) \leq \zeta$



Threshold Based Policies

$$G^f(t) = D^f(t) + r_0$$

Forecast supply is adjusted to
forecast demand

$$R(t) = G^a(t) - E^a(t)$$

$R(t)$:= reserve = excess of
demand over supply

Threshold policy:

if $R(t) < r^*$ * increase supply to come as close
to r^* as possible (considering ramp up
constraint)

else decrease supply to come as close to r^* as
possible (considering ramp down constraint)

Simulation

- Linearized system: 1 is eigenvalue

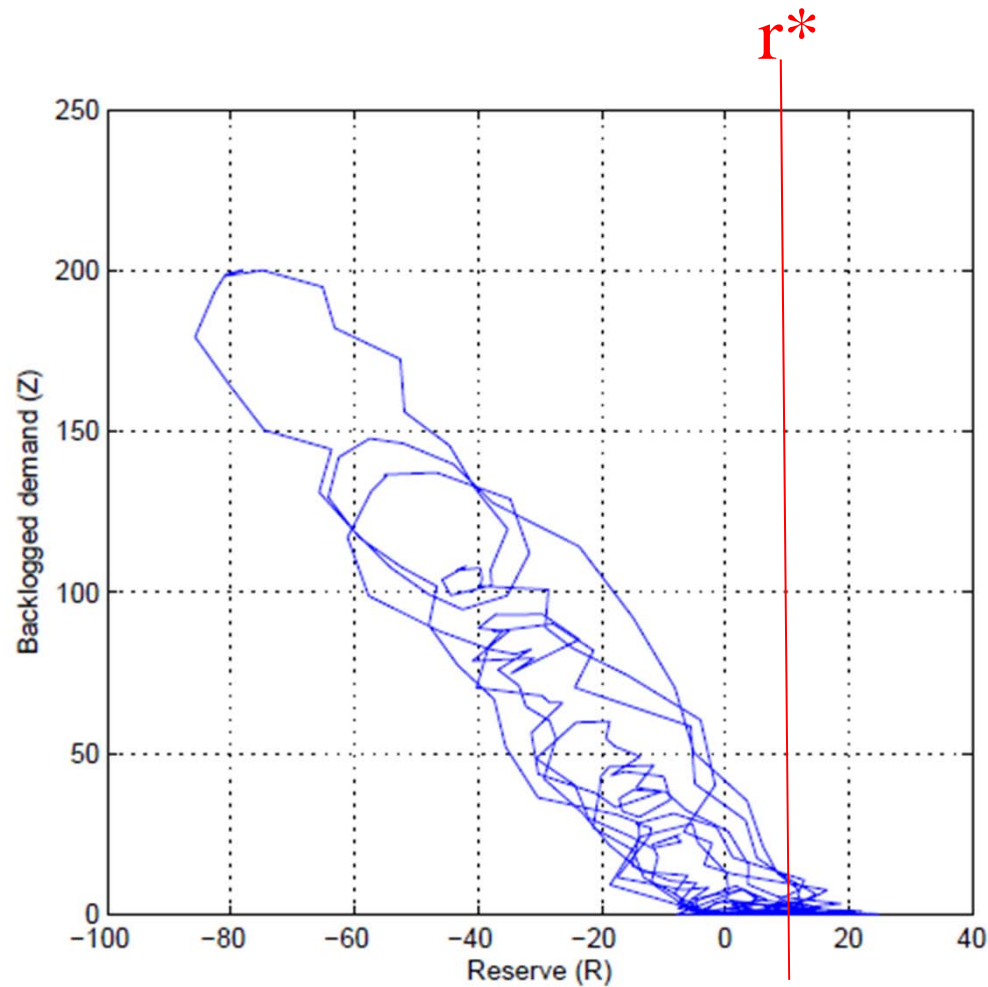


Figure 2. 500 iterations of the Markov process (13)-(14) for $\zeta = 1, r^* = 10, \sigma = 5, \lambda = 0.3, \mu = 0.1$

3.

