

PROBABILISTIC FORECAST OF REAL-TIME LOCATIONAL MARGINAL PRICE

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Acknowledgement:

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Overview

□ Objectives

- Investigate the **real-time LMP forecast** problem by an operator
- Develop **probabilistic forecasting techniques**.
- Incorporate **real-time measurements** and general load forecast models.
- Develop **scalable computation** techniques.

□ Summary of results

- Real-time LMP models for energy only and energy-reserve markets.
- Forecast methods:
 - A multiparametric linear program (MPLP) approach for ex ante LMP
 - A Markov chain approach for ex ante and ex post LMPs
- Simulations.

Operator provided real-time LMP forecast

□ Benefits

- Valuable for generation and demand response decisions
- Risk management
- Congestion relief and operating cost reduction
- Examples: ERCOT, AEMO,

□ Advantages of LMP forecast by an operator

- Access to real-time data and network operating conditions
- Access to internal LMP computation state
- Ability to incorporate load/variable generation forecast models.

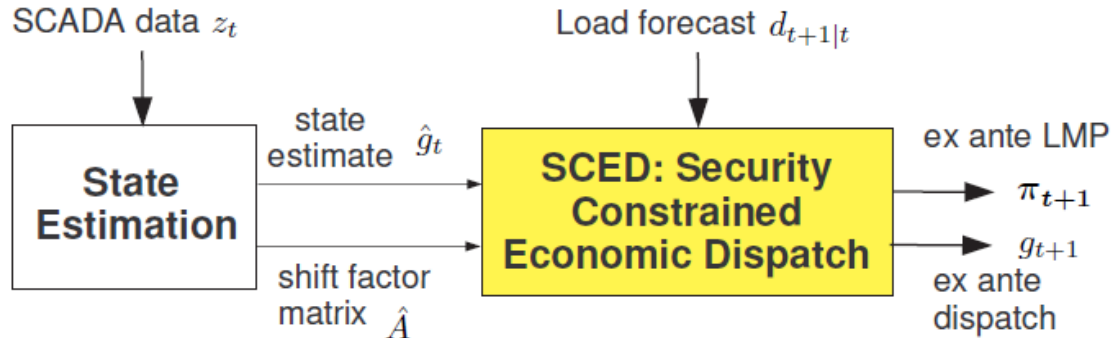
The state-of-the-art

- Extensive literature on load forecast
 - Quite accurate and robust performance (1-3%)
 - Both point and probabilistic forecast techniques exist
 - Renewable integration brings new challenges
- Considerable literature on day ahead LMP forecast
 - Mostly **black box** techniques
 - Relatively poor performance (10-20% MAPE)
 - Primarily providing **point forecasting**
 - Rarely deal with on **LMP and network effects**
- Very little published work on probabilistic LMP forecast

Outline

- Introduction
- Real-time LMP models (energy & reserve)
 - Ex ante LMP model
 - Ex post LMP model
- Probabilistic Real-time LMP forecast
 - Probabilistic forecast & performance measure
 - A multiparametric linear program (MPLP) approach
 - A Markov chain approach to ex post LMP forecast
- Simulation results

An ex ante real time LMP model



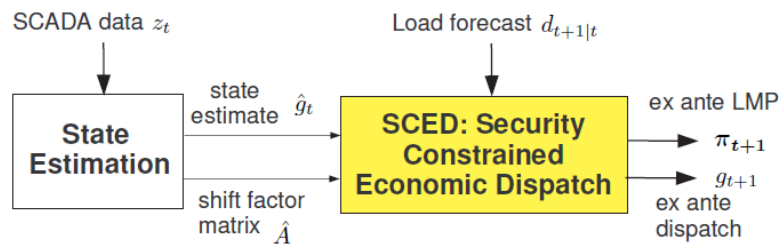
minimize $c^T g$

subject to

$$\begin{aligned}
 (\lambda) : \quad & \mathbf{1}^T (g - d_{t+1|t}) = 0, && \text{power balance,} \\
 (\mu^{+/-}) : \quad & -F^+ \leq \hat{A}(g - d_{t+1|t}) \leq F^+, && \text{transmission constraint,} \\
 (\gamma^{+/-}) : \quad & g^- \leq g \leq g^+, && \text{generation capacity,} \\
 (\eta^{+/-}) : \quad & \hat{g}_t - \Delta^- \leq g \leq \hat{g}_t + \Delta^+, && \text{ramp limit.}
 \end{aligned}$$

$$\pi_{t+1} = \lambda \mathbf{1} + \hat{A} \mu^+ - \hat{A} \mu^-.$$

An ex ante real time LMP model with co-optimization



P0:

subject to:

$(\lambda) :$

$(\mu^{+/-}) :$

$(\alpha) :$

$(\beta) :$

$(\gamma^+) :$

$(\nu) :$

$(\eta) :$

$(\gamma^-) :$

$$\min_{g,r,s} c_g^T g + c_r^T r + c_p^T s_l + c_p^T s_s$$

$$1^T (g - d_{t+1|t}) = 0,$$

$$-F^+ \leq \hat{A}(g - d_{t+1|t}) \leq F^+,$$

$$\delta_{\text{local}}^T r + (I^+ - I) + s_l \geq Q_l,$$

$$I = \hat{A}_I (g - d_{t+1|t}),$$

$$\delta_{\text{system}}^T r + s_s \geq Q_s,$$

$$g + r \leq g^+,$$

$$\hat{g}_t - \Delta^- \leq g \leq g_t + \Delta^+,$$

$$0 \leq r_i \leq r^+,$$

$$g \geq g^-,$$

$$s_l, s_s \geq 0.$$

energy balance,

transmission,

local reserve,

interface flow,

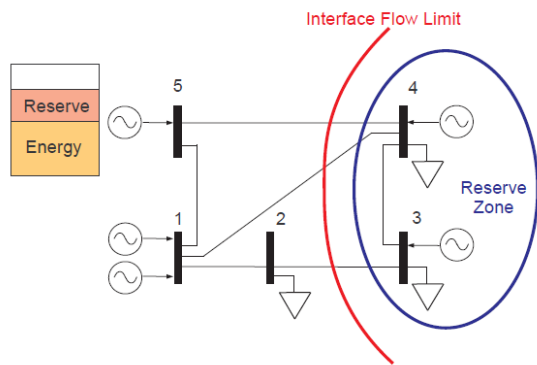
system reserve,

generator capacity,

generation ramp,

reserve ramp,

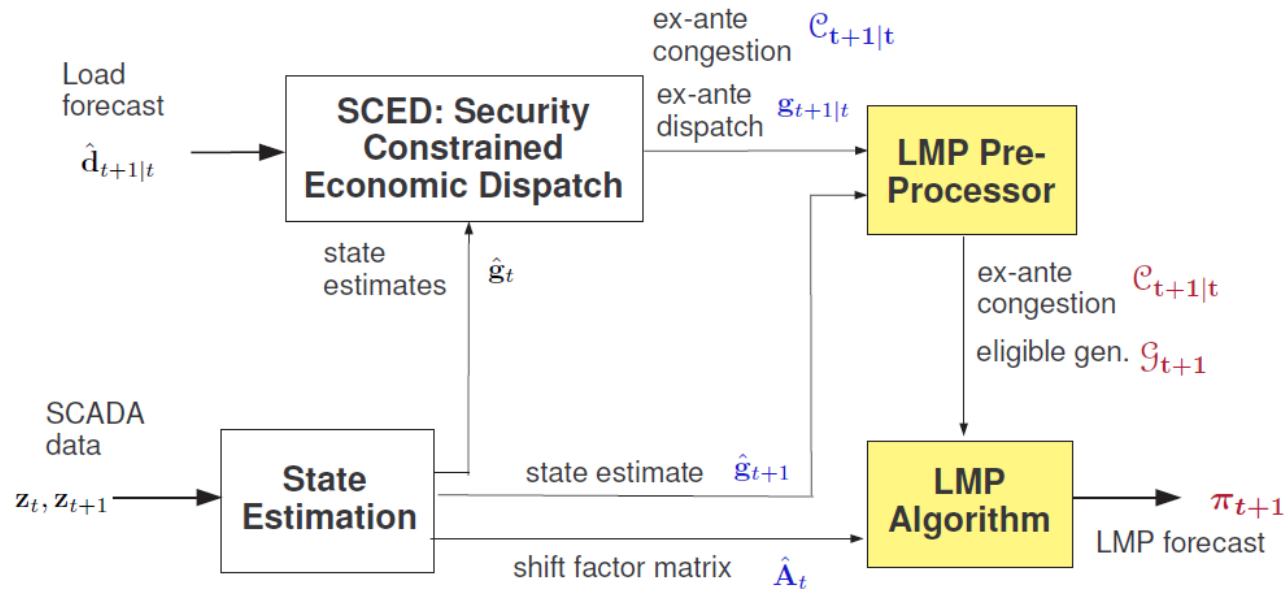
generation capacity,



$$\pi_{t+1} = \lambda \mathbf{1} + \hat{A} \mu^+ - \hat{A} \mu^- + \hat{A}_I \alpha$$



An ex post real time LMP model



$$\begin{pmatrix} z_{t+1} \\ z_t \\ \hat{d}_{t+1|t} \end{pmatrix} \xrightarrow{\text{WLS-SE}} \begin{pmatrix} \hat{g}_{t+1} \\ \hat{g}_t \\ \hat{d}_{t+1|t} \end{pmatrix} \xrightarrow{\text{SCED}} \begin{pmatrix} \hat{g}_{t+1} \\ g_{t+1|t} \\ c_{t+1|t} \end{pmatrix} \xrightarrow{\text{LMP Prep}} \begin{pmatrix} g_{t+1} \\ c_{t+1|t} \end{pmatrix} \xrightarrow{\text{IDCOPE}} \pi_{t+1}$$

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 - ▣ Ex ante LMP model
 - ▣ Ex post LMP model
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Point, categorical, and probabilistic forecasts

□ Point forecast:

- Predict the specific value of a random variable to be realized in the future
- Performance measure:
 - mean squared error (MSE), mean absolute error (MAE), and mean percentage error (MAPE).
- Often leads to parametric approach. Kalman filtering, time series, NN, etc.

