



XSICEL 2021

Transición energética en la 4ta revolución industrial



Universidad
Tecnológica
de Pereira



UNIVERSIDAD
NACIONAL
DE COLOMBIA

Voltage Forecasting Methods for Voltage Monitoring and Control Strategies

Authors: Víctor Meza and Ernesto Perez

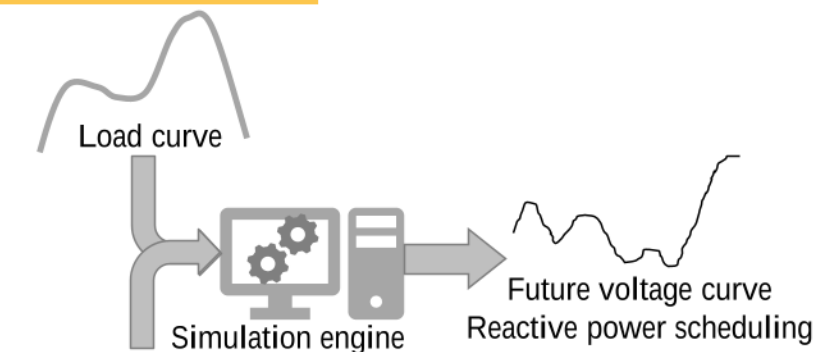
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Contents

- I. Introduction: Challenges**
- II. Proposed alternative**
- III. Cases and Results**
- IV. Improving accuracy and Covid-19 scenario**
- V. Questions**

