

Dynamic Optimization of Solar Thermal Energy Systems with Storage

Kody Powell Dr. John Hedengren Dr. Thomas Edgar

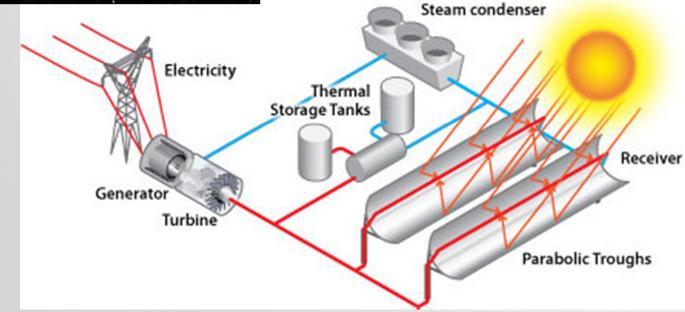
TWCCC: September 2011



Overview of Solar Thermal Power



- Promising technology
- Systems approach insight into plant design



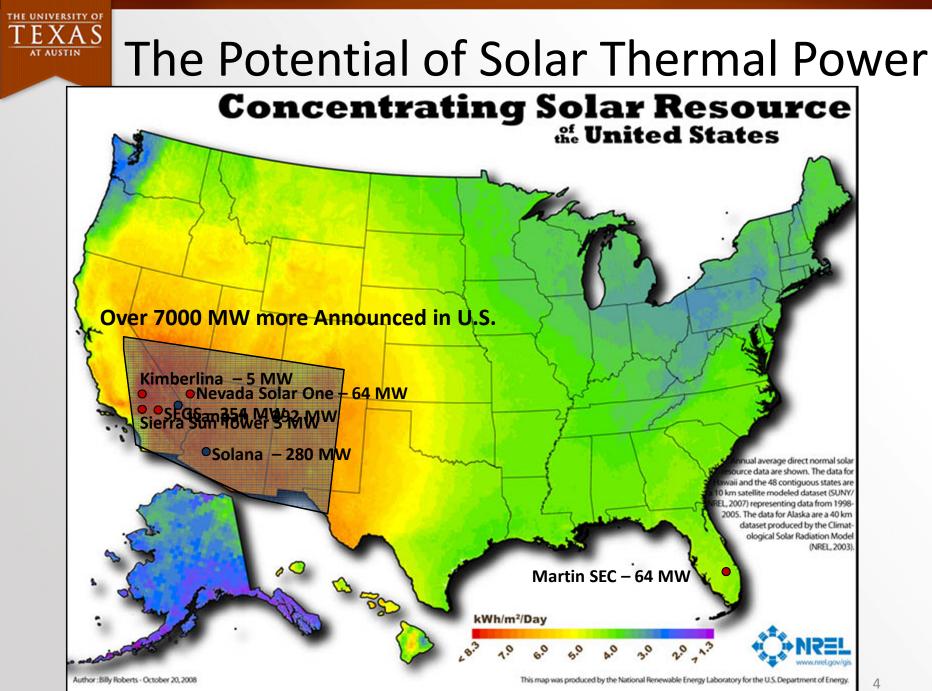
Solar Thermal vs Photovoltaic (PV)

	Solar Thermal	Photovoltaic
Energy Conversion	Sunlight \rightarrow Heat \rightarrow Mechanical \rightarrow Electricity	Sunlight \rightarrow Electricity
Cost (\$/kWh)	0.12 ¹ (0.06 Projected) ²	0.18-0.23 ¹
Efficiency ³	~18%	~12%
Solar Irradiance Used	Direct Normal Irradiance (DNI)	Global Horizontal (GHI)
Scale	Large Scale	Large Scale & Distributed
Storage	Thermal Storage	Battery Storage
Dispatchable on a large scale	Yes	No
Impact on grid	Small	Large

¹ http://www.window.state.tx.us/specialrpt/energy/exec/solar.html

² http://www.reuters.com/article/2009/08/24/us-energy-maghreb-desertec-sb-idUSTRE57N01720090824?sp=true

³ http://solarbuzz.com/facts-and-figures/markets-growth/cost-competitiveness





Can Forecasts Help Solar Thermal?