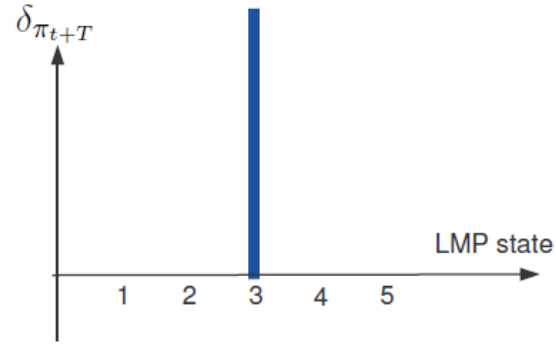
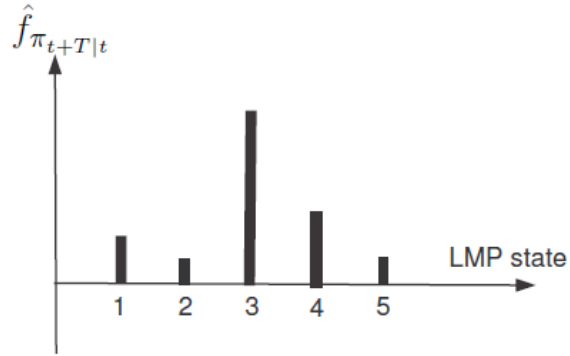
□ Categorical forecast:

- Predict the (discrete) category of random phenomenon (e.g. prediction interval)
- Performance measure: Categorical score
- Very difficult for high dimensional models

□ Probabilistic forecast:

- Predict the **distribution** of a random variable at a future time.
- Performance measure: (Brier score, KL divergence, total variation...)
- Very difficult to obtain structured solution in general
- **Bayesian Monte Carlo techniques + problem specific properties.**

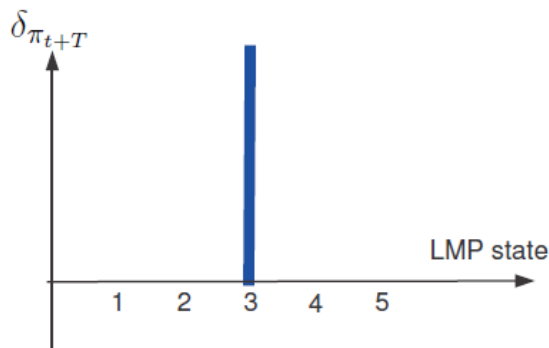
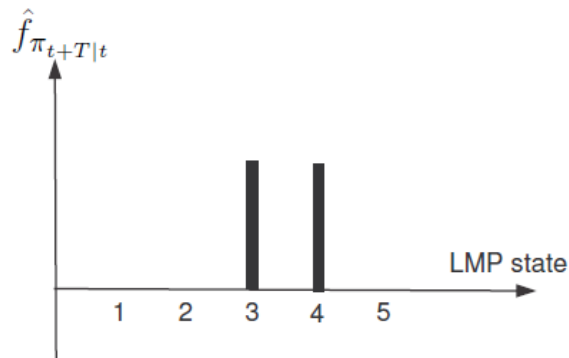
Distribution forecast



- The **Brier score** is defined by

$$\mathcal{B}(\hat{f}) = \mathbb{E} \left(||\hat{f}_{\pi_{t+T}|t} - \delta_{\pi_{t+T}}||^2 \right)$$

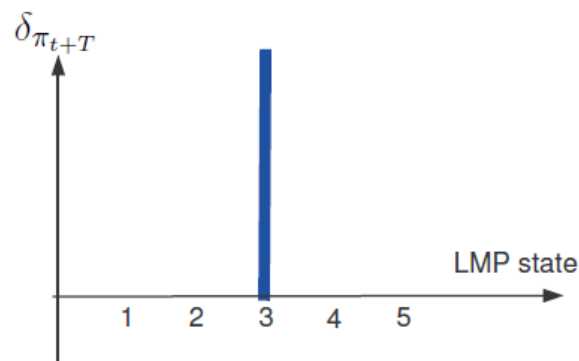
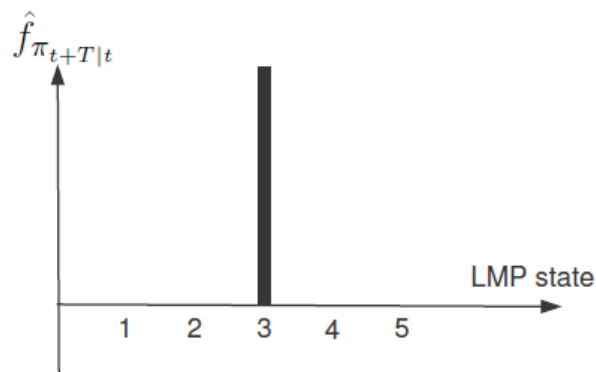
Categorical forecast



- The performance of a categorical forecast is measured by the tradeoff between **sharpness** and the **accuracy** of the forecaster.
- The **categorical score** of $\hat{P}_{t+T|t}$ is defined by

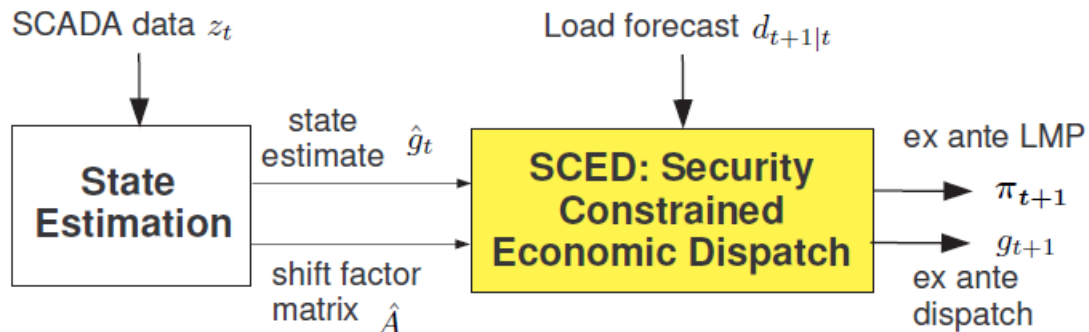
$$\mathcal{C}(\hat{P}_{t+T|t}) \triangleq \frac{|\hat{\mathcal{P}}_{t+T|t}|}{|\Omega|} + \Pr \left[\pi_{t+T} \notin \hat{\mathcal{P}}_{t+T|t} \right] \left(1 - \frac{|\hat{\mathcal{P}}_{t+T|t}|}{|\Omega|} \right)$$

Point forecast



- Point forecaster is a categorical forecaster of size 1.
- The performance of a point forecaster can be measured by
 - ▶ probability of error: $\mathcal{E}_p = \Pr(\hat{\pi}_{t+T|t} \neq \pi_{t+T})$.
 - ▶ mean squared error (MSE): $\mathcal{E}_{\text{MSE}} = \mathbb{E}(\|\hat{\pi}_{t+T|t} - \pi_{t+T}\|_2^2)$.
 - ▶ mean absolute error (MAE): $\mathcal{E}_{\text{MAE}} = \mathbb{E}(\|\hat{\pi}_{t+T|t} - \pi_{t+T}\|_1)$.
 - ▶ mean absolute percentage error (MAPE): $\mathcal{E}_{\text{MAPE}} = \frac{\mathbb{E}(\|\hat{\pi}_{t+T|t} - \pi_{t+T}\|_1)}{\mathbb{E}(\|\pi_{t+T}\|_1)}$.

MPLP forecast



minimize $c^T g$

subject to

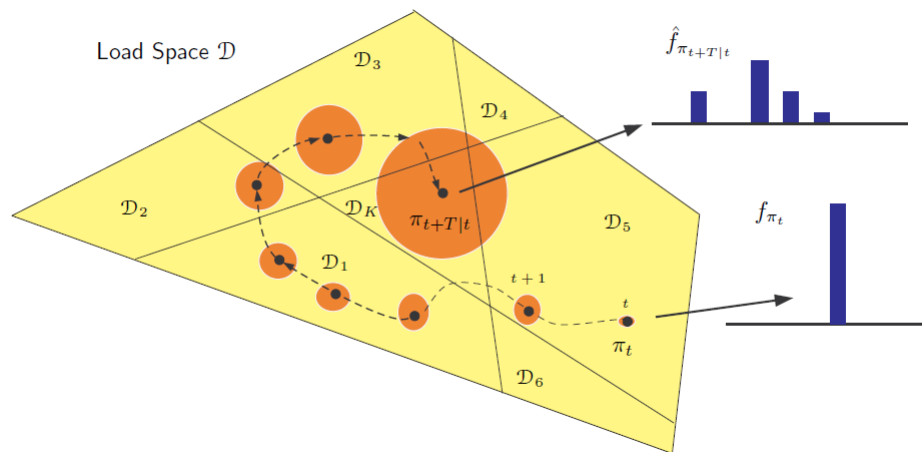
$$(\lambda) : \quad \mathbf{1}^T (g - d_{t+1}) = 0,$$

