# From Bad Models to Good Policies: an Intertwined Story about Energy and Reinforcement Learning 

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## Outline



## Intertwined Stories





## Deterministic RL

- Dynamics $x_{t+1}=f\left(x_{t}, u_{t}\right) \quad t=0, \ldots, T-1 \quad T \in \mathbb{N} \backslash\{0\}$

$$
\mathcal{X} \subset \mathbb{R}^{d} \quad \mathcal{U}=\left\{u^{(1)}, \ldots, u^{(m)}\right\}
$$

- Reward function $r_{t}=\rho\left(x_{t}, u_{t}\right) \in \mathbb{R}$
- Return

$$
\forall\left(u_{0}, \ldots, u_{T-1}\right) \in \mathcal{U}^{T}, \quad J\left(u_{0}, \ldots, u_{T-1}\right) \triangleq \sum_{t=0}^{T-1} \rho\left(x_{t}, u_{t}\right)
$$

- Optimality $J_{T}^{*} \triangleq \max _{\left(u_{0}, \ldots, u_{T-1}\right) \in \mathcal{U}^{T}} J\left(u_{0}, \ldots, u_{T-1}\right)$


## Batch Mode RL

- Dynamics and reward function are unknown
- Instead, we have access to trajectories (« bad model »):

$$
\begin{gathered}
\mathcal{F}^{(u)}=\left\{\left(x^{(u), k}, r^{(u), k}, y^{(u), k}\right)\right\}_{k=1}^{n^{(u)}} \\
y^{(u), k}=f\left(x^{(u), k}, u\right) \quad r^{(u), k}=\rho\left(x^{(u), k}, u\right) \\
\forall u \in \mathcal{U}, n^{(u)}>0 \quad \mathcal{F}=\mathcal{F}^{(1)} \cup \ldots \cup \mathcal{F}^{(m)}
\end{gathered}
$$

## Lipschitz Continuity

$$
\begin{aligned}
\forall\left(x, x^{\prime}\right) \in \mathcal{X}^{2}, \forall u \in \mathcal{U}, \quad\left\|f(x, u)-f\left(x^{\prime}, u\right)\right\| & \leq L_{f}\left\|x-x^{\prime}\right\| \\
\left|\rho(x, u)-\rho\left(x^{\prime}, u\right)\right| & \leq L_{\rho}\left\|x-x^{\prime}\right\|
\end{aligned}
$$

$$
L_{f}, L_{\rho} \in \mathbb{R}
$$

## Lipschitz Compatibility

$$
\begin{aligned}
& \mathcal{L}_{\mathcal{F}}^{f}=\left\{f^{\prime}: \mathcal{X} \times \mathcal{U} \rightarrow \mathcal{X} \left\lvert\,\left\{\begin{array}{l}
\forall x^{\prime}, x^{\prime \prime} \in \mathcal{X}, \forall u \in \mathcal{U}, \\
\left\|\prime^{\prime}\left(x^{\prime}, u\right)-f^{\prime}\left(x^{\prime \prime}, u\right)\right\| \leq L_{f}\left\|x^{\prime}-x^{\prime \prime}\right\|, \\
\forall k \in\left\{1, \ldots, n^{(u)}\right\}, f^{\prime}\left(x^{(u), k}, u\right)=f\left(x^{(u), k}, u\right)=y^{(u), k}
\end{array}\right\}\right.\right. \\
& \mathcal{L}_{\mathcal{F}}^{\rho}=\left\{\rho^{\prime}: \mathcal{X} \times \mathcal{U} \rightarrow \mathbb{R} \left\lvert\,\left\{\begin{array}{l}
\forall x^{\prime}, x^{\prime \prime} \in \mathcal{X}, \forall u \in \mathcal{U}, \\
\left|\rho^{\prime}\left(x^{\prime}, u\right)-\rho^{\prime}\left(x^{\prime \prime}, u\right)\right| \leq L_{\rho}\left\|x^{\prime}-x^{\prime \prime}\right\|, \\
\forall k \in\left\{1, \ldots, n^{(u)}\right\}, \rho^{\prime}\left(x^{(u), k}, u\right)=\rho\left(x^{(u), k}, u\right)=r^{(u), k}
\end{array}\right\}\right.\right.
\end{aligned}
$$

$$
\forall\left(f^{\prime}, \rho^{\prime}\right) \in \mathcal{L}_{\mathcal{F}}^{f} \times \mathcal{L}_{\mathcal{F}}^{\rho}, J_{\left(f^{\prime}, \rho^{\prime}\right)}\left(u_{0}, \ldots, u_{T-1}\right)=\sum_{t=0}^{T-1} \rho^{\prime}\left(x_{t}^{\prime}, u_{t}\right)
$$

$$
x_{t+1}^{\prime}=f^{\prime}\left(x_{t}^{\prime}, u_{t}\right)
$$

## Minmax Generalization

- Define:

$$
B^{*}\left(\mathcal{F}, u 0, \ldots, u_{T-1}\right)=\min _{\left(f^{\prime}, \rho^{\prime}\right) \in \mathcal{L}_{\mathcal{F}}^{f} \times \mathcal{L}_{\mathcal{F}}^{\rho}} J_{\left(f^{\prime}, \rho^{\prime}\right)\left(u_{0}, \ldots, u_{T-1}\right)}
$$

