Log(v) 3LPF: A linearized solution to train reinforcement learning algorithms for distribution systems

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SPADES Y1 workshop. December 2, 2020

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Reinforcement learning on distribution systems, PPO-Clip

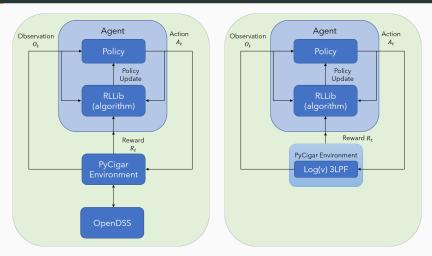


Figure 1: Current PyCigar modeling diagram (left) vs proposed architecture (right)

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Reinforcement learning on distribution systems, PPO-Clip

- Two neural networks. Weights obtained via SGD
 - Policy function: $\theta_{k+1} = \arg\min_{\theta} g(\hat{R}_t, s_{rl}, a_{rl}, \theta)$
 - Value function: $\phi_{k+1} = \arg\min_{\phi} h(\hat{R}_t, s_{rl}, a_{rl}, \phi)$
- Power Flow (PF) equations are used to compute the rewards
- Rewards are computed for every training iteration, every time step and every action sampled by the algorithm
- Efficient and accurate PF solvers are necessary

Our target:
$$\rightarrow \widehat{R}_t = f(s_p^1, s_{rl}^2, a_{rl}^3)$$

 $^{^{1}}s_{p}$: Power system state

 $^{^{2}}s_{rl}$: Reinforcement learning state

 $^{^3}a_{rl}$: Reinforcement learning action

Distribution systems are unbalanced

- ullet Unbalanced system o 3-phase solvers
- Need for AC modeling vs DC
 - AC PF equations are non-linear
 - Non-linearity is caused by ZIP load models (details later)
- Modeling of line losses through shunt elements (not negligible) needed

PF formulations:

Current: $i = Y_{bus}v$ Power: $s = D(vv^H Y_{bus}^H)$

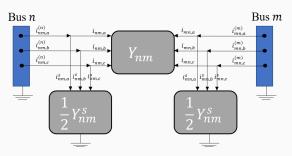


Figure 2: Pi-Model representation

Survey of ACPF solvers

- Commercial software PF solvers are iterative
 - Newton-Raphson (GridLab-D, PSLF)
 - Gauss-Seidel (OpenDSS,GridLab-D)
 - Forward-Backward Sweep (GridLab-D)
- Previous work, linear approximations
 - Lin3DistFlow [Sankur et al., 2016]. Nominal voltages and no losses.
 No ZIP models
 - NFA [Fobes et al., 2020]. Real-power only, no ZIP models
 - DCP [Fobes et al., 2020]. DC assumption, ignores reactive power, no ZIP models
 - Lossy Distflow [Schweitzer et al., 2019]. Not valid for complete ZIP models, cannot acommodate modeling of transformers, regulators, losses are parametrized.
 - LPF [Li et al., 2017]. Positive-sequence only, lossless, no ZIP models, doesn't exploit tree structure (use case is transmission systems).

Our contribution

Modeling capabilities

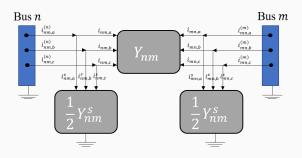
- Embedded linear power flow solver
- ZIP load models, shunt capacitors, regulators, transformers, smart inverters, batteries, and corresponding controls (as of today).
- Ability to fully control and understand unbalanced 3-phase distribution systems. Implementation of new attack vectors
- Linear PF and OPF. Applicable to transmission systems.