$$(\mu^{+/-}) : \quad -F^+ \leq \hat{A}(g - d_{t+1}) \leq F^+,$$

$$(\gamma^{+/-}) : \quad g^- \leq g \leq g^+,$$

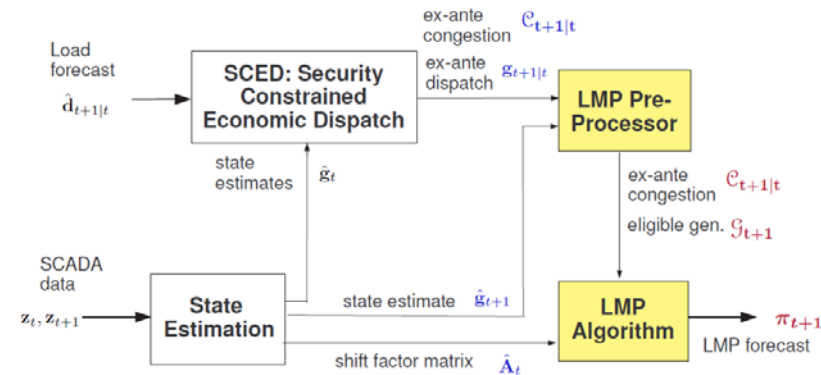
$$(\eta^{+/-}) : \quad B d_t - \Delta^- \leq g \leq B d_t + \Delta^+,$$

$$\pi_{t+1} = \lambda \mathbf{1} + \hat{A} \mu^+ - \hat{A} \mu^-.$$

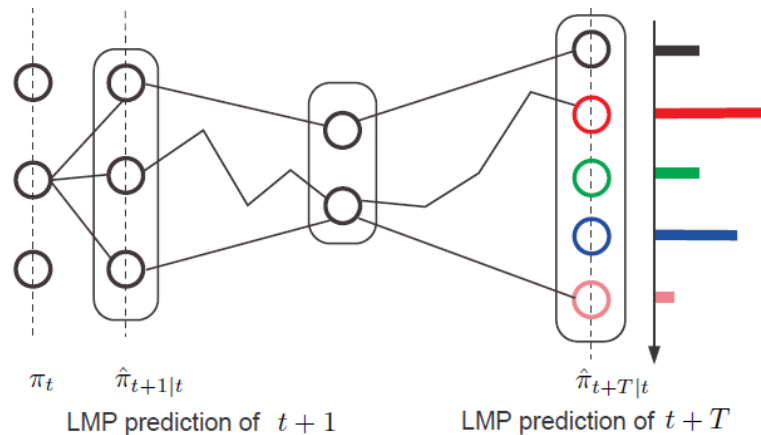
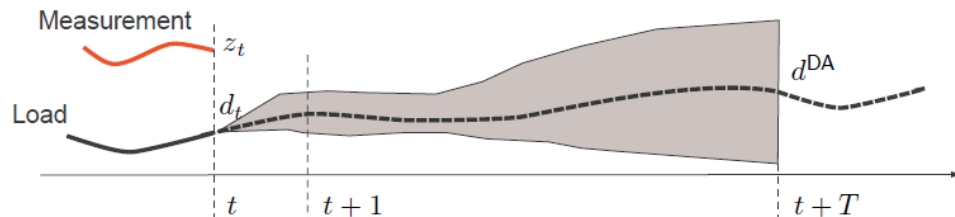




Forecast of ex post LMP via Markov chain



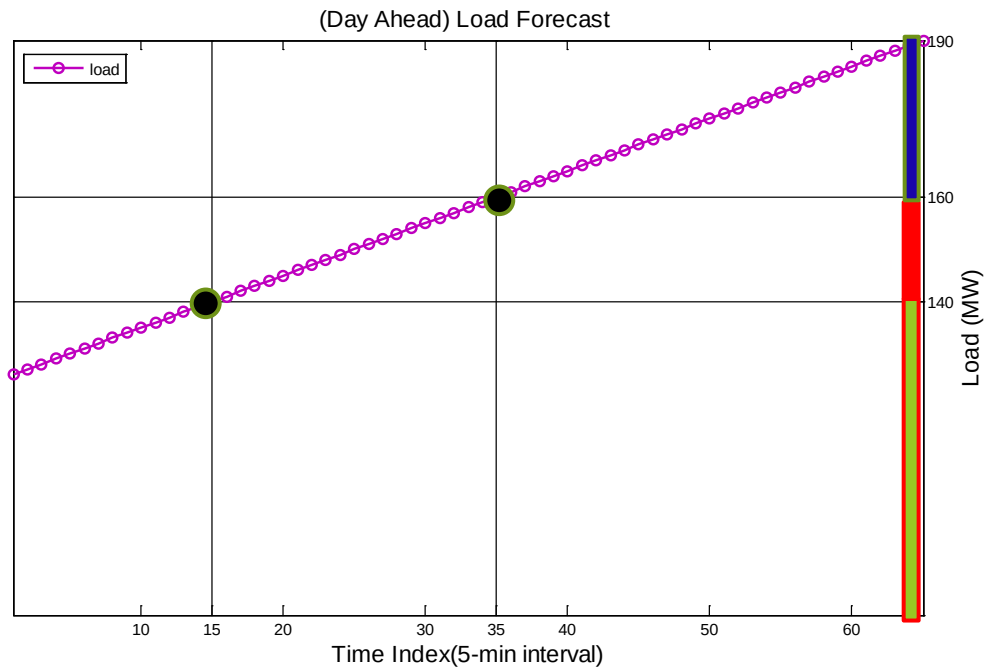
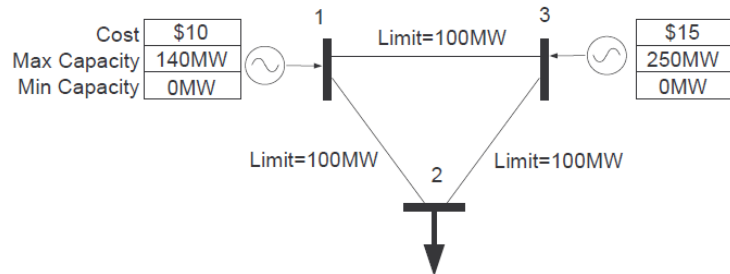
$$\begin{pmatrix} g_{t+1} \\ c_{t+1|t} \end{pmatrix} \xrightarrow{\text{LMPA}} \pi_{t+1} = \begin{bmatrix} \pi_{t+1}^{(1)} \\ \vdots \\ \pi_{t+1}^{(n)} \end{bmatrix}$$



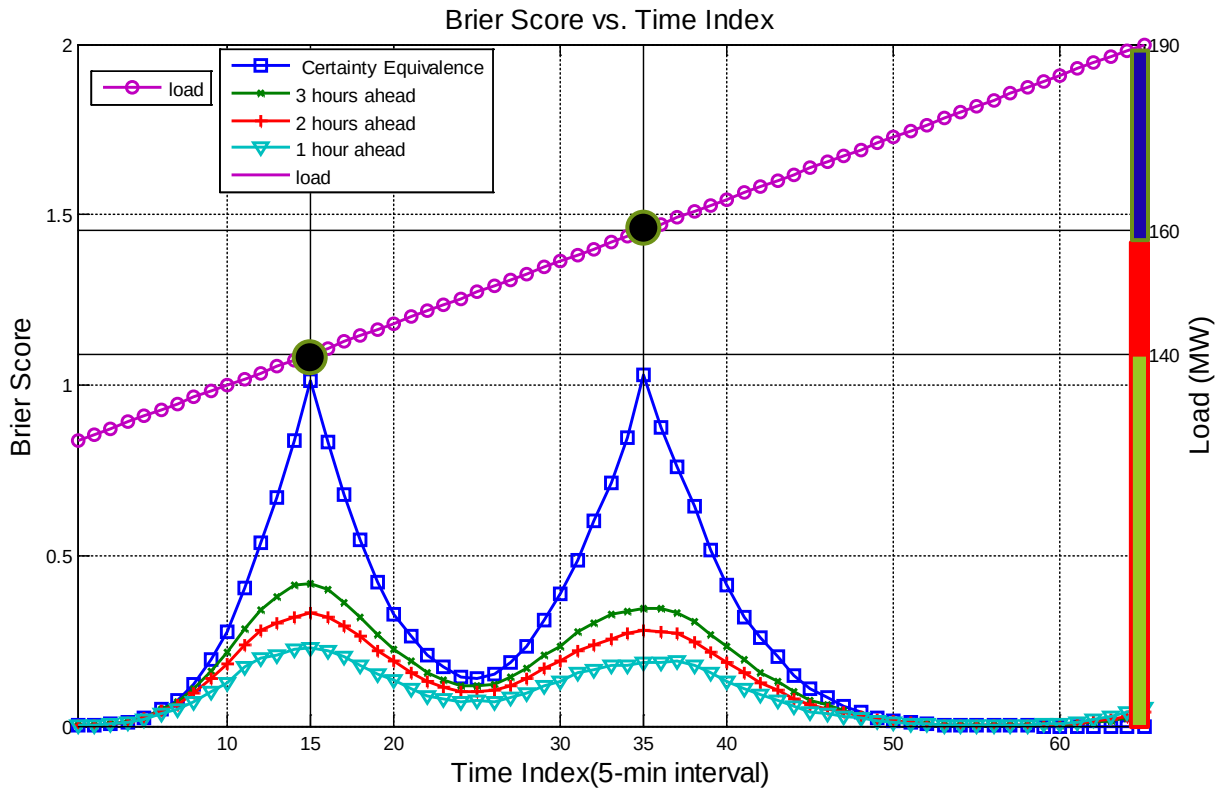
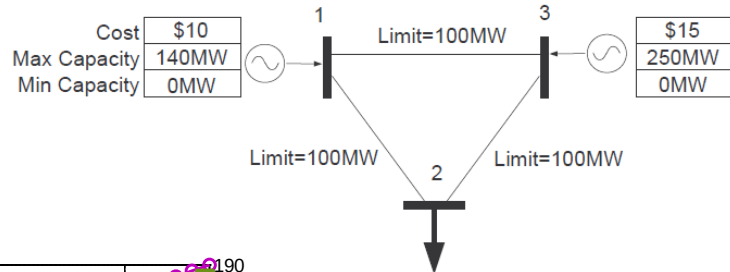
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1 D (mean) load trajectory