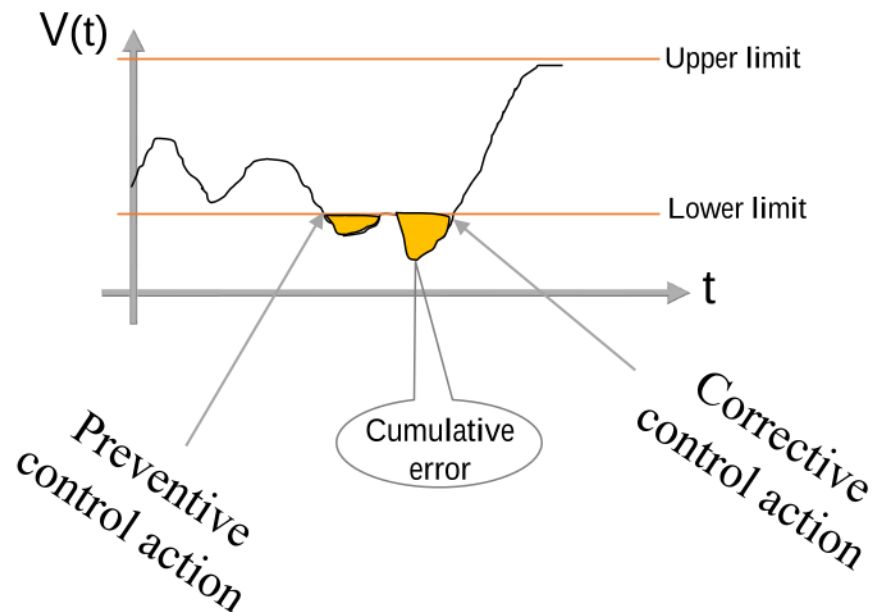
I. Introduction: Challenges

DER and DSM



Generation output curve

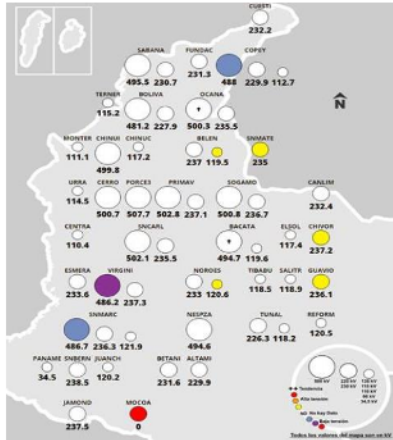
Uncertainty



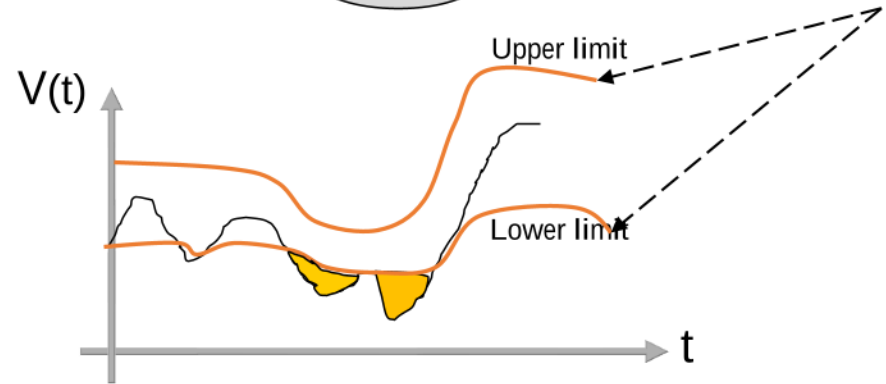
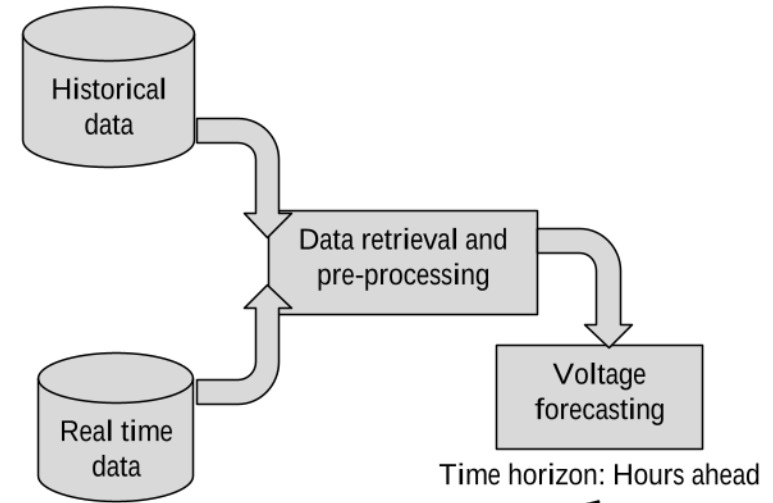
II. Proposed alternative

Control limits can vary:

- From node to node depending on:
 - Operational scenario
 - Network topology
 - SCR
 - Reactive power resources



Source: XM S.A. E.S.P.



III. Real-World Cases

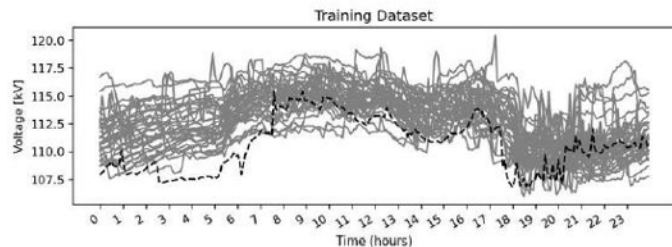


Fig.4.(a). Voltage time series with PV plant connected at 110 kV bus.

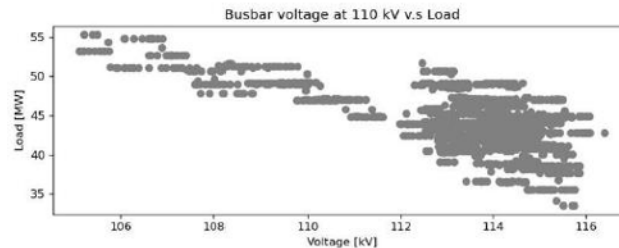


Fig.4.(b). Highly time-dependent voltage time series.

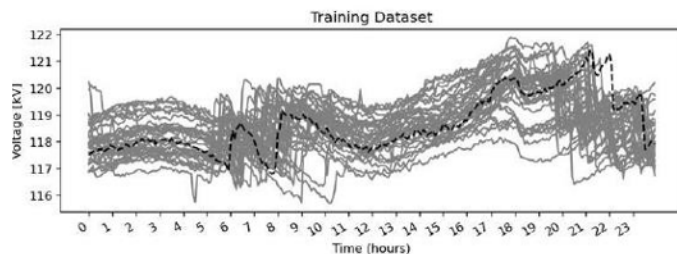


Fig.4.(c). Voltage time series with discrete control.