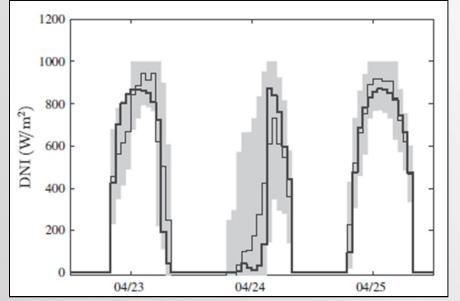
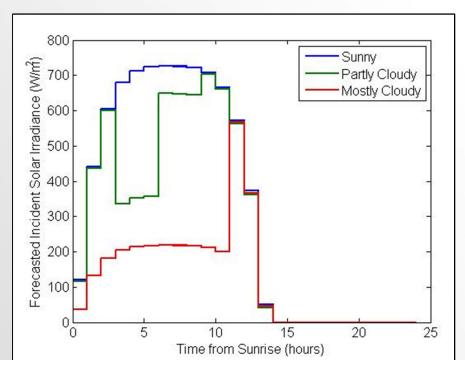


Image from: Marquez, R. and Coimbra, C. F. M., Forecasting of global and direct solar irradiance using stochastic learning methods, ground experiments and the NWS database, *Solar Energy*, Volume 85, 2011

- Look at extreme scenarios
- Compare Standard Control Approach to Dynamic Optimization

- Forecasting technology advancing
- How do we take advantage?



Constant Temperature and Constant Power Control

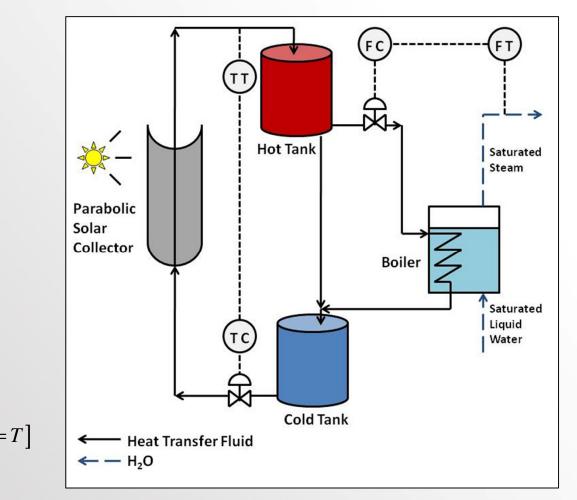
- NMPC approach
- Measure DNI, use as FF to plant
- Constant temperature, constant power
- Relief pipe used when hot tank fills

$$\min_{u(t)} \int_{t=0}^{t=T} \phi \left(\mathbf{x}(\tau), \mathbf{y}(\tau), \mathbf{u}(\tau), \mathbf{d}(\tau) \right) d\tau$$

$$\dot{\mathbf{x}} = \mathbf{f} \left(\mathbf{x}(\tau), \mathbf{y}(\tau), \mathbf{u}(\tau), \mathbf{d}(\tau) \right)$$

$$\mathbf{0} = \mathbf{g} \left(\mathbf{x}(\tau), \mathbf{y}(\tau), \mathbf{u}(\tau), \mathbf{d}(\tau) \right)$$

$$\mathbf{0} \ge \mathbf{h} \left(\mathbf{x}(\tau), \mathbf{y}(\tau), \mathbf{u}(\tau), \mathbf{d}(\tau) \right)$$

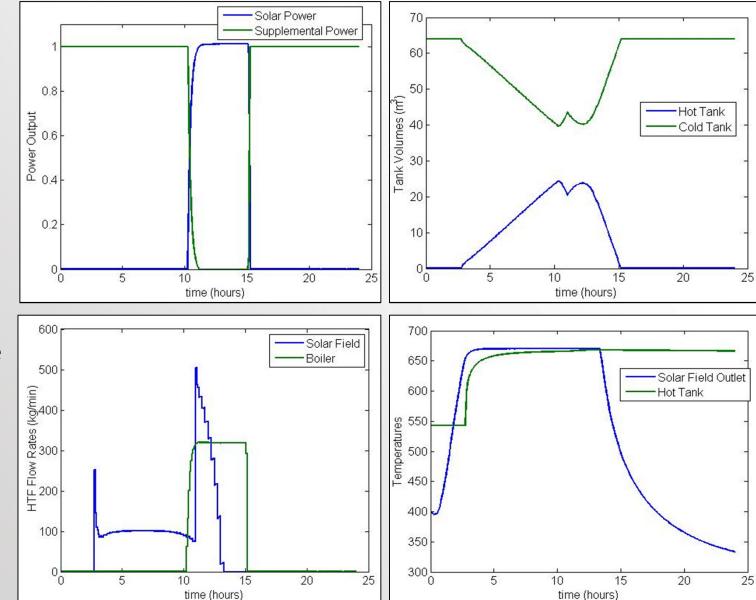


Constant T Constant Po Approach

Performance Improved by: •Optimal temperature •Hybrid Operation •Storage Bypass

Further Improvements Needed:

