





Online Inertia Estimation From PMU measurements

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Accurate inertia estimation is important

Higher renewable penetration \rightarrow lower inertia \rightarrow lower frequency nadir



a. The ERCOT simulated frequency responses (2,750 MW generation loss)

b. The WECC simulated frequency responses (2, 625 MW generation loss)

Fig. Frequency response under different renewable penetration levels

Fig. System equivalent inertia at different renewable penetration levels [SuNLAMP]

Problem Background

Renewable energy increase will significantly amplify the volatility and uncertainty of system inertia.



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Inertia Estimation Techniques

	In	put	data	Performance		
Methods	EMS	PM U	Event informa tion	Includes load, IBR, and other behind-the- meter inertia	Results impacted by FFR and load damping	Comments
Dispatch- based	\checkmark	×	×	1	×	Easy to do, but IBR control, load, other artificial inertia may be lost
Event-driven	×		\checkmark	\checkmark	\checkmark	Most accurate, needs to wait for events,
Probing signal	×	\checkmark	×	\checkmark	\checkmark	Better accuracy than ambient, real time, invasive, added costs
Ambient signal	×		×	\checkmark	\checkmark	Real time, low cost, data processing has challenges

1 The dispatch-based inertia can include the behind-the-meter inertia by including correction factors, but these behind-meter inertias are not measured directly.



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Machine learning based method

Minimum Volume Enclosing Ellipsoid (MVEE) from ambient-frequency features



Weather correlation

- Correlation Between
 Inertia and Weather
 Condition
- Average temperature of six cities in WECC: Los Angeles, Phoenix, Salt Lake City, Denver, Las Vegas and Seattle.



Fig. Correlation between daily average inertia and daily average temperature



Machine learning – WECC results vs NERC Data

Inputs to ML:

- Ambient frequency
- Weather
- Typical load profile



Performance of the machine-learning based inertia estimation using ambient frequency signal



25

25

Data + Physics based method

'relative inertia' results from island #1

/Noi

Local sunrise time Local sunset time





Potential Solution from Pumped Storage Hydro



Potential Solution from Pumped Storage Hydro

PMU data of ten Bath county pump switching off events show that the MW change is relatively constant.



Event #	Time EDT	Step change, MW
1	06/30/2021 13:13:30	347.7
5	06/28/2021 11:11:00	342.5
6	06/24/2021 05:52:23	339.2
7	06/18/2021 07:05:26	339.8
8	06/12/2021 08:51:15	339.1
9	05/30/2021 07:27:00	343.5
10	05/17/2021 02:25:00	344.8

MW step change difference

(Max-Min)/Average=(347.7-339.1)/342.4=**2.5%**



Online Inertia Estimation System Design