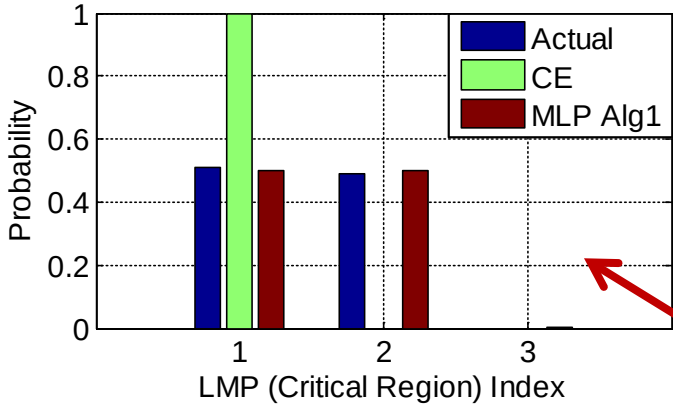


LMP forecast with 1D load

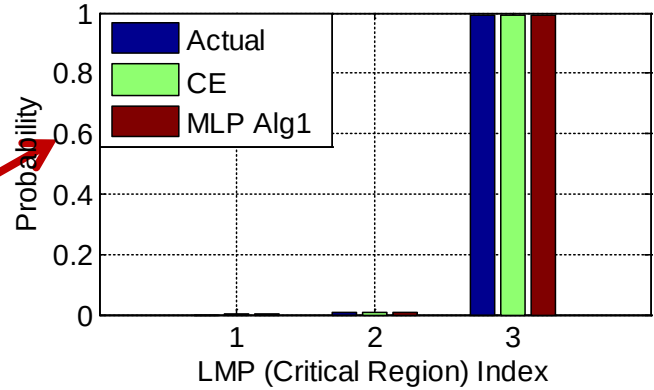


LMP forecast with 1D load

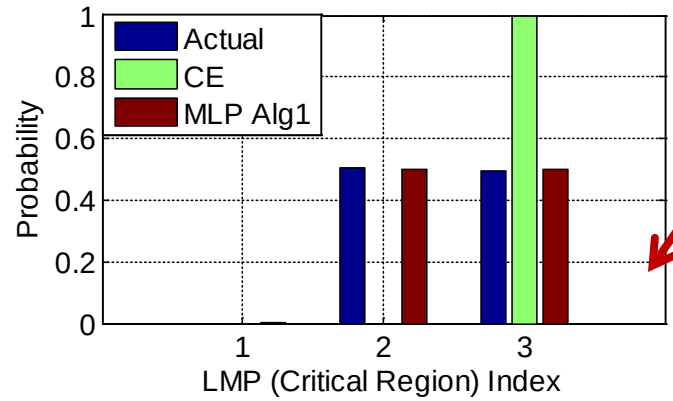
3hrs ahead Forecast at Time 15 (TV=0.0078)



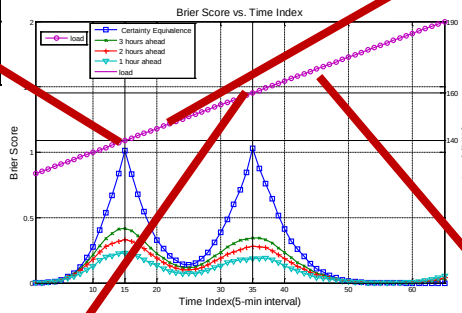
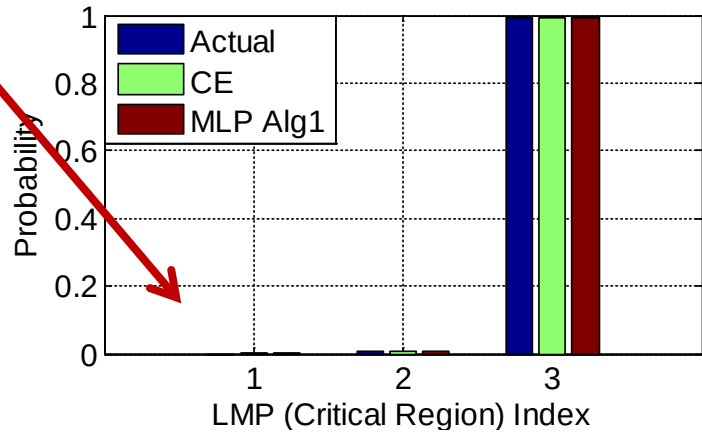
3hrs ahead Forecast at Time 25 (TV=0.0019)



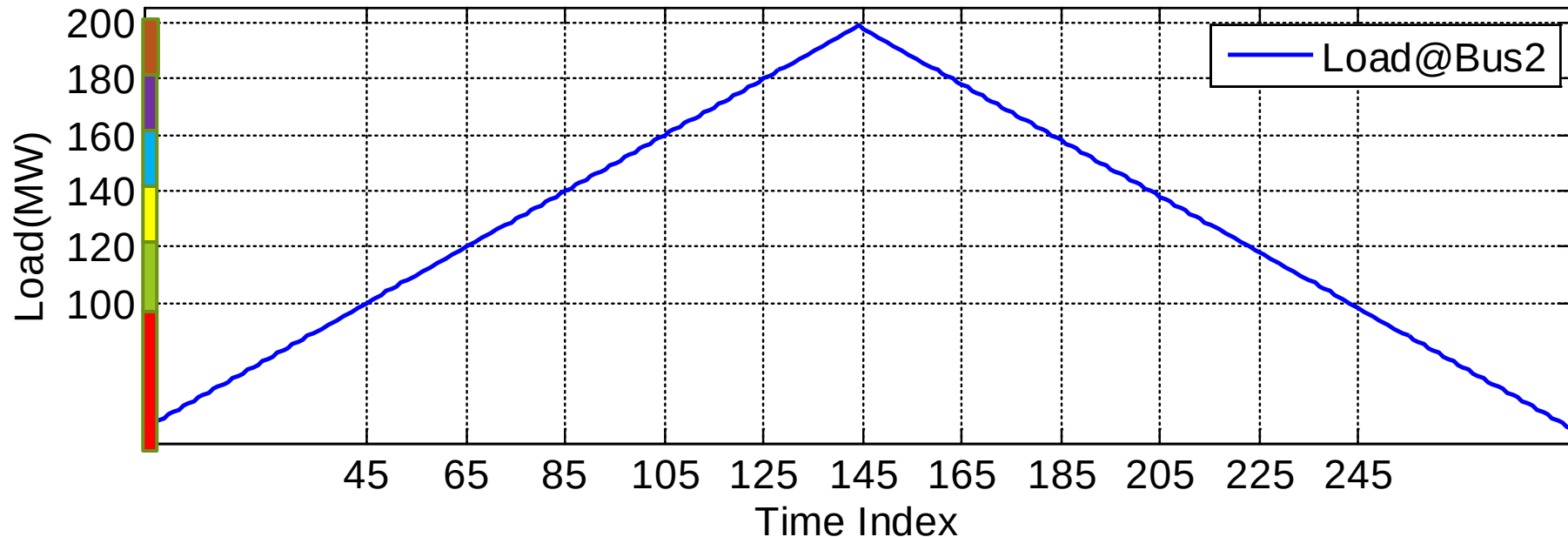
3hrs ahead Forecast at Time 35 (TV=0.007)