Modeling accuracy

- PF solution validated against OpenDSS
- Modeling of devices matches that of OpenDSS

Computational complexity

- Linear equations allow to solve the system in a single snapshot. In OpenDSS, IEEE-13 11 iterations, IEEE-8500 62 iterations reducing RL training times
- Ability to exploit graph tree structure
- Ability to efficiently compute inverse after perturbation
- Reduced overhead due to api calling external models



Kirchoff:
$$i_{nm}^{(n)} = i_{nm} + i_{nm}^{s}$$

Ohm: $i_{nm} = Y_{nm}^{(n)} \mathbf{v}_n - Y_{nm}^{(m)} \mathbf{v}_m$ $i_{nm}^{(n)} = \left(Y_{nm}^{(n)} + \frac{1}{2}Y_{nm}^{s}\right) \mathbf{v}_n - Y_{nm}^{(m)} \mathbf{v}_m$

Losses:
$$\boldsymbol{i}_{nm}^s = \frac{1}{2} Y_{nm}^s \boldsymbol{v}_n$$
 $\boldsymbol{S}_{nm}^{(n)} = \boldsymbol{v}_n \left(\boldsymbol{i}_{nm}^{(n)} \right)^H$

$$\boldsymbol{S}_{nm}^{(n)} = \boldsymbol{v}_n \boldsymbol{v}_n^H \left(Y_{nm}^{(n)} + \frac{1}{2} Y_{nm}^s \right)^H - \boldsymbol{v}_n \boldsymbol{v}_m^H \left(Y_{nm}^{(m)} \right)^H$$

$$\boldsymbol{S}_{nm}^{(n)} = \boldsymbol{v}_n \boldsymbol{v}_n^H \left(\boldsymbol{Y}_{nm}^{(n)} + \frac{1}{2} \boldsymbol{Y}_{nm}^s \right)^H - \boldsymbol{v}_n \boldsymbol{v}_m^H \left(\boldsymbol{Y}_{nm}^{(m)} \right)^H$$

We want to remove the non-linearity $\mathbf{v}_n \mathbf{v}_n^H$ from the equation that relates power flows to voltage

$$\mathbf{v}_n := \begin{bmatrix} v_n^a, v_n^b, v_n^c \end{bmatrix}^T, \quad v_n^p = |v_n^p| \, e^{j\theta_p} \tag{1}$$

$$|v_n^p| = e^{\log|v_n^p|}, \quad u_n^p := \log|v_n^p|$$
 (2)

$$v_n^p = e^{u_n^p} e^{j\theta_p}$$
 (3)

$$\mathbf{S}_{nm}^{(n)} = \mathbf{v}_n \mathbf{v}_n^H \left(\mathbf{Y}_{nm}^{(n)} + \frac{1}{2} \mathbf{Y}_{nm}^s \right)^H - \mathbf{v}_n \mathbf{v}_m^H \left(\mathbf{Y}_{nm}^{(m)} \right)^H$$

We know voltages are around 1 p.u., thus $\log(1) = 0$, and we approximate the voltage magnitude in $v_n^p = e^{u_n^p} e^{j\theta_p}$ using first-order Taylor expansion.

$$\mathbf{v}_{n} := \begin{pmatrix} e^{u_{n}^{a}} e^{j\theta_{n}^{a}} \\ e^{u_{n}^{b}} e^{j\theta_{n}^{b} + \frac{2\pi}{3} - \frac{2\pi}{3}} \\ e^{u_{n}^{c}} e^{j\theta_{n}^{c} - \frac{2\pi}{3} + \frac{2\pi}{3}} \end{pmatrix} = \mathbf{\Delta}_{3} \operatorname{diag} \begin{pmatrix} e^{u_{n}^{a}} \\ e^{u_{n}^{b}} \\ e^{u_{n}^{c}} \end{pmatrix} \begin{pmatrix} e^{j\tilde{\theta}_{n}^{a}} \\ e^{j\tilde{\theta}_{n}^{c}} \\ e^{j\tilde{\theta}_{n}^{c}} \end{pmatrix}$$
(4)

$$\mathbf{v}_n \approx \mathbf{\Delta}_3 \left(\mathsf{I} + \mathsf{diag} \left(\mathbf{u}_n \right) \right) e^{j\theta_n}$$
 (5)

