

# **Optimization Algorithm Design via Electric Circuits**

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## Distributed convex optimization problem

$$\begin{array}{ll}\text{minimize} & f(x) \\ \text{subject to} & x \in \mathcal{R}(E^\top)\end{array}$$

- $f: \mathbf{R}^m \rightarrow \mathbf{R} \cup \{\infty\}$  is closed, convex, and proper
- $n$  nets  $N_1, \dots, N_n$  forming a partition of  $\{1, \dots, m\}$
- $E \in \mathbf{R}^{n \times m}$  is a selection matrix

$$E_{ij} = \begin{cases} +1 & \text{if } j \in N_i \\ 0 & \text{otherwise} \end{cases}$$

## Example: Consensus problem

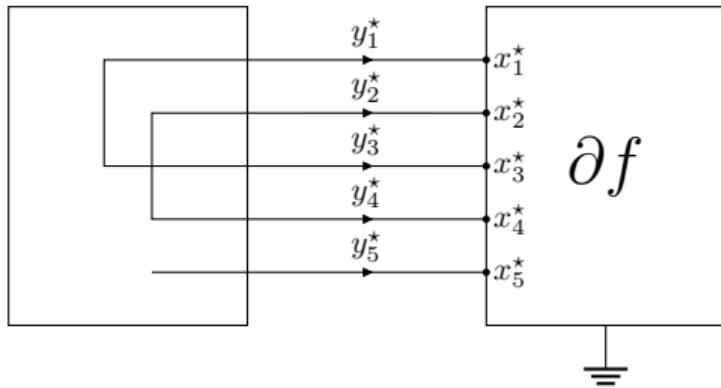
$$\begin{array}{ll}\text{minimize}_{x_1, \dots, x_N \in \mathbf{R}^{m/N}} & f_1(x_1) + \dots + f_N(x_N) \\ \text{subject to} & x_1 = \dots = x_N\end{array}$$

- $x = (x_1, \dots, x_N) \in \mathbf{R}^m$  is the decision variable
- $f(x) = f_1(x_1) + \dots + f_N(x_N)$  is block-separable
- $E^\top = (I, \dots, I) \in \mathbf{R}^{m \times m/N}$

## Circuit interpretation: KKT conditions

- $y \in \partial f(x)$  (stationarity)
- $x \in \mathcal{R}(E^\top)$  (primal feasibility)
- $y \in \mathcal{N}(E)$  (dual feasibility)

Static interconnect



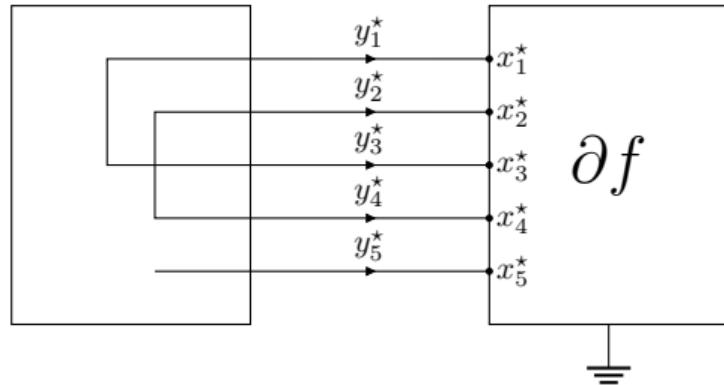
## Circuit interpretation: KKT conditions

$y \in \partial f(x)$  (nonlinear resistor)

$x \in \mathcal{R}(E^T)$  (KVL)

$y \in \mathcal{N}(E)$  (KCL)

Static interconnect



## Circuit interpretation: Dynamic interconnect

$$y(t) \in \partial f(x(t)) \quad (\text{nonlinear resistor})$$

$$v(t) = A^\top \begin{bmatrix} x(t) \\ e(t) \end{bmatrix} \quad (\text{KVL})$$

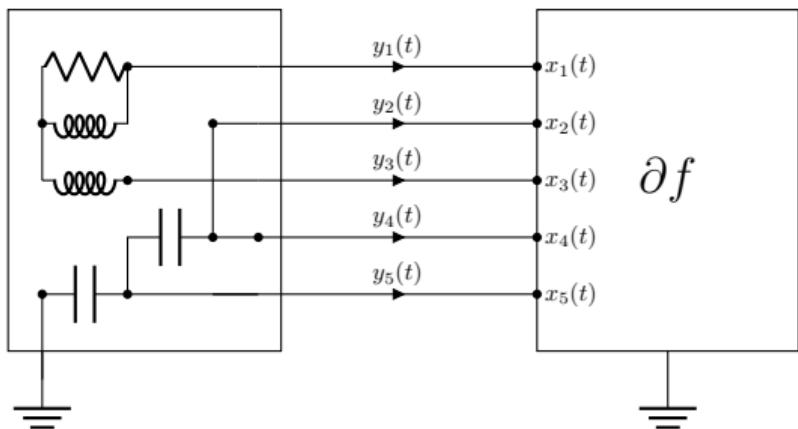
$$Ai(t) = \begin{bmatrix} -y(t) \\ 0 \end{bmatrix} \quad (\text{KCL})$$