- The minmax generalization solution is defined as:

$$
\left(u_{0}, \ldots, u_{T-1}\right) \in \underset{\left(u_{0}, \ldots, u_{T-1}\right) \in \mathcal{U}^{T}}{\arg \max } B^{*}\left(\mathcal{F}, u 0, \ldots, u_{T-1}\right)
$$

- Here, we focus on the min part


## Minmax Generalization

$$
\left(\mathcal{P}\left(\mathcal{F}, L_{f}, L_{\rho}, x_{0}, u_{0}, \ldots, u_{T-1}\right)\right):
$$

$$
\begin{array}{lll} 
& & \min _{\hat{r}^{2}} \\
\begin{array}{ccc}
\hat{\mathbf{r}}_{0} & \ldots & \hat{\mathbf{r}}_{T-1} \in \mathbb{R} \\
\hat{\mathbf{x}}_{0} & \ldots & \hat{\mathbf{x}}_{T-1} \in \mathcal{X}
\end{array} & \sum_{t=0}^{T-1} \hat{\mathbf{r}}_{t}, \\
\end{array}
$$

subject to

$$
\begin{align*}
& \left|\hat{\mathbf{r}}_{t}-r^{\left(u_{t}\right), k_{t}}\right|^{2} \leq L_{\rho}^{2}\left\|\hat{\mathbf{x}}_{t}-x^{\left(u_{t}\right), k_{t}}\right\|^{2}, \forall\left(t, k_{t}\right) \in\{0, \ldots, T-1\} \times\left\{1, \ldots, n^{\left(u_{t}\right)}\right\} \\
& \left\|\hat{\mathbf{x}}_{t+1}-y^{\left(u_{t}\right), k_{t}}\right\|^{2} \leq L_{f}^{2}\left\|\hat{\mathbf{x}}_{t}-x^{\left(u_{t}\right), k_{t}}\right\|^{2}, \forall\left(t, k_{t}\right) \in\{0, \ldots, T-1\} \times\left\{1, \ldots, n^{\left(u_{t}\right)}\right\}  \tag{3.2}\\
& \left|\hat{\mathbf{r}}_{t}-\hat{\mathbf{r}}_{t^{\prime}}\right|^{2} \leq L_{\rho}^{2}\left\|\hat{\mathbf{x}}_{t}-\hat{\mathbf{x}}_{t^{\prime}}\right\|^{2}, \forall t, t^{\prime} \in\left\{0, \ldots, T-1 \mid u_{t}=u_{t^{\prime}}\right\}  \tag{3.3}\\
& \left\|\hat{\mathbf{x}}_{t+1}-\hat{\mathbf{x}}_{t^{\prime}+1}\right\|^{2} \leq L_{f}^{2}\left\|\hat{\mathbf{x}}_{t}-\hat{\mathbf{x}}_{t^{\prime}}\right\|^{2}, \forall t, t^{\prime} \in\left\{0, \ldots, T-2 \mid u_{t}=u_{t^{\prime}}\right\}  \tag{3.4}\\
& \hat{\mathbf{x}}_{0}=x_{0} \tag{3.5}
\end{align*}
$$




$)^{3}+\frac{3}{x}+5$




## Minmax Generalization

$$
\left(\mathcal{P}\left(\mathcal{F}, L_{f}, L_{\rho}, x_{0}, u_{0}, \ldots, u_{T-1}\right)\right):
$$

$$
\begin{array}{llll} 
& & \min _{\hat{\mathbf{r}}_{T-1} \in \mathbb{R}} & \sum_{t=0}^{T-1} \hat{\mathbf{r}}_{t}, \\
\hat{\mathbf{r}}_{0} & \ldots & \hat{\mathbf{x}}_{T-1} \\
\hat{\mathbf{x}}_{0} & \ldots & \hat{\mathbf{x}}_{T-1} \in \mathcal{X}
\end{array}
$$