$$\mathbf{v}_{n}\mathbf{v}_{n}^{H} \approx \mathbf{\Delta}_{3} \left(11^{T} + \mathbf{u}_{n}1^{T} + 1^{T}\mathbf{u}_{n}^{T} + j\tilde{\boldsymbol{\theta}}_{n}1^{T} - j1\tilde{\boldsymbol{\theta}}_{n}^{T}\right) \mathbf{\Delta}_{3}^{H}$$

$$\mathbf{v}_{n}\mathbf{v}_{m}^{H} \approx \mathbf{\Delta}_{3} \left(11^{T} + \mathbf{u}_{n}1^{T} + 1^{T}\mathbf{u}_{m}^{T} + j\tilde{\boldsymbol{\theta}}_{n}1^{T} - j1\tilde{\boldsymbol{\theta}}_{m}^{T}\right) \mathbf{\Delta}_{3}^{H}$$
(6)

$$\textbf{Non-linear ACPF:} \quad \boldsymbol{S}_{nm}^{(n)} = \boldsymbol{v}_{n}\boldsymbol{v}_{n}^{H} \left(\boldsymbol{Y}_{nm}^{(n)} + \frac{1}{2}\boldsymbol{Y}_{nm}^{s}\right)^{H} - \boldsymbol{v}_{n}\boldsymbol{v}_{m}^{H} \left(\boldsymbol{Y}_{nm}^{(m)}\right)^{H}$$

Reordering and defining the corresponding matrices and vectors

Log(v) 3LPF (Linear):

$$\tilde{\mathbf{s}}_{nm} \approx \tilde{\mathbf{Y}}_{\text{bus}} \mathbf{x}$$
 where $\mathbf{x} \triangleq \begin{bmatrix} \mathbf{u} \\ \tilde{\mathbf{\theta}} \end{bmatrix}$ (7)

$$extbf{ iny x} pprox ilde{ extsf{Y}}_{ extsf{bus}}^{-1} ilde{ extbf{ iny s}}_{ extsf{nm}}$$

and we may recover the voltage phasors as follows

$$oldsymbol{v}pproxoldsymbol{\Delta}_3 {\sf diag}\left(e^{oldsymbol{u}}
ight)e^{j ilde{ heta}_n}$$

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ZIP models

ZIP models. Wye-connected loads

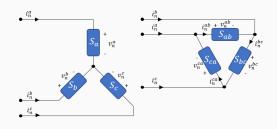


Figure 3: Wye (left) and delta-connected (right) load

Wye-connected loads:
$$S_n^Y = S_n^{Z,Y} + S_n^{I,Y} + S_n^{P,Y}$$

Z:
$$S_n^{Z,Y} = (y_n^Y)^* + 2\operatorname{diag}(y_n^Y)^* \boldsymbol{u}_n$$

I:
$$S_n^{I,Y} \approx \Delta_3 \left(i_n^Y\right)^* + \Delta_3 \text{diag} \left(i_n^Y\right)^* u_n + j\Delta_3 \text{diag} \left(i_n^Y\right)^* \tilde{\theta}_n$$

P:
$$S_n^{P,Y} = s_{\ell}^{Y}$$

ZIP models. Delta-connected loads

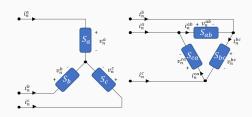


Figure 4: Wye (left) and delta-connected (right) load

Delta-connected loads:
$$S_n^{\Delta} = S_n^{Z,\Delta} + S_n^{I,\Delta} + S_n^{P,\Delta}$$

$$\begin{aligned} \textbf{Z:} \quad & \mathsf{S}_n^{\mathsf{Z},\Delta} \approx \left(\mathsf{diag} \left(\boldsymbol{\Delta}_3 \left(\tilde{\mathsf{Y}}_n^{\Delta} \right)^\mathsf{T} \mathbf{1} \right) + \boldsymbol{\Delta}_3 \left(\tilde{\mathsf{Y}}_n^{\Delta} \right)^\mathsf{T} \right) \boldsymbol{u}_n \\ & \quad + \left(\mathsf{diag} \left(\boldsymbol{\Delta}_3 \left(\tilde{\mathsf{Y}}_n^{\Delta} \right)^\mathsf{T} \mathbf{1} \right) - \boldsymbol{\Delta}_3 \left(\tilde{\mathsf{Y}}_n^{\Delta} \right)^\mathsf{T} \right) \tilde{\boldsymbol{\theta}}_n \end{aligned}$$