$$v_{\mathcal{R}}(t) = D_{\mathcal{R}} i_{\mathcal{R}}(t) \quad (\text{resistor})$$

$$v_{\mathcal{L}}(t) = D_{\mathcal{L}} \frac{d}{dt} i_{\mathcal{L}}(t) \quad (\text{inductor})$$

$$i_C(t) = D_C \frac{d}{dt} v_C(t) \quad (\text{capacitor})$$

Dynamic interconnect



## Circuits for classical algorithms: DRS

- V-I relations

$$x_1 = \text{prox}_{Rg}(x_2 + Ri_{\mathcal{L}})$$

$$x_2 = \text{prox}_{Rf}(x_1 - Ri_{\mathcal{L}})$$

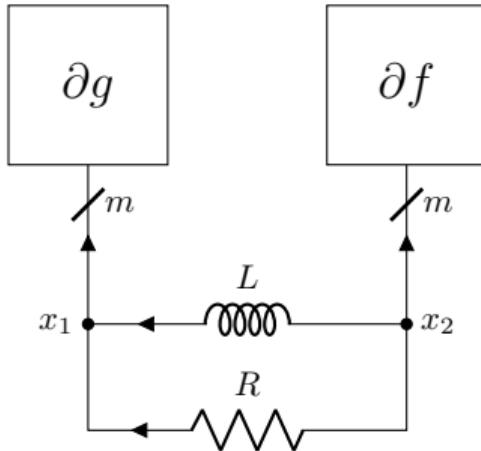
$$\frac{d}{dt}i_{\mathcal{L}} = \frac{1}{L}(x_2 - x_1)$$

- Douglas–Rachford splitting

$$x_1^{k+1} = \text{prox}_{Rg}(x_2^k + Ri_{\mathcal{L}}^k)$$

$$x_2^{k+1} = \text{prox}_{Rf}(x_1^{k+1} - Ri_{\mathcal{L}}^k)$$

$$i_{\mathcal{L}}^{k+1} = i_{\mathcal{L}}^k + \frac{h}{L}(x_2^{k+1} - x_1^{k+1})$$



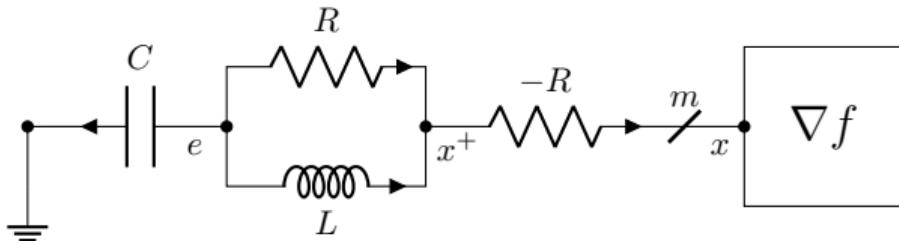
## Circuits for classical algorithms: Nesterov acceleration

- V-I relations

$$\begin{aligned}\frac{d}{dt} i_{\mathcal{L}} &= D_{\mathcal{L}}^{-1}(v_{\mathcal{C}} - x^+) \\ \frac{d}{dt} v_{\mathcal{C}} &= -D_{\mathcal{C}}^{-1} \nabla f(x).\end{aligned}$$

- Nesterov acceleration

$$\frac{d^2}{dt^2}x + 2\sqrt{\mu}\frac{d}{dt}x + \sqrt{s}\frac{d}{dt}\nabla f(x) + (1 + \sqrt{\mu s})\nabla f(x) = 0$$