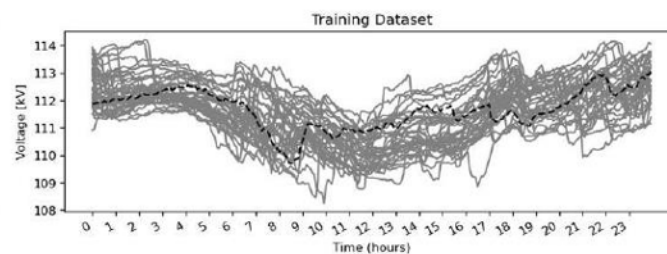
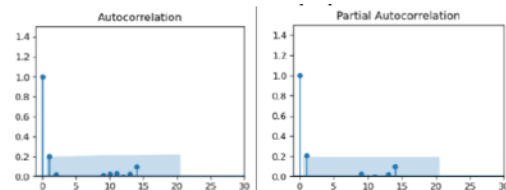


Fig.4.(d). Voltage time series in meshed grid (pilot node).

II. Model's performance comparison

ARIMA(1,0,0) or Autoregressive (AR)



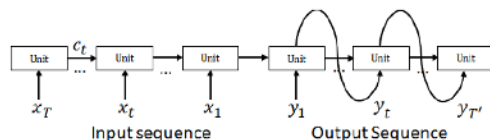
Simple Exponential Smoothing (SES)

$$V_t = \sum_{i=0}^{T-1} \alpha(1-\alpha)^i V_{T-t} + (1-\alpha)^T \hat{V}_1, \quad 0 < \alpha \leq 1$$

Seasonal Historical Median (SHM)

$$V_t = V_{t-1} \frac{1}{V_{max}^H T} \sum_{i=1}^H V_{t-i,D}, \quad V_{t-i,D} > 0.5 p.u$$

Recurrent Neural Networks (RNN)



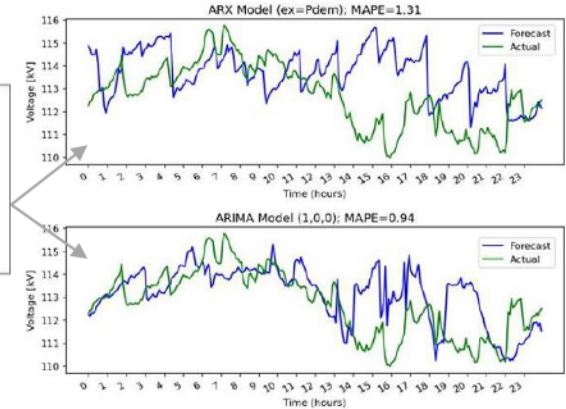
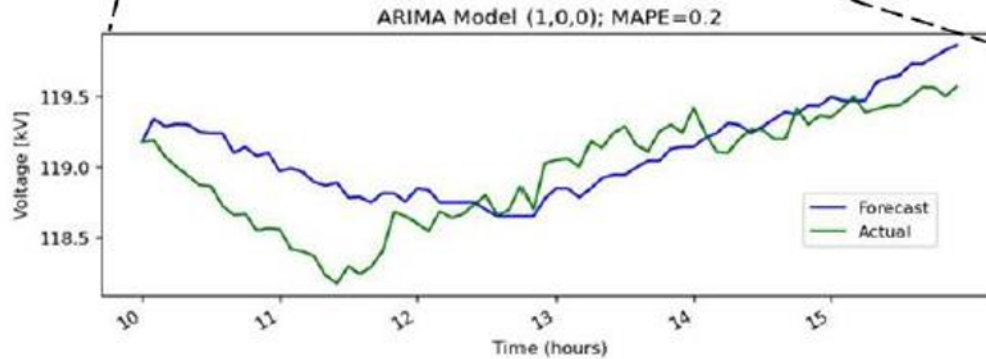
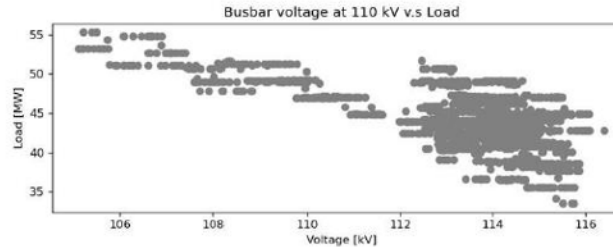
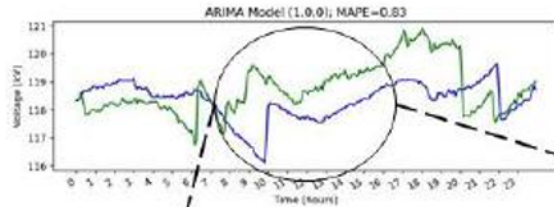
Model name	Metrics		
	Processing time [s]	Training dataset size	Time Horizon [h]
ARIMA(1,0,0)	Less than 1s	50 series	6,12,24
SES	Less than 1s		
SHM	Less than 1s		
RNN	73,76,92		
ARIMA(1,0,0)	Less than 1s	200 series	6,12,24
SES	Less than 1s		
SHM	Less than 1s		
RNN	29,94,23		