I:
$$S_n^{I,\Delta} \approx \Delta_3 \Lambda \left(i_n^Y \right)^* + \Delta_3 \text{diag} \left(\Lambda i_n^Y \right)^* u_n + j \Delta_3 \text{diag} \left(\Lambda i_n^Y \right)^* \tilde{\theta}_n$$

P:
$$S_n^{P,\Delta} \approx \Lambda s_\ell^{\Delta}$$

Modeling and control of power delivery elements

Transformers and voltage regulators

Modeled through the **admittance matrix**

$$\boldsymbol{S}_{nm}^{(n)} = \boldsymbol{v}_n \boldsymbol{v}_n^H \left(\boldsymbol{\mathsf{Y}}_{nm}^{(n)} + \frac{1}{2} \boldsymbol{\mathsf{Y}}_{nm}^s \right)^H - \boldsymbol{v}_n \boldsymbol{v}_m^H \left(\boldsymbol{\mathsf{Y}}_{nm}^{(m)} \right)^H$$

Transformers:

$$\mathbf{Y}_{\mathsf{prim}} = \mathsf{ANB} \left(\mathbf{Z}_{nm}^{t} \right)^{-1} \mathbf{B}^{\mathsf{T}} \mathbf{N}^{\mathsf{T}} \mathbf{A}^{\mathsf{T}}$$

$$\mathbf{Y}_{\mathsf{prim}} = \left(\begin{array}{cc} \mathbf{Y}_{nm}^{(n)} & \mathbf{Y}_{nm}^{(m)} \\ \mathbf{Y}_{mn}^{(n)} & \mathbf{Y}_{mn}^{(m)} \end{array} \right)$$

Voltage regulators:

$$\begin{aligned} \mathsf{Y}_{\mathsf{prim}} &= \Gamma \mathsf{ANB} \left(\mathsf{Z}_{nm}^t \right)^{-1} \mathsf{B}^{\mathrm{T}} \mathsf{N}^{\mathrm{T}} \mathsf{A}^{\mathrm{T}} \Gamma^{\mathrm{T}} \\ \mathsf{where} \quad \Gamma &= \mathsf{diag} \left(\gamma \right) \quad \mathsf{and} \quad \gamma_i = 1 \pm 0.00625 \tau_i \\ \tau_i &= f \big(\mathsf{v}_{\mathsf{reg}}, \mathsf{v}_b, \mathsf{v}_n \big) \end{aligned}$$

Attack vectors

 v_{reg} setpoint, v_b bandwidth , v_n measurement

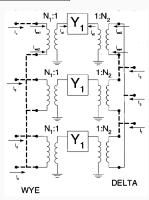


Figure 5: WyeG-Delta Transformer

Shunt capacitors and cap controls

Modeled as a wye or delta-connected constant impedance load

$$\left[\mathsf{S}_{n}^{\mathsf{Z},\mathsf{Y}}=\mathsf{D}\left(\boldsymbol{\mathit{v}}_{n}\boldsymbol{\mathit{v}}_{n}^{H}\boldsymbol{\Pi}\left(\mathsf{Y}_{n}^{\mathsf{Y}}\right)^{H}\boldsymbol{\Pi}^{T}\right)\right]\quad\mathsf{or}\quad\left[\mathsf{S}_{n}^{\mathsf{Z},\Delta}=\mathsf{D}\left(\boldsymbol{\mathit{v}}_{n}\boldsymbol{\mathit{v}}_{n}^{H}\boldsymbol{\Pi}\left(\mathsf{Y}_{n}^{\Delta}\right)^{H}\boldsymbol{\Pi}^{T}\right)\right]$$