## Circuits for classical algorithms: Proximal gradient

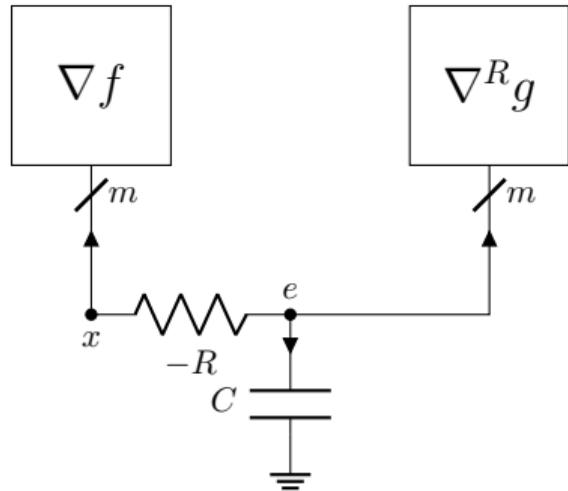
- V-I relations

$$i_C = -\nabla f(x) - \nabla^R g(e)$$

$$v_C = x - R\nabla f(x)$$

- Proximal gradient method

$$x^{k+1} = \text{prox}_{Rg}(I - R\nabla f)(x^k)$$



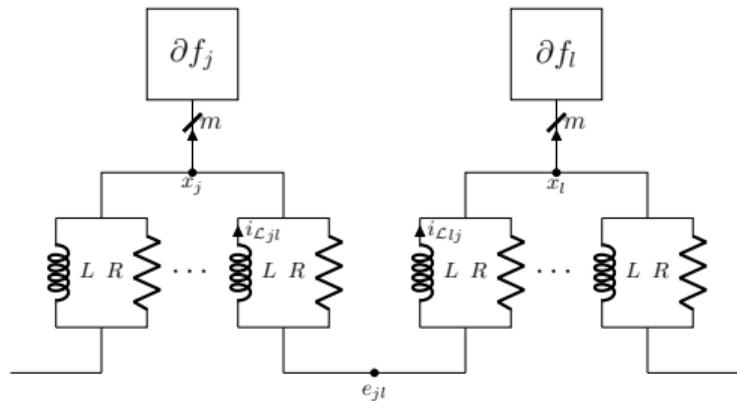
## Circuits for classical algorithms: DADMM

- V-I relations

$$\begin{aligned}
 x_j &= \text{prox}_{(R/|\Gamma_j|)f_j} \left( \frac{1}{|\Gamma_j|} \sum_{l \in \Gamma_j} (R i_{\mathcal{L}jl} + e_{jl}) \right) \\
 e_{jl} &= \frac{1}{2}(x_j + x_l) \\
 \frac{d}{dt} i_{\mathcal{L}jl} &= \frac{1}{L} (e_{jl} - x_j)
 \end{aligned}$$

- Decentralized ADMM

$$\begin{aligned}
 x_j^{k+1} &= \text{prox}_{(R/|\Gamma_j|)f_j} \left( \frac{1}{|\Gamma_j|} \sum_{l \in \Gamma_j} (R i_{\mathcal{L}jl}^k + e_{jl}^k) \right) \\
 e_{jl}^{k+1} &= \frac{1}{2}(x_j^{k+1} + x_l^{k+1}) \\
 i_{\mathcal{L}jl}^{k+1} &= i_{\mathcal{L}jl}^k + \frac{1}{R} (e_{jl}^{k+1} - x_j^{k+1})
 \end{aligned}$$

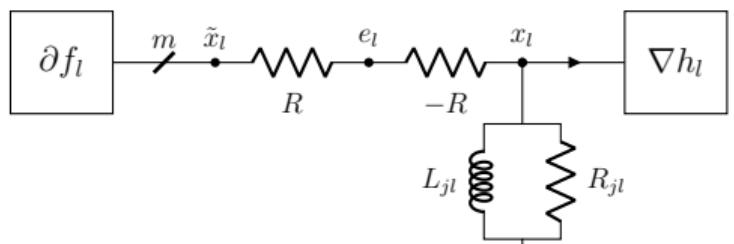


## Circuits for classical algorithms: PG-EXTRA

- V-I relations

$$x_j = \text{prox}_{Rf_j} \left( \sum_{l=1}^N W_{jl} x_l - R \nabla h_j(x_j) - w_j \right)$$

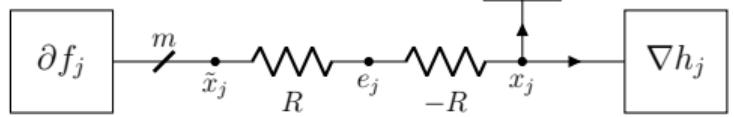
$$\frac{d}{dt} w_j = x_j - \sum_{l=1}^N W_{jl} x_l$$



- PG-EXTRA

$$x^{k+1} = \text{prox}_{Rf} \left( Wx^k - R \nabla h(x^k) - w^k \right)$$

$$w^{k+1} = w^k + \frac{1}{2}(I - W)x^k$$



## Energy dissipation

- in continuous time, energy dissipation leads to convergence (Thm 2.2)
  - $\mathcal{E}(t) = \frac{1}{2} \|v_C(t) - v_C^*\|_{D_C}^2 + \frac{1}{2} \|i_L(t) - i_L^*\|_{D_L}^2$
  - $\frac{d}{dt} \mathcal{E} \leq -\langle x(t) - x^*, y(t) - y^* \rangle \leq 0$
  - $\lim_{t \rightarrow \infty} x(t) = x^*$
- not every discretization leads to a convergent algorithm