subject to

$$
\begin{align*}
& \left|\hat{\mathbf{r}}_{t}-r^{\left(u_{t}\right), k_{t}}\right|^{2} \leq L_{\rho}^{2}\left\|\hat{\mathbf{x}}_{t}-x^{\left(u_{t}\right), k_{t}}\right\|^{2}, \forall\left(t, k_{t}\right) \in\{0, \ldots, T-1\} \times\left\{1, \ldots, n^{\left(u_{t}\right)}\right\} \\
& \left\|\hat{\mathbf{x}}_{t+1}-y^{\left(u_{t}\right), k_{t}}\right\|^{2} \leq L_{f}^{2}\left\|\hat{\mathbf{x}}_{t}-x^{\left(u_{t}\right), k_{t}}\right\|^{2}, \forall\left(t, k_{t}\right) \in\{0, \ldots, T-1\} \times\left\{1, \ldots, n^{\left(u_{t}\right)}\right\},  \tag{3.2}\\
& \left|\hat{\mathbf{r}}_{t}-\hat{\mathbf{r}}_{t^{\prime}}\right|^{2} \leq L_{\rho}^{2}\left\|\hat{\mathbf{x}}_{t}-\hat{\mathbf{x}}_{t^{\prime}}\right\|^{2}, \forall t, t^{\prime} \in\left\{0, \ldots, T-1 \mid u_{t}=u_{t^{\prime}}\right\}  \tag{3.3}\\
& \left\|\hat{\mathbf{x}}_{t+1}-\hat{\mathbf{x}}_{t^{\prime}+1}\right\|^{2} \leq L_{f}^{2}\left\|\hat{\mathbf{x}}_{t}-\hat{\mathbf{x}}_{t^{\prime}}\right\|^{2}, \forall t, t^{\prime} \in\left\{0, \ldots, T-2 \mid u_{t}=u_{t^{\prime}}\right\}  \tag{3.4}\\
& \hat{\mathbf{x}}_{0}=x_{0} \tag{3.5}
\end{align*}
$$

## Minmax Generalization

- One can show that constraint (3.3) are redundant

LEMMA 4.1. Consider $\left(\hat{\mathbf{r}}^{*}, \hat{\mathbf{x}}^{*}\right) \in \mathbb{R}^{T} \times \mathcal{X}^{T}$ an optimal solution to $\overline{\mathcal{P}}\left(\mathcal{F}, u_{0}, \ldots, u_{T-1}\right)$. Then, for all $t, t^{\prime}$ such that $u_{t}=u_{t^{\prime}}$,

$$
\left|\hat{\mathbf{r}}_{t}^{*}-\hat{\mathbf{r}}_{t^{\prime}}^{*}\right|^{2} \leq L_{\rho}^{2}\left\|\hat{\mathbf{x}}_{t}^{*}-\hat{\mathbf{x}}_{t^{\prime}}^{*}\right\|^{2}
$$

- In particular, this implies that optimal reward for the first stage ( $\mathrm{t}=0$ ) can also be computed

Lemma 4.2. The solution of the problem $\left(\mathcal{P}^{\prime}\left(\mathcal{F}, u_{0}\right)\right)$ is

$$
\hat{\mathbf{r}}_{0}^{*}=\max _{k_{0} \in\left\{1, \ldots, n^{\left(u_{0}\right)}\right\}} r^{\left(u_{0}\right), k_{0}}-L_{\rho}\left\|x_{0}-x^{\left(u_{0}\right), k_{0}}\right\| .
$$

## Minmax Generalization

$\left(\mathcal{P}^{\prime \prime}\left(\mathcal{F}, u_{0}, \ldots, u_{T-1}\right)\right):$

$$
\begin{array}{lll} 
& & \min _{\hat{\mathbf{r}}^{2}} \\
\hat{\mathbf{r}}_{1} & \ldots & \sum_{T-1} \in \mathbb{R} \\
\hat{\mathbf{x}}_{0} & \ldots & \hat{\mathbf{x}}_{T-1} \in \mathcal{X}
\end{array} \quad \hat{\mathbf{r}}_{t},
$$

subject to

$$
\begin{align*}
& \left|\hat{\mathbf{r}}_{t}-r^{\left(u_{t}\right), k_{t}}\right|^{2} \leq L_{\rho}^{2}\left\|\hat{\mathbf{x}}_{t}-x^{\left(u_{t}\right), k_{t}}\right\|^{2}, \forall\left(t, k_{t}\right) \in\{1, \ldots, T-1\} \times\left\{1, \ldots, n^{\left(u_{t}\right)}\right\} \\
& \left\|\hat{\mathbf{x}}_{t+1}-y^{\left(u_{t}\right), k_{t}}\right\|^{2} \leq L_{f}^{2}\left\|\hat{\mathbf{x}}_{t}-x^{\left(u_{t}\right), k_{t}}\right\|^{2}, \forall\left(t, k_{t}\right) \in\{0, \ldots, T-1\} \times\left\{1, \ldots, n^{\left(u_{t}\right)}\right\}  \tag{5.2}\\
& \left\|\hat{\mathbf{x}}_{t+1}-\hat{\mathbf{x}}_{t^{\prime}+1}\right\|^{2} \leq L_{f}^{2}\left\|\hat{\mathbf{x}}_{t}-\hat{\mathbf{x}}_{t^{\prime}}\right\|^{2}, \forall t, t^{\prime} \in\left\{0, \ldots, T-2 \mid u_{t}=u_{t^{\prime}}\right\}  \tag{5.3}\\
& \hat{\mathbf{x}}_{0}=x_{0} \tag{5.4}
\end{align*}
$$