Strategy for more stable operation
Consider stochastic problem

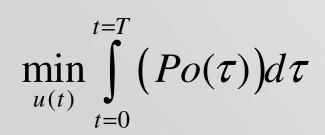


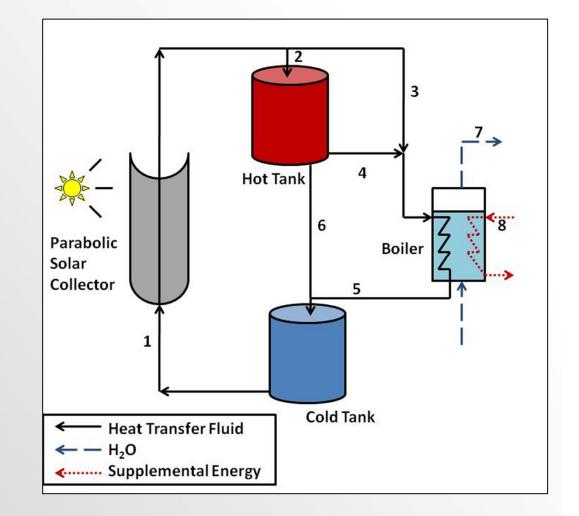


Dynamic Optimization w/ Forecasts

Hypothesis:

- Performance can be improved by:
 - Controlling to optimal temperatures
 - Hybrid operation
 - Ability to bypass storage
 - More DOFs







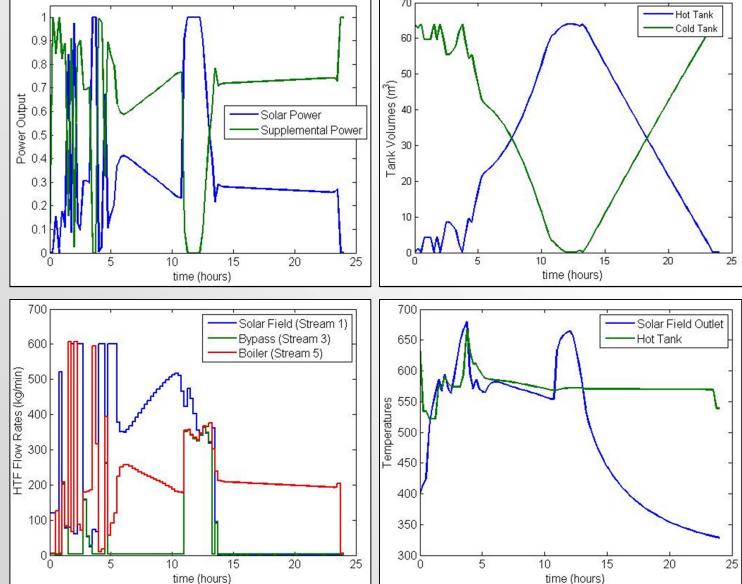
Performance

Improved by:
Optimal
temperature
Hybrid Operation
Storage Bypass

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Dynamic Optimization w/ Forecast





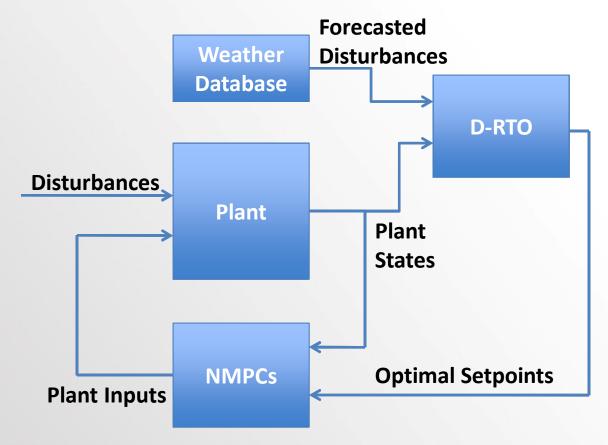
Dynamic Optimization Improves Performance