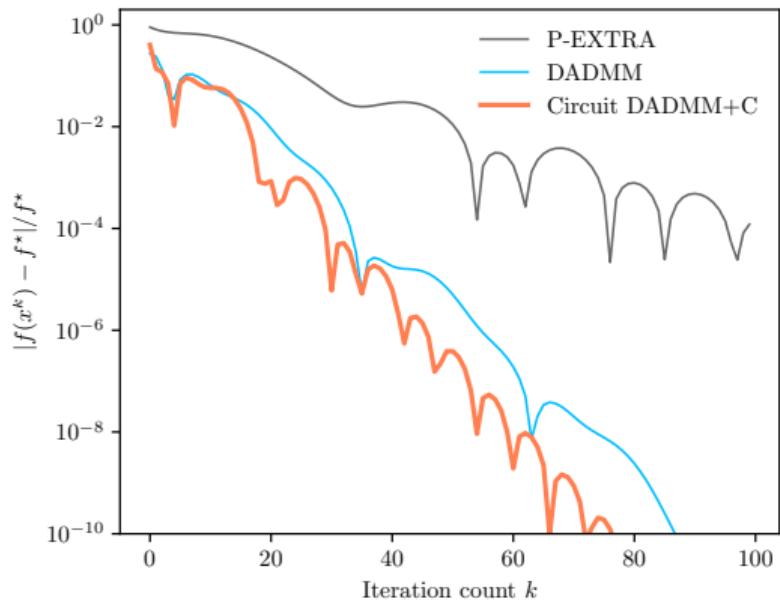
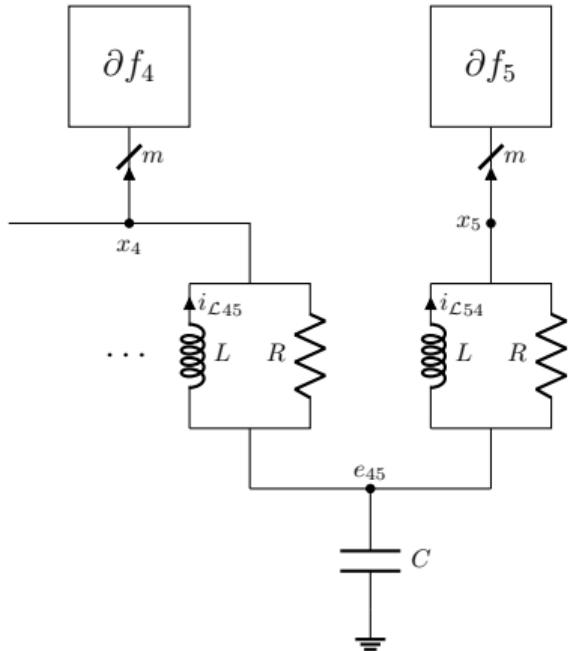
## Automatic discretization

- find discretization preserving the proof structure (Lemma 4.1)
  - $\mathcal{E}_k = \frac{1}{2} \|v_{\mathcal{C}}^k - v_{\mathcal{C}}^*\|_{D_{\mathcal{C}}}^2 + \frac{1}{2} \|i_{\mathcal{L}}^k - i_{\mathcal{L}}^*\|_{D_{\mathcal{L}}}^2$
  - $\mathcal{E}_{k+1} - \mathcal{E}_k + \eta \langle x^k - x^*, y^k - y^* \rangle \leq 0$  for some  $\eta > 0$
  - $\lim_{k \rightarrow \infty} x^k = x^*$
- automate using computer-assisted proof framework PEP
  - open-source package ciropt:  
[https://github.com/cvxgrp/optimization\\_via\\_circuits](https://github.com/cvxgrp/optimization_via_circuits)

## Previous discretizations

- previous discretization studies can be divided into two categories
  - special rules tailored to the specific dynamics
  - apply standard discretization schemes or their variants
- our discretization methodology is novel
  - aim to find parameters that preserve the proof structure
  - find such parameters automatically by leveraging PEP

## Numerical results: DADMM+C



## Contributions

- introduce a framework for designing optimization algorithms via RLC circuits
  - design dynamic circuit that converges to the solution
  - discretize to obtain convergent algorithm
- electric circuits for standard methods
  - Nesterov acceleration, proximal point method, prox-gradient, primal decomposition, dual decomposition, DYS, DRS, decentralized gradient descent, diffusion, DADMM and PG-EXTRA
- convergence proof of circuit dynamics based on energy dissipation
- PEP-based automated discretization that preserves proof structure
  - open-source package `ciropt`  
[https://github.com/cvxgrp/optimization\\_via\\_circuits](https://github.com/cvxgrp/optimization_via_circuits)