## Minmax Generalization

- We show that this problem is NP-hard
- Reduction from $\{0,1\}$-programming feasibility problem
- We then decide to look for relaxation schemes of polynomial complexity
- We want these relaxation schemes to preserve the philosophy of the original problem
- Lower bounds


## Relaxation Schemes

- First approach: remove constraints until the problem becomes polynomial

$$
\left(\mathcal{P}^{\prime \prime}\left(\mathcal{F}, u_{0}, \ldots, u_{T-1}\right)\right):
$$

$$
\begin{array}{lll} 
& & \\
\hat{r}_{1} & \sum_{t=1} \hat{\mathbf{r}}_{T-1} \in \mathbb{R} \\
\hat{\mathbf{x}}_{1} & \ldots & \hat{\mathbf{x}}_{T-1} \in \mathcal{X}
\end{array}
$$

Only one
subject to

$$
\begin{align*}
& \left|\hat{\mathbf{r}}_{t}-r^{\left(u_{t}\right), k_{t}}\right|^{2} \leq L_{\rho}^{2}\left\|\hat{\mathbf{x}}_{t}-x^{\left(u_{t}\right), k_{t}}\right\|^{2}, \forall\left(t, k_{t}\right) \in\{1, \ldots, T-1\} \times\left\{1, \ldots, n^{\left(u_{t}\right)}\right\}  \tag{5.1}\\
& \left\|\hat{\mathbf{x}}_{t+1}-y^{\left(u_{t}\right), k_{t}}\right\|^{2} \leq L_{f}^{2}\left\|\hat{\mathbf{x}}_{t}-x^{\left(u_{t}\right), k_{t}}\right\|^{2}, \forall\left(t, k_{t}\right) \in\{0, \ldots, T-1\} \times\left\{1, \ldots, n^{\left(u_{t}\right)}\right\}  \tag{5.2}\\
& \hat{\mathbf{x}}_{0}=x_{0}
\end{align*}
$$

## Relaxation Schemes

- We get the «Intertwined Trust-Region» scheme:

$$
\begin{aligned}
& \hline\left(\mathcal{P}_{I T R}^{\prime \prime}\left(\mathcal{F}, u_{0}, \ldots, u_{T-1}, \bar{k}_{0}, \ldots, \bar{k}_{T-1}\right)\right): \\
& \min \\
& \hat{\mathbf{r}}_{1}, \ldots, \hat{\mathbf{r}}_{T-1} \in \mathbb{R} \\
& \hat{\mathbf{x}}_{0}, \ldots, \hat{\mathbf{x}}_{T-1} \in \mathcal{X}
\end{aligned} \quad \sum_{t=1}^{T-1} \hat{\mathbf{r}}_{t},
$$

subject to

$$
\begin{align*}
\left|\hat{\mathbf{r}}_{t}-r^{\left(u_{t}\right), \bar{k}_{t}}\right|^{2} & \leq L_{\rho}^{2}\left\|\hat{\mathbf{x}}_{t}-x^{\left(u_{t}\right), \bar{k}_{t}}\right\|^{2} & & t \in\{1, \ldots, T-1\}  \tag{5.5}\\
\left\|\hat{\mathbf{x}}_{t}-y^{\left(u_{t-1}\right), \bar{k}_{t-1}}\right\|^{2} & \leq L_{f}^{2}\left\|\hat{\mathbf{x}}_{t-1}-x^{\left(u_{t-1}\right), \bar{k}_{t-1}}\right\|^{2} & & t \in\{1, \ldots, T-1\}  \tag{5.6}\\
\hat{\mathbf{x}}_{0} & =x_{0} & & \tag{5.7}
\end{align*}
$$

## Relaxation Schemes

- This problem can be solved by induction. Define:

$$
\begin{aligned}
& \left(\mathcal{Q}_{I T R}^{\prime \prime}\left(\mathcal{F}, u_{0}, \ldots, u_{j}, \bar{k}_{0}, \ldots, \bar{k}_{j}\right)\right): \\
& \max ^{\hat{\mathbf{r}}_{1}, \ldots, \hat{\mathbf{r}}_{j} \in \mathbb{R}} \begin{array}{l}
\hat{\mathbf{x}}_{0}, \ldots, \hat{\mathbf{x}}_{j} \in \mathcal{X}
\end{array} \quad\left\|\hat{\mathbf{x}}_{j}-x^{\left(u_{j}\right), \bar{k}_{j} \|}\right\|
\end{aligned}
$$

subject to

$$
\begin{array}{rlrl}
\left|\hat{\mathbf{r}}_{t}-r^{\left(u_{t}\right), \bar{k}_{t}}\right|^{2} & \leq L_{\rho}^{2}\left\|\hat{\mathbf{x}}_{t}-x^{\left(u_{i}\right), \bar{k}_{t}}\right\|^{2} & & t \in\{1, \ldots, j\} \\
\left\|\hat{\mathbf{x}}_{t}-y^{\left(u_{t-1}\right), \bar{k}_{t-1}}\right\|^{2} & \leq L_{f}^{2}\left\|\hat{\mathbf{x}}_{t-1}-x^{\left(u_{t-1}\right), \bar{k}_{t-1}}\right\|^{2} & t \in\{1, \ldots, j\} \\
\hat{\mathbf{x}}_{0} & =x_{0} & & \tag{5.10}
\end{array}
$$