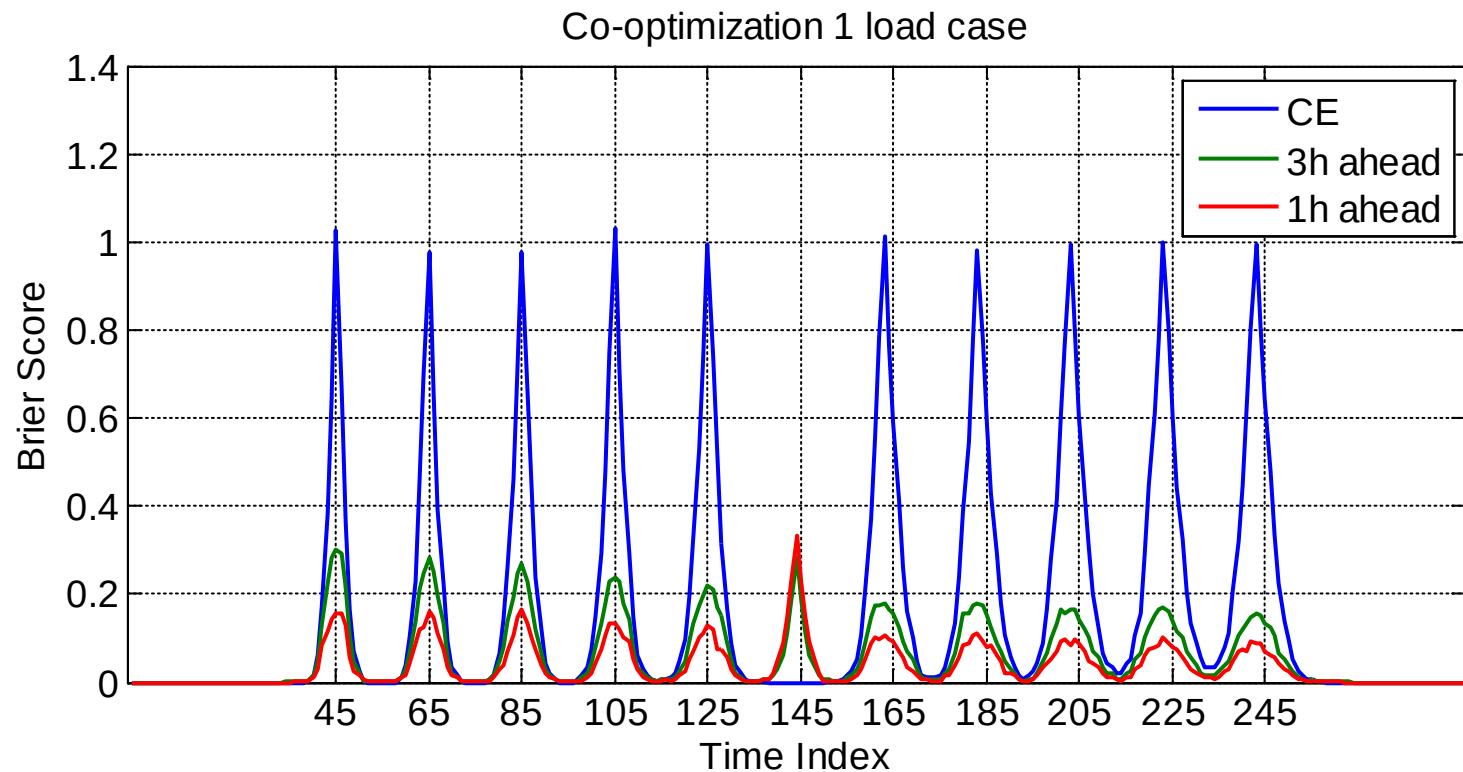
3hrs ahead Forecast at Time 45 (TV=0.0014)



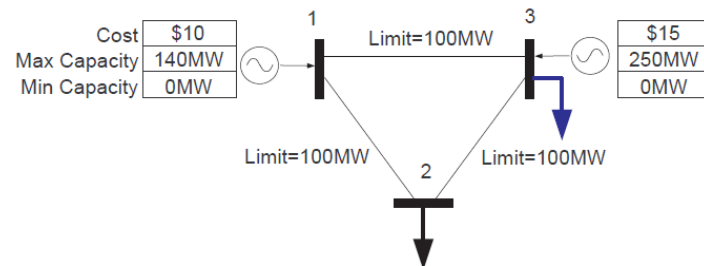
LMP forecast with energy-reserve co-optimization



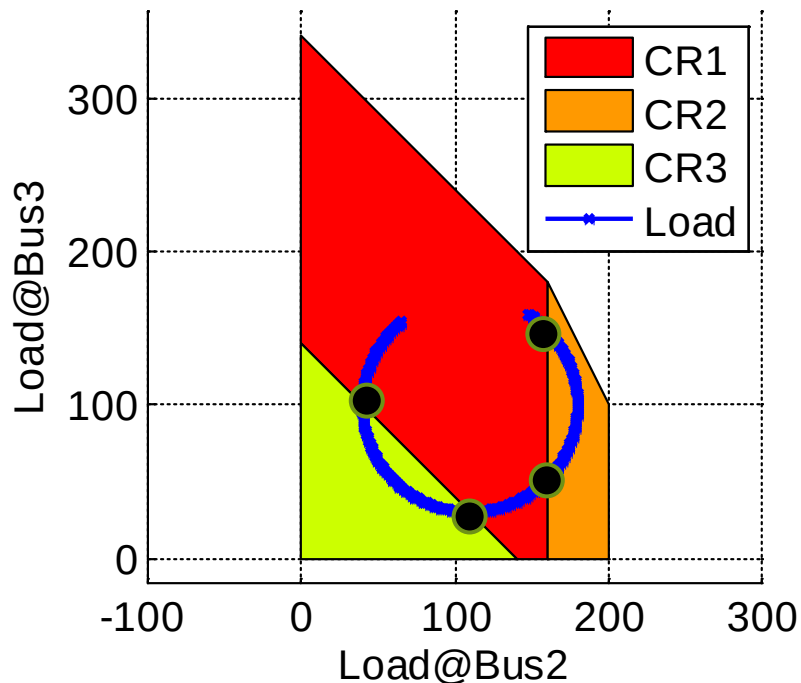
LMP forecast with energy-reserve co-optimization



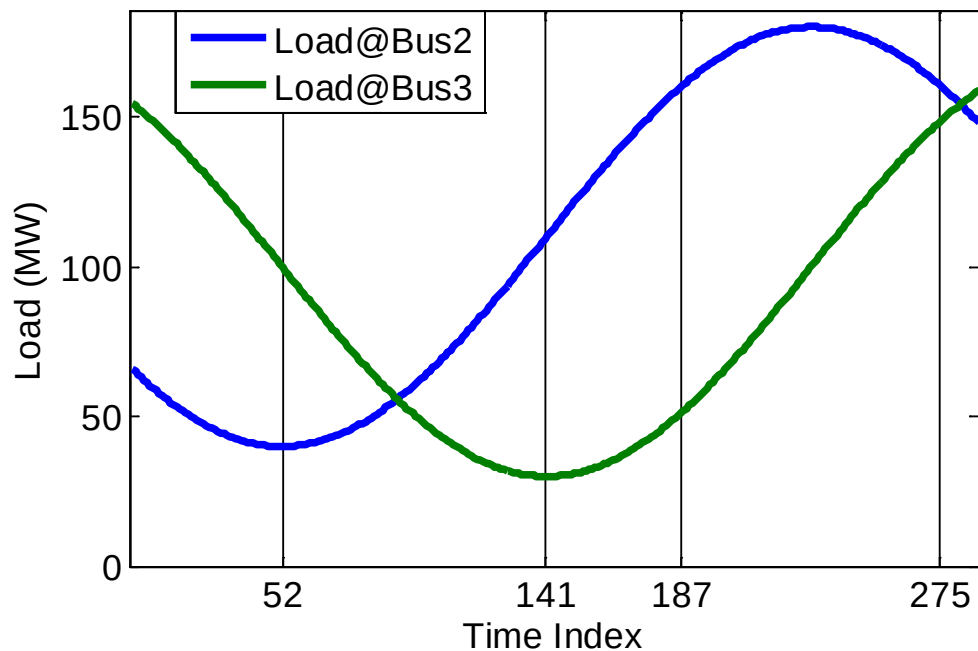
2D (mean) load trajectory



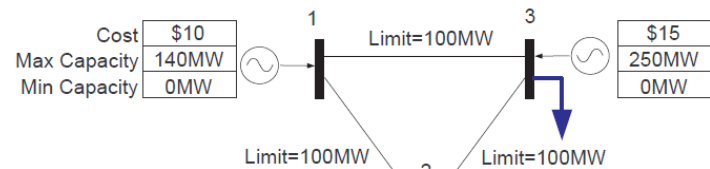
Critical Regions for 3Bus 2Load MLP-Alg1



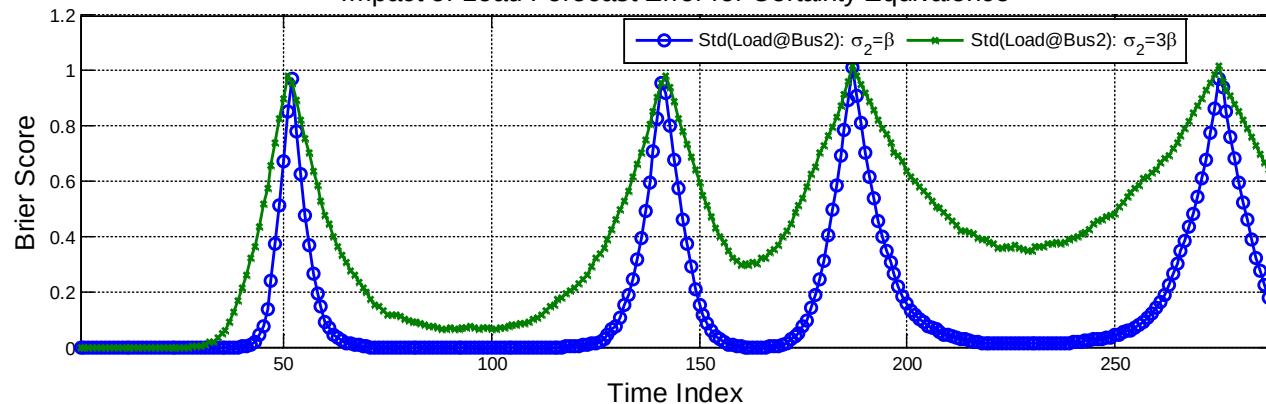
Load Forecast for 3Bus 2Load alg1



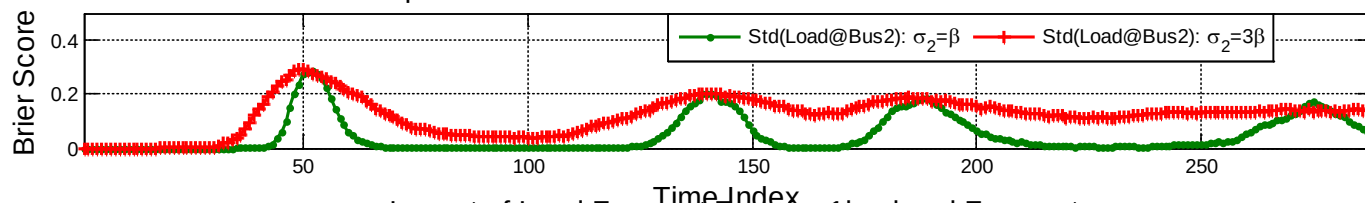
Load uncertainties



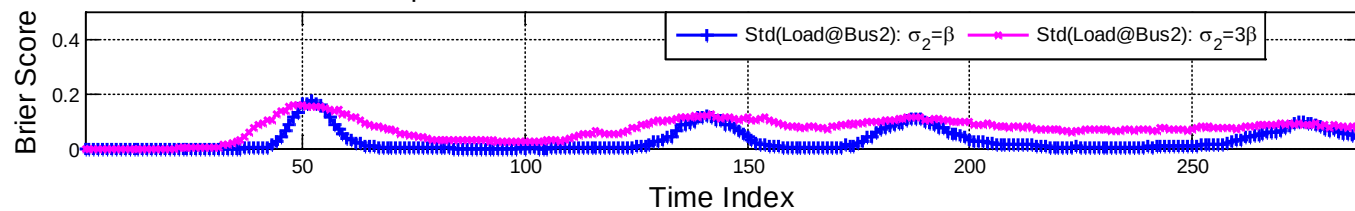
Impact of Load Forecast Error for Certainty Equivalence