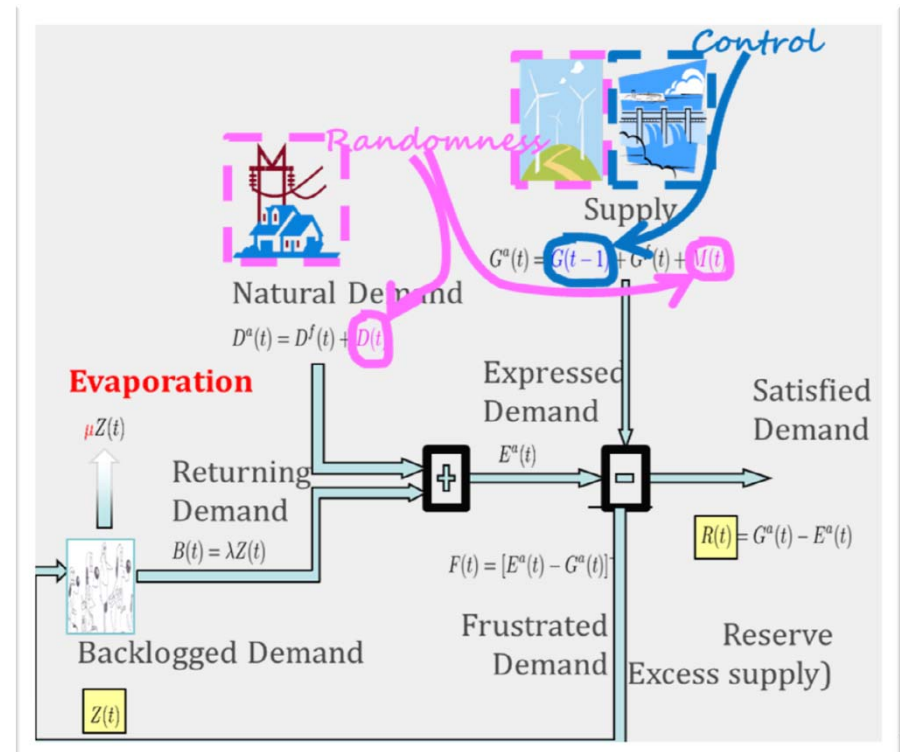
STABILITY RESULTS

Findings

- If evaporation μ is positive, system is stable (ergodic, positive recurrent Markov chain) for any threshold r^*

- If evaporation μ is negative, system unstable for any threshold r^*

- Delay does not play a role in stability
- Nor do ramp-up / ramp down constraints or size of reserve



More Detailed Findings

► Case 1: $\mu > 0$

Postponing a task = discount

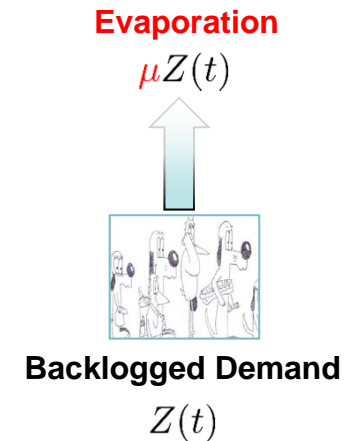
■ **Theorem 1:** The Markov chain (R, Z) is Harris recurrent and ergodic. It converges to the unique steady state probability distribution, for *any threshold and any strictly positive ramp-up constraint*.

► Case 2: $\mu < 0$

Postponing a task = penalty

■ **Theorem 2:** The Markov chain (R, Z) is non-positive, for *any threshold*.

Method of Proof: quadratic Lyapunov (case 1) or logarithmic L. (case 2)



Evaporation

■ *Negative* evaporation μ means:
delaying a load makes the
returning load larger than the
original one.

■ Could this happen ?

Q. Does letting your house cool down
now imply
spending more heat in total
compared to
keeping temperature constant ?

■ \neq return of the load:

Q. Does letting your house
cool down now imply
spending more heat later ?

A. Yes

(you will need to heat up
your house later -- delayed
load)

■ Assume the house model of [6]

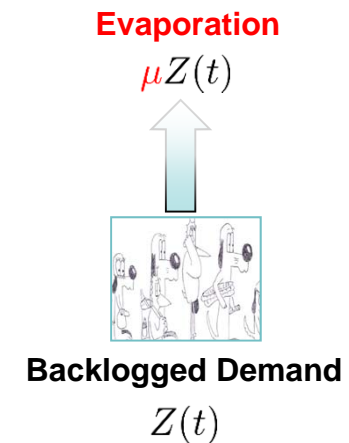
heat provided to building $d(t)\epsilon = \underbrace{K}_{\text{leakiness}}(T(t) - \underbrace{\theta(t)}_{\text{outside}}) + \underbrace{C}_{\text{inertia}}(T(t) - T(t-1))$

efficiency $\epsilon \underbrace{\sum_{t=1}^{\tau} d(t)}_{\text{E, total energy provided}} = K \sum_{t=1}^{\tau} (\underbrace{T(t)}_{\text{achieved } t^0} - \theta(t)) + C(T(\tau) - T(0))$