## Relaxation Schemes

LEMMA 5.2. The optimal solution $D_{I T R}^{\prime \prime}\left(u_{0}, u_{1}, \bar{k}_{0}, \bar{k}_{1}\right)$ to $\left(\mathcal{Q}_{I T R}^{\prime \prime}\left(\mathcal{F}, u_{0}, u_{1}, \bar{k}_{0}, \bar{k}_{1}\right)\right)$ is given by

$$
D_{I T R}^{\prime \prime}\left(u_{0}, u_{1}, \bar{k}_{0}, \bar{k}_{1}\right)=\left\|\hat{\mathbf{x}}_{1}^{*}\left(\bar{k}_{0}, \bar{k}_{1}\right)-x^{\left(u_{1}\right), \bar{k}_{1}}\right\|,
$$

where

$$
\hat{\mathbf{x}}_{1}^{*}\left(\bar{k}_{0}, \bar{k}_{1}\right) \doteq y^{\left(u_{0}\right), \bar{k}_{0}}+L_{f} \frac{\left\|x_{0}-x^{\left(u_{0}\right), \bar{k}_{0}}\right\|}{\left\|y^{\left(u_{0}\right), \bar{k}_{0}}-x^{\left(u_{1}\right), \bar{k}_{1}}\right\|}\left(y^{\left(u_{0}\right), \bar{k}_{0}}-x^{\left(u_{1}\right), \bar{k}_{1}}\right) \text { if } y^{\left(u_{0}\right), \bar{k}_{0}} \neq x^{\left(u_{1}\right), \bar{k}_{1}}
$$

and, if $y^{\left(u_{0}\right), \bar{k}_{0}}=x^{\left(u_{1}\right), \bar{k}_{1}}, \hat{\mathbf{x}}_{1}^{*}\left(\bar{k}_{0}, \bar{k}_{1}\right)$ can be any point of the sphere centered in $y^{\left(u_{0}\right), \bar{k}_{0}}=$ $x^{\left(u_{1}\right), \bar{k}_{1}}$ with radius $L_{f}\left\|x_{0}-x^{\left(u_{0}\right), \bar{k}_{0}}\right\|$.

# Relaxation Schemes 

$$
+x^{\left(u_{1}\right), \bar{k}_{1}}
$$

$$
\begin{gathered}
\vdots y^{\left(u_{0}\right), \bar{k}_{0}} \begin{array}{c}
\vdots \\
L_{f} \| x_{0}-x^{\left(u_{0}\right), \bar{k}_{0}} \\
\ddots
\end{array}+ \\
\hat{\mathbf{x}}_{1}^{*}\left(\bar{k}_{0}, \bar{k}_{1}\right)
\end{gathered}
$$

A simple geometric algorithm to solve $\left(\mathcal{Q}_{I T R}^{\prime \prime}\left(\mathcal{F}, u_{0}, u_{1}, \bar{k}_{0}, \bar{k}_{1}\right)\right)$

## Relaxation Schemes

LEMMA 5.3. The optimal solution to $\left(\mathcal{Q}_{I T R}^{\prime \prime}\left(\mathcal{F}, u_{0}, \ldots, u_{j}, \bar{k}_{0}, \ldots, \bar{k}_{j}\right)\right)$ is given by:

$$
\begin{aligned}
\forall t \in\{1, \ldots, j\}, \quad & \hat{\mathbf{x}}_{t}^{*}\left(\bar{k}_{0}, \ldots, \bar{k}_{t}\right) \doteq y^{\left(u_{t-1}\right), \bar{k}_{t-1}} \\
& +L_{f} \frac{\left\|\hat{\mathbf{x}}_{t-1}^{*}\left(\bar{k}_{0}, \ldots, \bar{k}_{t-1}\right)-x^{\left(u_{t-1}\right), \bar{k}_{t-1}}\right\|}{\left\|y^{\left(u_{t-1}\right), \bar{k}_{t-1}}-x^{\left(u_{t}\right), \bar{k}_{t}}\right\|}\left(y^{\left(u_{t-1}\right), \bar{k}_{t-1}}-x^{\left(u_{t}\right), \bar{k}_{t}}\right) \\
& \text { if } y^{\left(u_{t-1}\right), \bar{k}_{t-1}} \neq x^{\left(u_{t}\right), \bar{k}_{t}}
\end{aligned}
$$

and, if $y^{\left(u_{t-1}\right), \bar{k}_{t-1}}=x^{\left(u_{t}\right), \bar{k}_{t}}, \hat{\mathbf{x}}_{t}^{*}\left(\bar{k}_{0}, \ldots, \bar{k}_{t}\right)$ can be any point of the sphere centered in $y^{\left(u_{t-1}\right), \bar{k}_{t-1}}=x^{\left(u_{t}\right), \bar{k}_{t}}$ with radius $L_{f}\left\|\hat{\mathbf{x}}_{t-1}^{*}\left(\bar{k}_{0}, \ldots, \bar{k}_{t-1}\right)-x^{\left(u_{t-1}\right), \bar{k}_{t-1}}\right\|$.