Model name	Metrics		
	Average MAPE [%]	Cases 1,2,3,4,5 MAPE [%]	Time Horizon [h]
ARIMA(1,0,0)	0.70	1.14, 0.92, 0.52, 0.41, 0.51	6
SES	0.71	1.18, 0.90, 0.52, 0.42, 0.52	
SHM	0.67	1.04, 1.10, 0.41, 0.39, 0.42	
RNN	1.02	1.72, 1.63, 0.6, 0.55, 0.60	
Persistence	1.10	2.01, 1.77, 0.6, 0.59, 0.51	
ARIMA(1,0,0)	0.89	1.1, 1.46, 0.68, 0.49, 0.70	12
SES	0.90	1.18, 1.47, 0.70, 0.50, 0.69	
SHM	0.68	0.93, 1.19, 0.42, 0.41, 0.47	
RNN	1.20	2.01, 1.97, 0.57, 0.66, 0.78	
Persistence	1.22	2.23, 1.99, 0.6, 0.65, 0.63	
ARIMA(1,0,0)	0.83	1.16, 1.20, 0.65, 0.52, 0.64	24
SES	0.84	1.18, 1.21, 0.66, 0.52, 0.64	
SHM	0.68	0.95, 1.04, 0.48, 0.41, 0.53	
RNN	1.27	2.2, 2.06, 0.69, 0.58, 0.84	
Persistence	1.27	2.40, 1.81, 0.68, 0.60, 0.85	

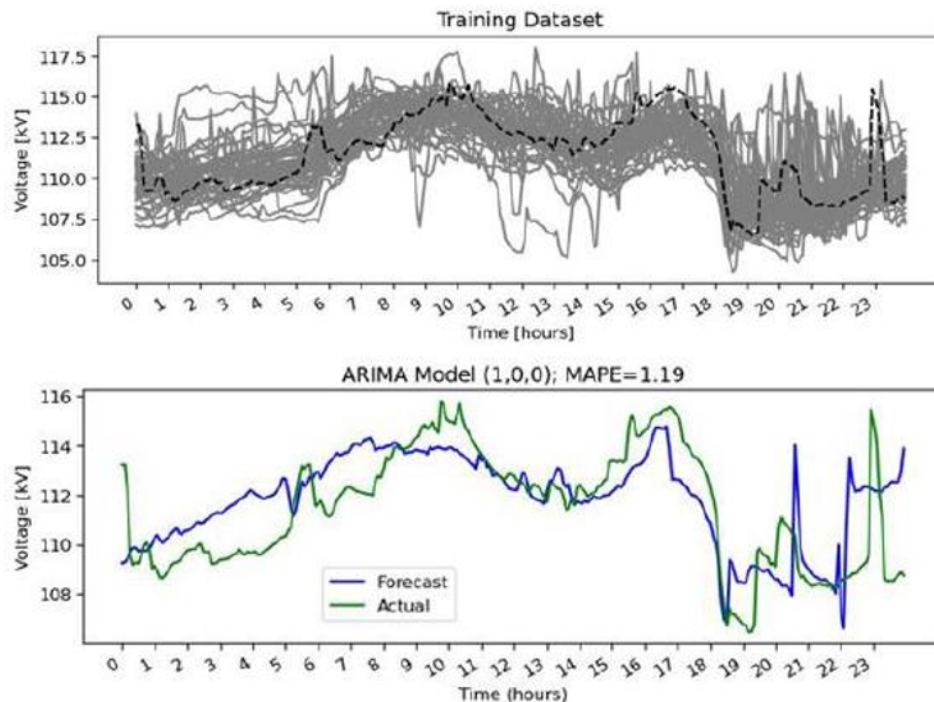
IV. Improving accuracy

AR or ARX Models

Time horizon:
6 hours ahead



II. Covid 19 Scenario



Gracias

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