<i>Scenario</i>	<i>Optimal</i>	<i>Frustrated</i>
Building temperature	$T^*(t), t = 0 \dots \tau$	$T(t), t = 0 \dots \tau,$ $T(t) \leq T^*(t)$
Heat provided	$E^* = \frac{1}{\epsilon} \left(K \sum_{t=1}^{\tau} (T^*(t) - \theta(t)) + C(T^*(\tau) - T^*(0)) \right)$	$E < E^*$

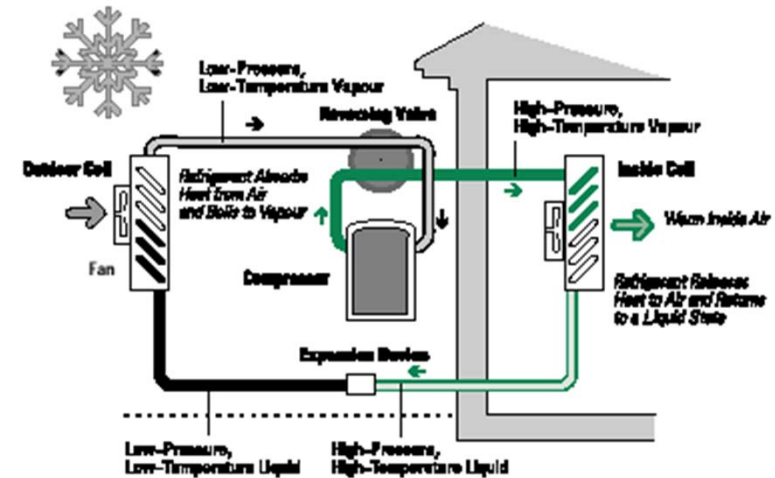
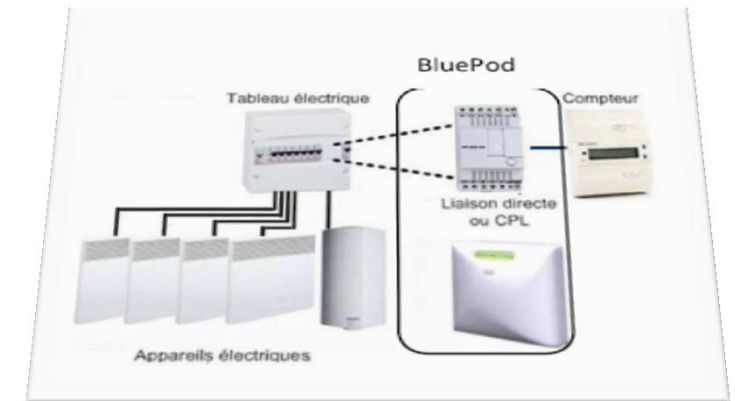
When Delayed Heating is Less Heat

- With constant coefficient of performance ϵ , total energy provided is less if let building cool down and warm up again
- Assume some demand is frustrated (second scenario)
update process replaces backlogged demand by *what is needed to recover the target temperature T^**
- Update process decreases backlog, evaporation is positive



The Sign of Evaporation

- Resistive heating system: evaporation is positive.
This is why Voltalis bluepod is accepted by users
- If heat = heat pump, coefficient of performance ϵ may be variable negative evaporation is possible
- Electric vehicle: delayed charge may have to be faster, less efficient, negative evaporation is possible



Conclusions

- A first model of adaptive appliances with volatile demand and supply
- Suggests that negative evaporation makes system unstable
Existing demand-response positive experience (with Voltalis/PeakSaver) might not carry over to other loads
- Model suggests that large backlogs are possible
Backlogged load is a new threat to grid operation
Need to measure and forecast backlogged load

Questions ?

- [1] Cho, Meyn – *Efficiency and marginal cost pricing in dynamic competitive markets with friction*, Theoretical Economics, 2010
- [2] Le Boudec, Tomozei, *Satisfiability of Elastic Demand in the Smart Grid*, Energy 2011 and ArXiv.1011.5606
- [3] Le Boudec, Tomozei, *Demand Response Using Service Curves*, IEEE ISGT-EUROPE, 2011
- [4] Le Boudec, Tomozei, *A Demand-Response Calculus with Perfect Batteries*, WoNeCa, 2012
- [5] Papavasiliou, Oren - *Integration of Contracted Renewable Energy and Spot Market Supply to Serve Flexible Loads*, 18th World Congress of the International Federation of Automatic Control, 2011
- [6] David MacKay, *Sustainable Energy – Without the Hot Air*, UIT Cambridge, 2009