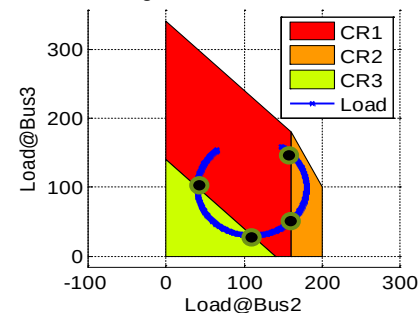
Impact of Load Forecast Error for 3hrs ahead Forecast



Impact of Load Forecast Error for 1hr ahead Forecast

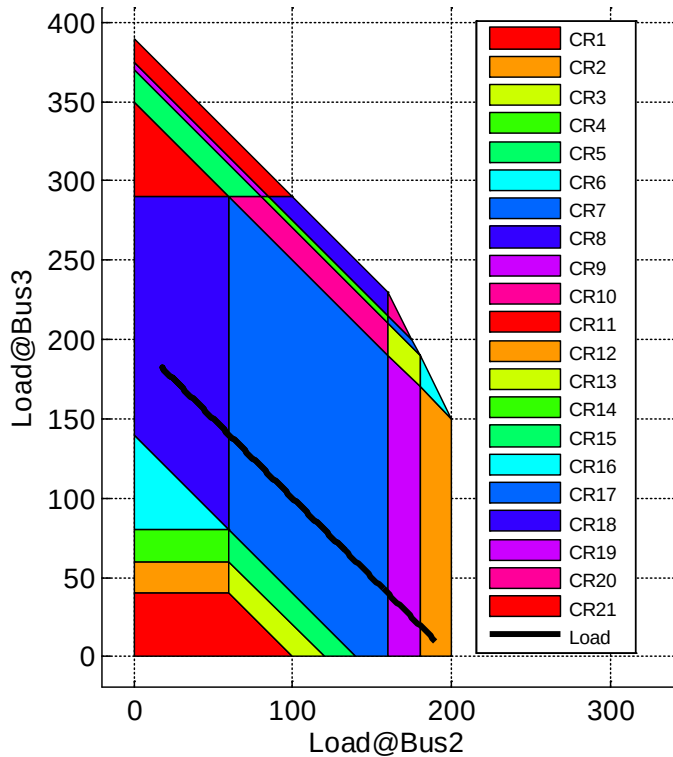


Critical Regions for 3Bus 2Load MLP-Alg1

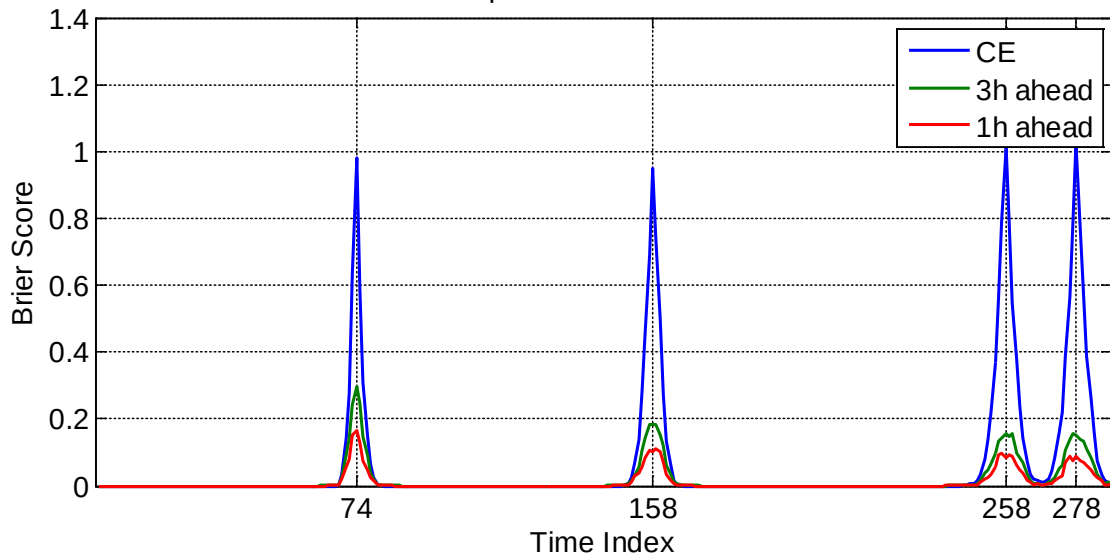


3 bus 2D system with energy-reserve co-optimization

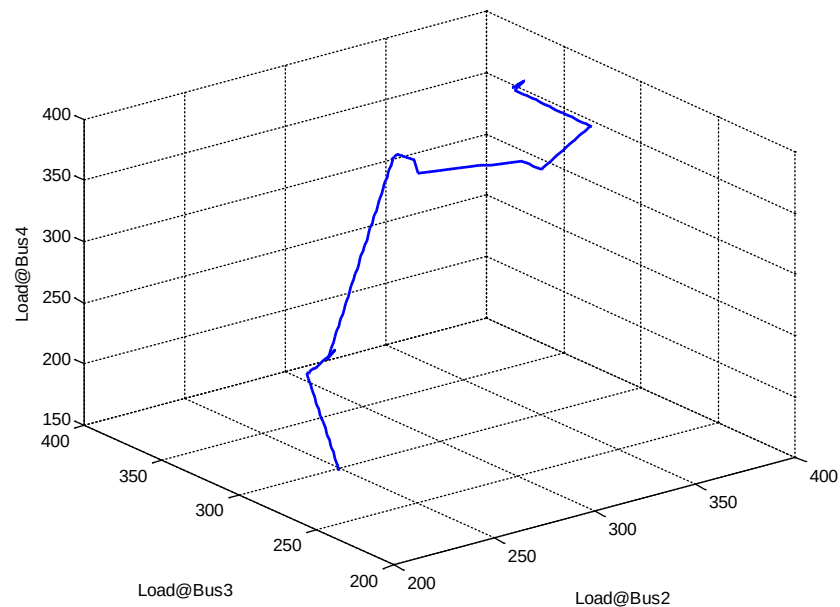
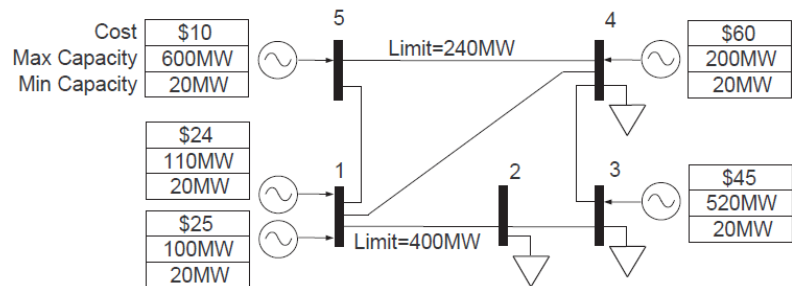
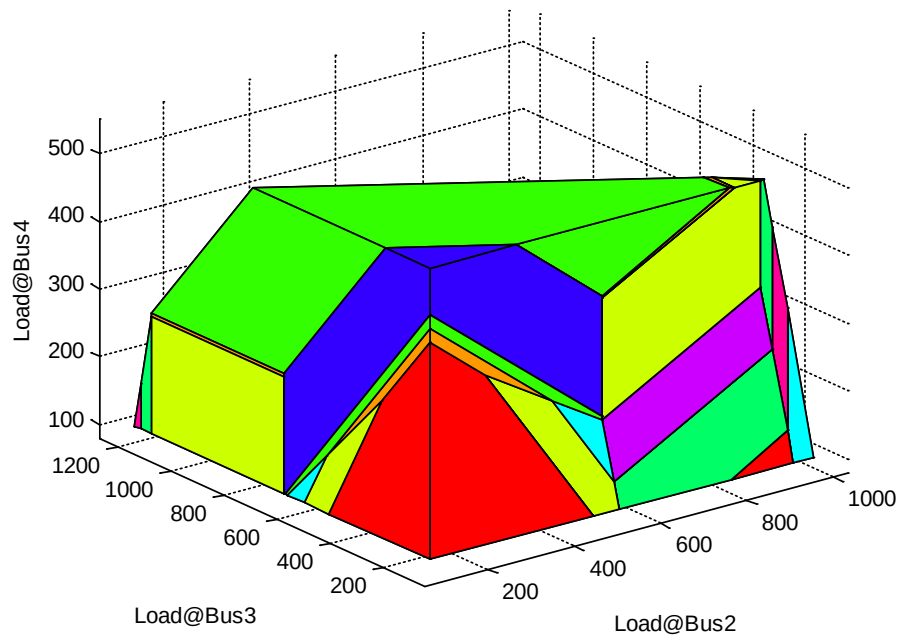
Critical Regions for 3Bus 2Load System in Co-Optimization