## Relaxation Schemes

Theorem 5.4. The solution to $\left(\mathcal{P}_{I T R}^{\prime \prime}\left(\mathcal{F}, u_{0}, \ldots, u_{T-1}, \bar{k}_{0}, \ldots, \bar{k}_{T-1}\right)\right)$ is given by:

$$
B_{I T R}^{\prime \prime}\left(\mathcal{F}, u_{0}, \ldots, u_{T-1}, \bar{k}_{0}, \ldots, \bar{k}_{T-1}\right)=\sum_{t=1}^{T-1} \hat{\mathbf{r}}_{t}^{*}
$$

where

$$
\begin{aligned}
& \hat{\mathbf{r}}_{t}^{*}=r^{\left(u_{t}\right), \bar{k}_{t}}-L_{\rho}\left\|\hat{\mathbf{x}}_{t}^{*}\left(\bar{k}_{0}, \ldots, \bar{k}_{t}\right)-x^{\left(u_{t}\right), \bar{k}_{t}}\right\| \\
& \hat{\mathbf{x}}_{t}^{*}\left(\bar{k}_{0}, \ldots, \bar{k}_{t}\right) \doteq y^{\left(u_{t-1}\right), \bar{k}_{t-1}} \\
& +L_{f} \frac{\left\|\hat{\mathbf{x}}_{t-1}^{*}\left(\bar{k}_{0}, \ldots, \bar{k}_{t-1}\right)-x^{\left(u_{t-1}\right), \bar{k}_{t-1}}\right\|}{\left\|y^{\left(u_{t-1}\right), \bar{k}_{t-1}}-x^{\left(u_{t}\right), \bar{k}_{t}}\right\|}\left(y^{\left(u_{t-1}\right), \bar{k}_{t-1}}-x^{\left(u_{t}\right), \bar{k}_{t}}\right) \\
& \text { if } y^{\left(u_{t-1}\right), \bar{k}_{t-1}} \neq x^{\left(u_{t}\right), \bar{k}_{t}}
\end{aligned}
$$

and, if $y^{\left(u_{t-1}\right), \bar{k}_{t-1}}=x^{\left(u_{t}\right), \bar{k}_{t}}, \hat{\mathbf{x}}_{t}^{*}\left(\bar{k}_{0}, \ldots, \bar{k}_{t}\right)$ can be any point of the sphere centered in $y^{\left(u_{t-1}\right), \bar{k}_{t-1}}=x^{\left(u_{t}\right), \bar{k}_{t}}$ with radius $L_{f}\left\|\hat{\mathbf{x}}_{t-1}^{*}\left(\bar{k}_{0}, \ldots, \bar{k}_{t-1}\right)-x^{\left(u_{t-1}\right), \bar{k}_{t-1}}\right\|$.

## Relaxation Schemes

- One can look for the best bound among all possible configurations

Definition 5.5 (Intertwined Trust-region Bound $B_{I T R}\left(\mathcal{F}, u_{0}, \ldots, u_{T-1}\right)$ ).

$$
\begin{aligned}
& B_{I T R}\left(\mathcal{F}, u_{0}, \ldots, u_{T-1}\right) \triangleq \hat{\mathbf{r}}_{0}^{*} \\
& +\begin{array}{c}
\max \\
\bar{k}_{T-1} \in\left\{1, \ldots, n^{\left(u_{T-1}\right)}\right\} \\
\quad \ldots \\
\quad \bar{k}_{0} \in\left\{1, \ldots, n^{\left(u_{0}\right)}\right\}
\end{array} B_{I T R}^{\prime \prime}\left(\mathcal{F}, u_{0}, \ldots, u_{T-1}, \bar{k}_{0}, \ldots, \bar{k}_{T-1}\right) .
\end{aligned}
$$

## Relaxation Schemes

$$
\left(\mathcal{P}_{L D}^{\prime \prime}\left(\mathcal{F}, u_{0}, \ldots, u_{T-1}\right)\right):
$$

| $\max$ | $\min$ |
| :---: | :---: |
| $\nu_{t, t^{\prime}} \in \mathbb{R}$ | $\hat{\mathbf{r}}_{1}, \ldots, \hat{\mathbf{r}}_{T-1} \in \mathbb{R}$ |
| $\lambda_{t, k_{t}} \in \mathbb{R}$ | $\hat{\mathbf{x}}_{1}, \ldots, \hat{\mathbf{x}}_{T-1} \in \mathcal{X}$ |
| $\mu_{t, k_{t}} \in \mathbb{R}$ |  |