	Solar Energy Collected (MWh)	Energy Collected/ Total Incident Energy (%)
Sunny Day		
Standard Control	18.02	76.8%
Dynamic Optimization	18.59	79.2%
Partly Cloudy Day		
Standard Control	14.60	75.8%
Dynamic Optimization	15.83	81.1%
Mostly Cloudy Day		
Standard Control	4.75	52.1%
Dynamic Optimization	7.80	85.4%



Proposed D-RTO Formulation

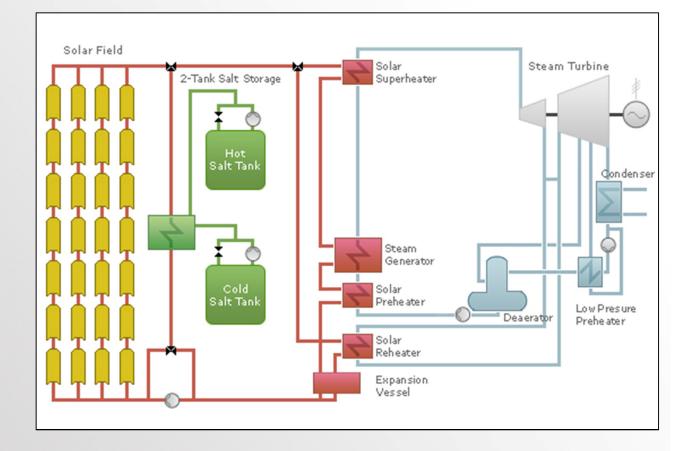
- Supervisory Control
- NMPC for stability (run every 2-5 minutes)
- D-RTO runs every 1-2 hours
 - Fewer variables may help solver find global min
- Forecast and plant states updated regularly

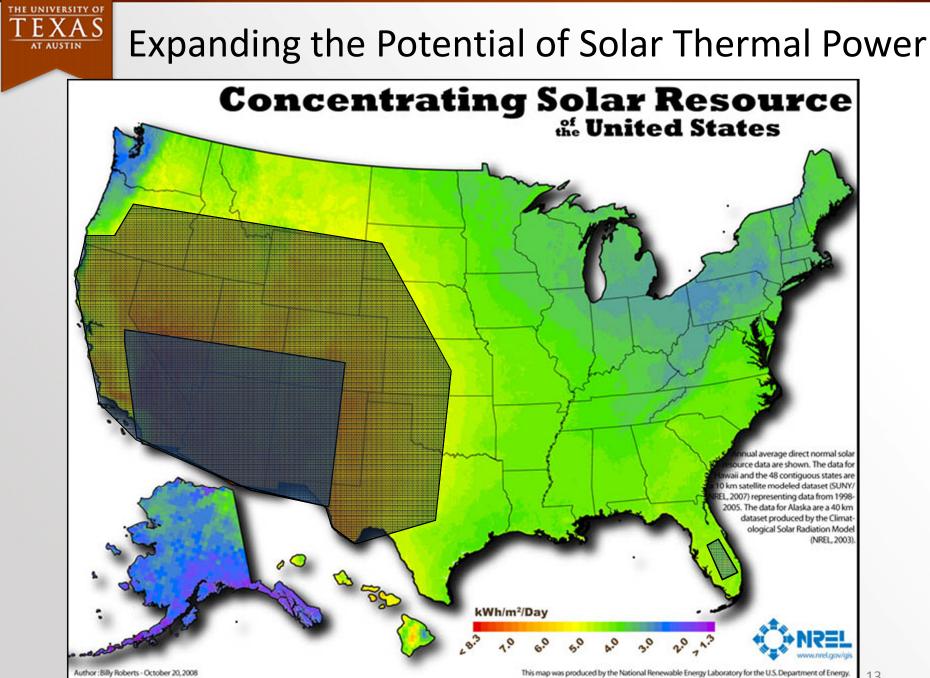




Opportunities for Further Improvement

- More DOFs → more optimal solutions
- Consider entire plant
- Include electric grid and demand forecasts
- Economic optimization
- Apply to other systems





Conclusions

Better utilization of renewable resources

Particularly on cloudy days

- Systems approach leads to design insights
- Hybridization can greatly expand solar thermal utilization
- Supervisory D-RTO needed
- Thanks to APMonitor, National Science Foundation, Cockrell School of Engineering, UT-Austin, Edgar Group