Co-optimization 2 load case



5 bus 3D load system



Sample paths of ex post LMP

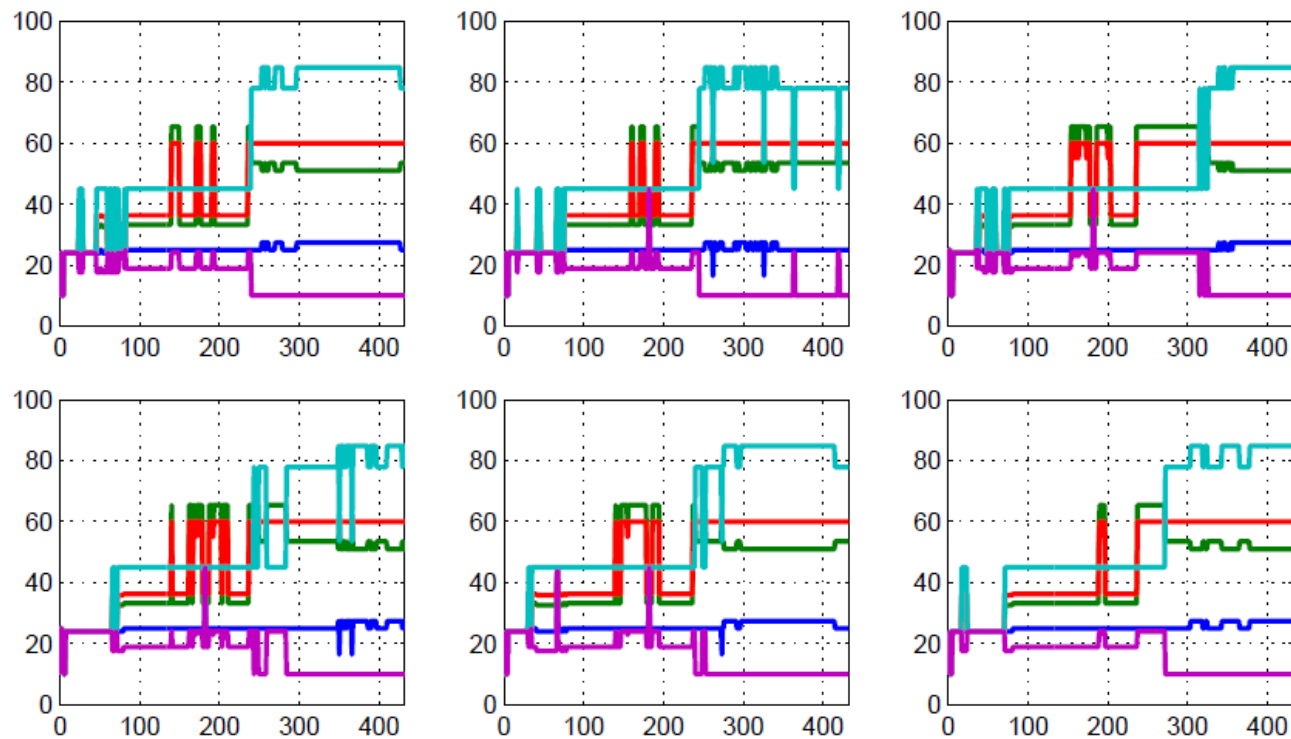
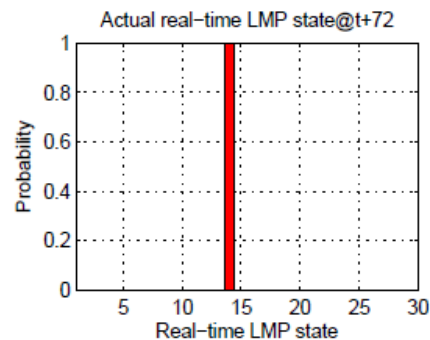
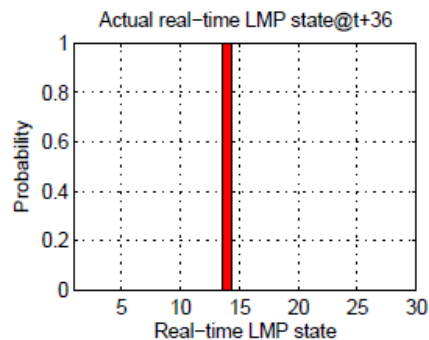
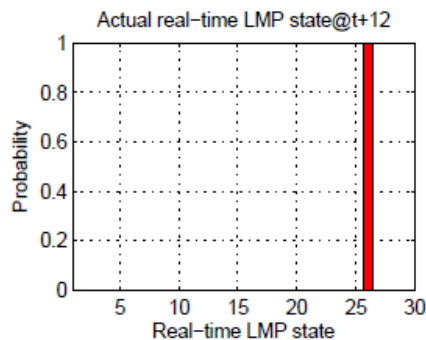
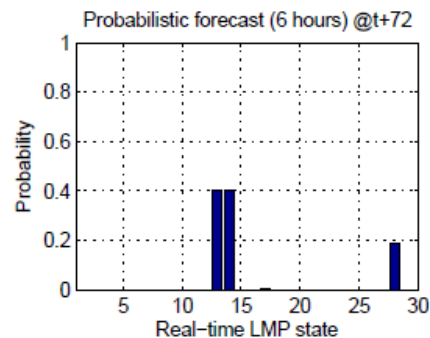
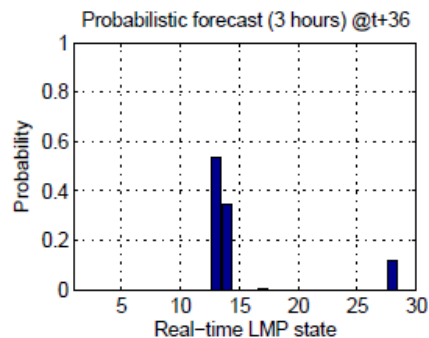
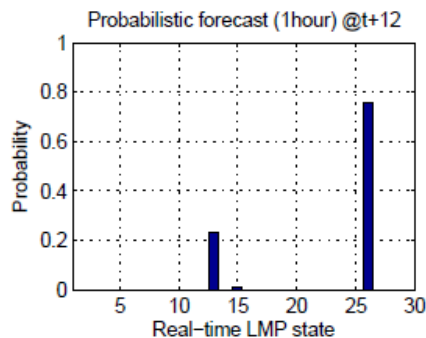
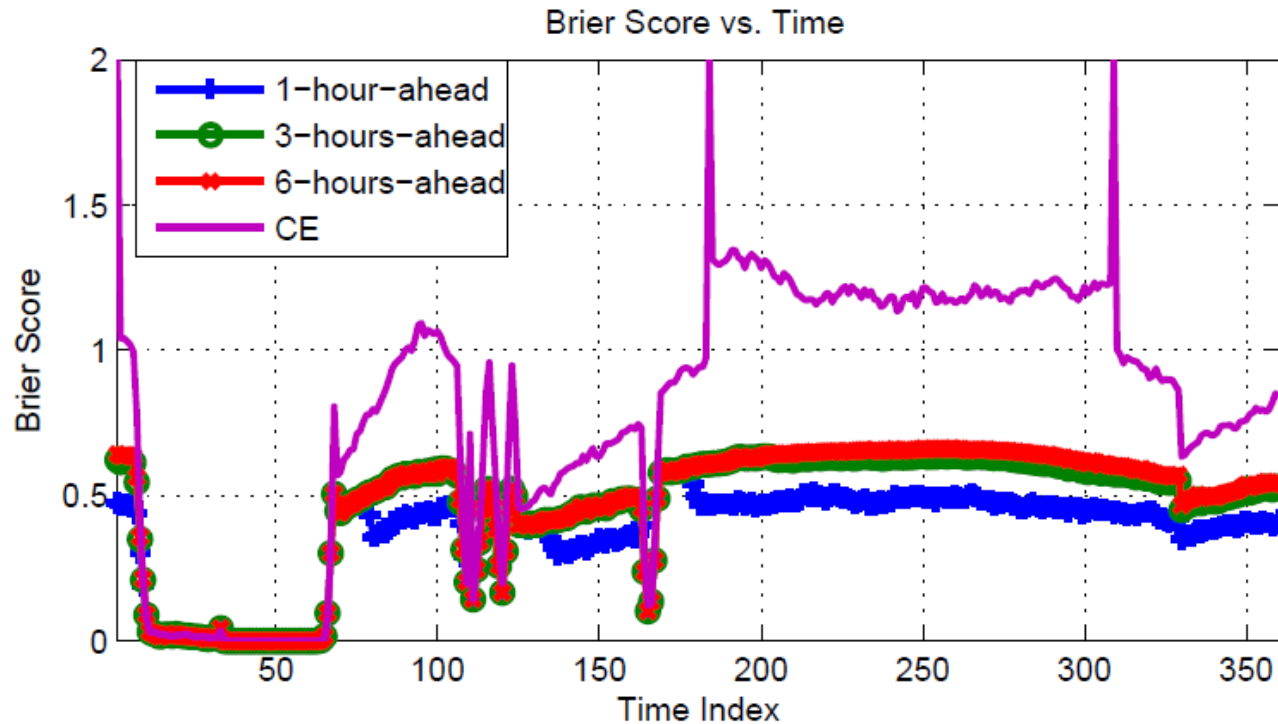