$$
\begin{aligned}
& \hat{\mathbf{r}}_{1}+\cdots+\hat{\mathbf{r}}_{T-1}+ \\
+ & \sum_{\left(t, k_{t}\right) \in\{1, \ldots, T-1\} \times\left\{1, \ldots, n^{\left(u_{t}\right)}\right\}} \mu_{t, k_{t}}\left(\left|\hat{\mathbf{r}}_{t}-r^{\left(u_{t}\right), k_{t}}\right|^{2}-L_{\rho}^{2}\left\|\hat{\mathbf{x}}_{t}-x^{\left(u_{t}\right), k_{t}}\right\|^{2}\right)
\end{aligned}
$$

$$
+\sum_{\left(t, k_{t}\right) \in\{1, \ldots, T-1\} \times\left\{1, \ldots, n^{\left(u_{t}\right)}\right\}} \lambda_{t, k_{t}}\left(\left\|\hat{\mathbf{x}}_{t+1}-y^{\left(u_{t}\right), k_{t}}\right\|^{2}-L_{f}^{2}\left\|\hat{\mathbf{x}}_{t}-x^{\left(u_{t}\right), k_{t}}\right\|^{2}\right)
$$

$$
+\sum_{t, t^{\prime} \in\left\{0, \ldots, T-2 \mid u_{t}=u_{t^{\prime}}\right\}} \nu_{t, t^{\prime}}\left(\left\|\hat{\mathbf{x}}_{t+1}-\hat{\mathbf{x}}_{t^{\prime}+1}\right\|^{2}-L_{f}^{2}\left\|\hat{\mathbf{x}}_{t}-\hat{\mathbf{x}}_{t^{\prime}}\right\|^{2}\right)
$$

## Relaxation Schemes

- The Lagrangian Relaxation provides a lower bound on the optimal bound, in a polynomial time

DEFINITION 5.7 (Lagrandian Bound $\left.B_{L D}\left(\mathcal{F}, u_{0}, \ldots, u_{T-1}\right)\right)$. Let $B_{L D}^{\prime \prime}\left(\mathcal{F}, u_{0}, \ldots, u_{T-1}\right)$ be the optimal Lagrangian dual of $\left(\mathcal{P}_{L D}^{\prime \prime}\left(\mathcal{F}, u_{0}, \ldots, u_{T-1}\right)\right)$. Then,

$$
B_{L D}\left(\mathcal{F}, u_{0}, \ldots, u_{T-1}\right)=\mathbf{r}_{0}^{*}+B_{L D}^{\prime \prime}\left(\mathcal{F}, u_{0}, \ldots, u_{T-1}\right)
$$

## Bounds Tightness \& Convergence

- Tightness

Theorem 5.18. $\forall\left(u_{0}, \ldots, u_{T-1}\right) \in \mathcal{U}^{T}$,

$$
\begin{aligned}
B_{C G R L}\left(\mathcal{F}, u_{0}, \ldots, u_{T-1}\right) & \leq B_{I T R}\left(\mathcal{F}, u_{0}, \ldots, u_{T-1}\right) \\
& \leq B_{L D}\left(\mathcal{F}, u_{0}, \ldots, u_{T-1}\right) \\
& \leq B^{*}\left(\mathcal{F}, u_{0}, \ldots, u_{T-1}\right) \\
& \leq J\left(u_{0}, \ldots, u_{T-1}\right)
\end{aligned}
$$

- Convergence as the sample dispersion of the sample of trajectories goes to 0

THEOREM 5.21. $\forall\left(u_{0}, \ldots, u_{T-1}\right) \in \mathcal{U}^{T}$, $\forall \beta \in\left\{B_{C G R L}\left(\mathcal{F}, u_{0}, \ldots, u_{T-1}\right), B_{I T R}\left(\mathcal{F}, u_{0}, \ldots, u_{T-1}\right), B_{L D}\left(\mathcal{F}, u_{0}, \ldots, u_{T-1}\right)\right\}$,

$$
\lim _{\alpha^{*}(\mathcal{F}) \rightarrow 0} \quad J\left(u_{0}, \ldots, u_{T-1}\right)-\beta=0
$$

sixk

## Trajectories of Societies



# From Bad Models to Good Policies 

The Energy Transition Case

## 1. World primary energy consumption

Non renewable
$>80 \%-<20 \%$

Renewable

## 2. Energy <-> Economy

- Recent research in Economics has shown that:
- The empirical elasticity (measured from time series among OECD countries over the last 50 years) of the consumption of primary energy into the GDP is about 60\%, which is 10 times higher that what is predicted by the Cost Share Theorem

Elasticity can be quantified as the ratio of the percentage change in one variable to the percentage change in another variable

- There is a causality link between the consumption of primary energy and the GDP in the direction Energy -> GDP



Variation of the world oil consumption (red) and GDP per inhabitant (blue) - Data from the the World Bank for GDP and BP stat for energy