Capacitor banks:

$$\begin{aligned} \textbf{Y}_{n}^{\textbf{Y}} &= \mathsf{diag}\left(\textbf{y}_{n}^{\textbf{Y}}\right), \textbf{Y}_{n}^{\boldsymbol{\Delta}} &= \mathsf{diag}\left(\textbf{y}_{n}^{\boldsymbol{\Delta}}\right), \\ \textbf{y}_{n}^{\textbf{Y}} &= \left[\textbf{y}_{n,a}^{\textbf{Y}}, \textbf{y}_{n,b}^{\textbf{Y}}, \textbf{y}_{n,c}^{\textbf{Y}}\right]^{T}, \textbf{y}_{n}^{\boldsymbol{\Delta}} &= \left[\textbf{y}_{n,a}^{\boldsymbol{\Delta}}, \textbf{y}_{n,b}^{\boldsymbol{\Delta}}, \textbf{y}_{n,c}^{\boldsymbol{\Delta}}\right]^{T} \end{aligned}$$

Cap controls:

$$\mathbf{y}_{n}^{\mathsf{Y}} = \sum_{i=1}^{n_{c}^{\mathsf{S}}} \eta_{c,i} \mathbf{y}_{c,i}^{\mathsf{Y}}, \quad \mathbf{y}_{n}^{\Delta} = \sum_{i=1}^{n_{c}^{\mathsf{S}}} \eta_{c,i} \mathbf{y}_{c,i}^{\Delta}$$

where
$$n_c^s \to N$$
 steps, $\eta_{c,i} \in \{0,1\}$ $\eta_{c,i} = f(\vartheta^4, \overline{\vartheta}, \underline{\vartheta}^5)$

⁴Control input (current, voltage, kvar, PF, time)

⁵Attack vector. Upper and lower limits (operation is outside limits)

DER and smart inverters

Modeled as a wye-connected constant power load

$$\boxed{\mathsf{S}^{\mathsf{P},\mathsf{Y}}_n(t) = \mathsf{s}^\mathsf{Y}_\ell(t)}$$

Solar resources and batteries:

$$\mathsf{S}^{\mathsf{P},\mathsf{Y}}_n(t) = \mathsf{s}^{\mathsf{Y}}_\ell(t) \quad \text{and} \quad \mathsf{S}^{\mathsf{P},\mathsf{Y}}_n(t) = f(\eta_{c,t},\eta_{d,t},s_{oc}(t-1)^6)$$

Smart Inverters:

$$S_n^{P,Y}(t) = f(s_\ell^Y(t-1), v_{n,t}^7)^8$$

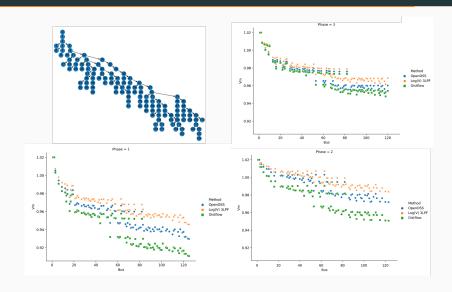
⁶State of charge

⁷Voltage measurement

⁸Attack vector. Changes in drop curve settings

Preliminary results

IEEE-123 test case



OpenDSS vs Log(V) 3LPF

RMSE (IEEE-123)	Distflow	Log(V) 3LPF
Phase 1	0.014	0.012
Phase 2	0.016	0.08
Phase 3	0.006	0.04

Case	Devices	Buses	Nodes	Losses (%)
IEEE-123	238	132	278	2.63
IEEE-8500	7282	4876	8531	10.58
European LV	965	907	2721	0.2703

Case	Iterations	OpenDSS (s)	Log(V) 3LPF (s)
IEEE-123	3	0.011	0.001
IEEE-8500	62	0.244	n/a
European LV	3	0.087	n/a

Current efforts. Future work

- Exploring techniques to solve the linear system of equations. E.g.
 - Forward-backward sweep
 - Truncated SVD
 - Parallel computation on leaf nodes
- Re-centering around 1 to obtain better approximation
- Sherman-Morrison for matrix inversion after perturbation
- Solving large cases directly from .dss files

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