Figure : Real-Time LMP (\$/MW) vs. Time Index

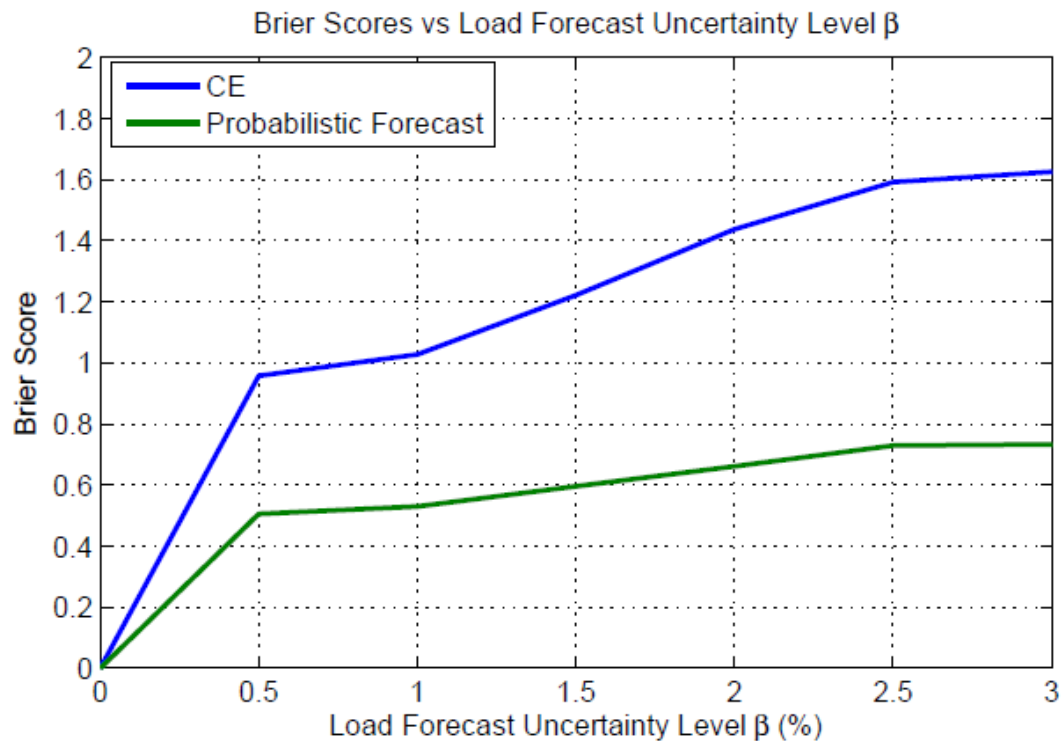
Sample forecasts



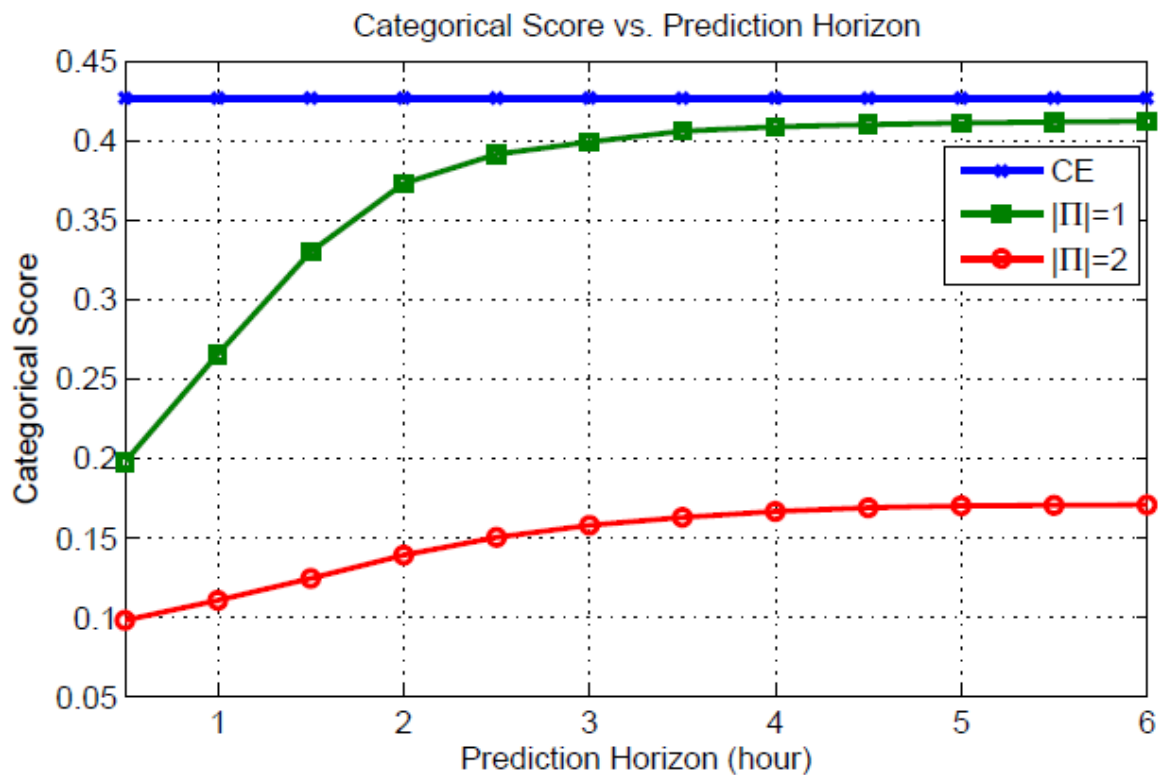
Performance of Markov distribution forecast



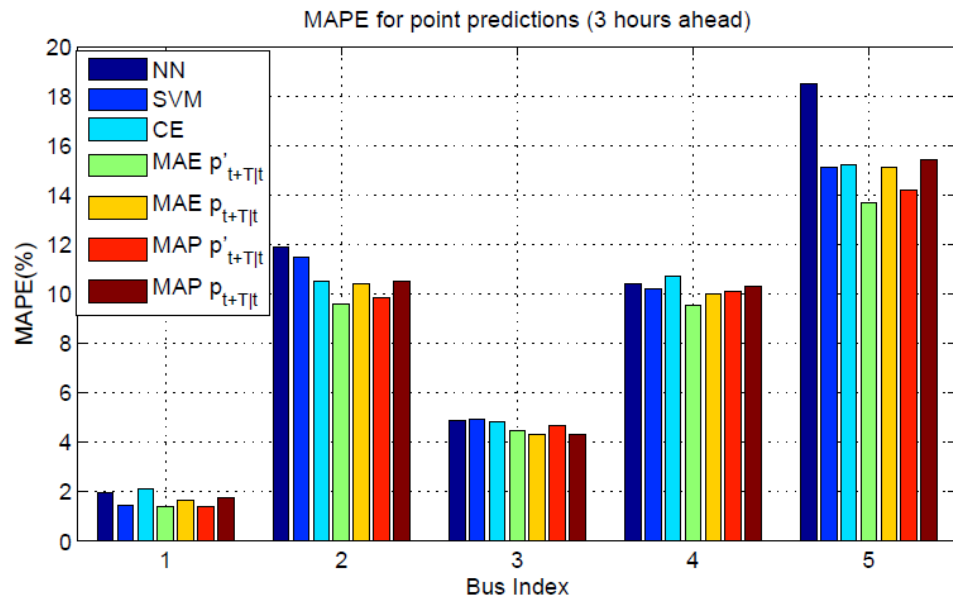
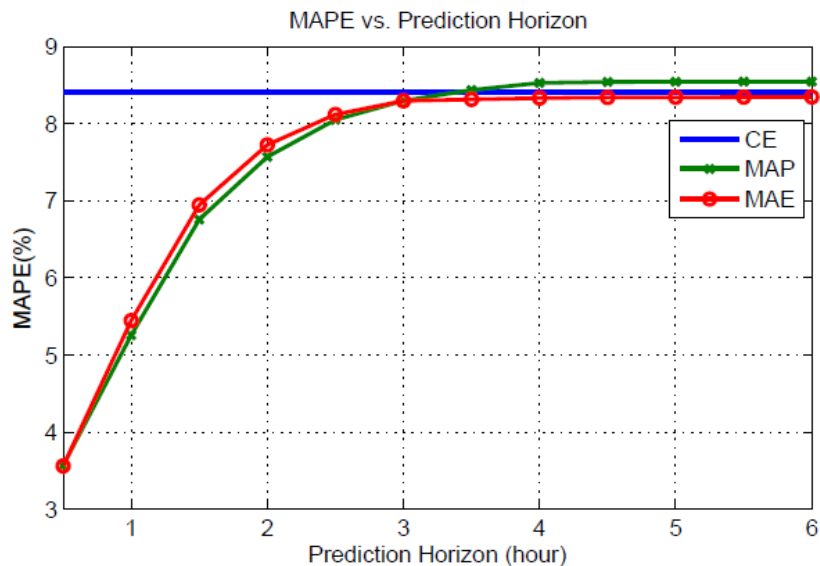
Effects of load uncertainty



Categorical score of MAP and “best of two”



MAPE performance



Summary and future work

□ Summary of results

- ▣ Real-time LMP process is highly structured
- ▣ Real-time LMP models that incorporating forecasting and measurement uncertainties and real-time measurements (e.g. SCADA/PMU).
- ▣ Probabilistic forecast techniques
 - A multi-parametric linear program approach for ex ante LMP forecast
 - A Markov chain approach for ex post LMP forecast

□ Future work

- ▣ LMP forecast with renewable generation
- ▣ Toward a scalable solution
 - Approximate MPLP
 - Efficient Monte Carlo techniques