Source (in French): Jean-Marc Jancovici, « L'économie aurait-elle un vague rapport avec l'énergie? », LH Forum, 27 septembre 2013

## 3. ERoEI

- ERoEl for « Energy Return over Energy Investment » (also called EROI) is the ratio of the amount of usable energy acquired from a particular energy resource to the amount of energy expended to obtain that energy resource:

$$
E R O I=\frac{\text { Usable Acquired Energy }}{\text { Energy Expended }}
$$

- The highest this ratio, the more energy a technology brings back to society
- Notation : 1:X

$\mathrm{EROI}_{\text {st }}$
$\square$ Energy Used to Procure Energy
Source: EROI of Global Energy Resources - Preliminary Status and Trends - Jessica Lambert, Charles Hall, Steve Balogh, Alex Poisson, and Ajay Gupta State University of New York, College of Environmental Science and Forestry Report 1 - Revised Submitted - 2 November 2012 DFID - 59717


## Another Bad Model

- A discrete-time model of the deployment of «renewable energy » production capacities
- Budget of non-renewable energy

$$
\begin{aligned}
& \forall t \in\{0, \ldots, T-1\}, B_{t} \geq 0 \\
& \exists r>0, \exists \tau>0, \exists t_{0} \in \mathbb{R}: \forall t \in\{0, \ldots, T-1\}, \\
& B_{t}=\frac{1}{r} \frac{e^{\frac{-\left(t-t_{0}\right)}{\tau}}}{\left(1+e^{\frac{-\left(t-t_{0}\right)}{\tau}}\right)^{2}}
\end{aligned}
$$

## Another Bad Model

- Set of renewable energy production technologies:

$$
\forall n \in\{1, \ldots, N\}, \forall t \in\{0, \ldots, T-1\}, R_{n, t} \geq 0
$$

- Characteristics

$$
\Delta_{n, t} \geq 0
$$

$$
E R o E I_{n, t} \geq 0
$$

- Deployment strategy
$R_{n, t+1}=\left(1+\alpha_{n, t}\right) R_{n, t} \quad \alpha_{n, t} \in[-1, \infty[$


## Another Bad Model

- Energy costs for growth and long-term replacement $\forall n \in\{1, \ldots, N\}, \forall t \in\{0, \ldots, T-1\}$,

$$
C_{n, t}\left(R_{n, t}, \alpha_{n, t}\right) \geq 0 \quad M_{n, t} \geq 0
$$

- Total energy and net energy to society

$$
\begin{array}{r}
\forall t \in\{0, \ldots, T-1\}, E_{t}=B_{t}+\sum_{n=1}^{N} R_{n, t} \\
S_{t}=E_{t}-\left(\sum_{n=1}^{N} C_{n, t}\left(R_{n, t}, \alpha_{n, t}\right)+M_{n, t}\right)
\end{array}
$$

## Another Bad Model

- Constraint on the quantity of energy invested for energy production

$$
\begin{aligned}
\forall t \in & \{0, \ldots, T-1\} \\
& \exists \sigma_{t}: C_{n, t}\left(R_{n, t}, \alpha_{n, t}\right)+M_{n, t} \leq \frac{1}{\sigma_{t}} E_{t}
\end{aligned}
$$

## Another Bad Model

- Further assumptions
- Energy cost for growth is proportional to growth, and done initially:

$$
C_{n, t}\left(R_{n, t}, \alpha_{n, t}\right)=\frac{\Delta_{n, t}}{E R o E I_{n, t}} \alpha_{n, t} R_{n, t} \text { if } \alpha_{n, t} \geq 0
$$

- Long-term replacement cost is (i) proportional and (ii) annualized

$$
M_{n, t}\left(R_{n, t}\right)=\frac{1}{E R o E I_{n, t}} R_{n, t}
$$

$$
\begin{aligned}
& E_{0}=1 \\
& B_{0}=0.85 E_{0} \\
& R_{1,0}=0.01 E_{0}
\end{aligned}
$$

$$
\sum_{n=2}^{N} R_{n, 0}=0.14 E_{0}
$$

$E R o E I_{1, t}=9$
$\Delta_{1, t}=20$
$\sigma_{t}=14$
Constant growth if possible, else max admissible


Fig. 2. Scenario "peak at time $t=0$ "


Fig. 4. Scenario "peak at time $t=20$ "


Fig. 3. Scenario "plateau at time $t=0$ "


Fig. 5. Scenario "plateau at time $t=20$ "

## Another Bad Model

- Increasing the ERoEl parameter

$\forall t \in\{0, \ldots, T-1\}, E R o E I_{1, t}=9+\frac{t}{T}(12-9)$


## Good Policies?

- What kind of « good policy » can be suggested by such a « bad model»?
- Energy efficiency: « do better with less »
-> Lots of decision making under uncertainty problems to solve here
- For people interested in Smart Grids: below is link toward a simulator for Active Network Management (ANM) developed by my colleagues at the University of Liège:
http://www.montefiore.ulg.ac.be/~anm/


## Epilogue



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