

Biogeography-Based Optimization of Power Flow



All authors should
be listed on the first
page, just like in
your paper

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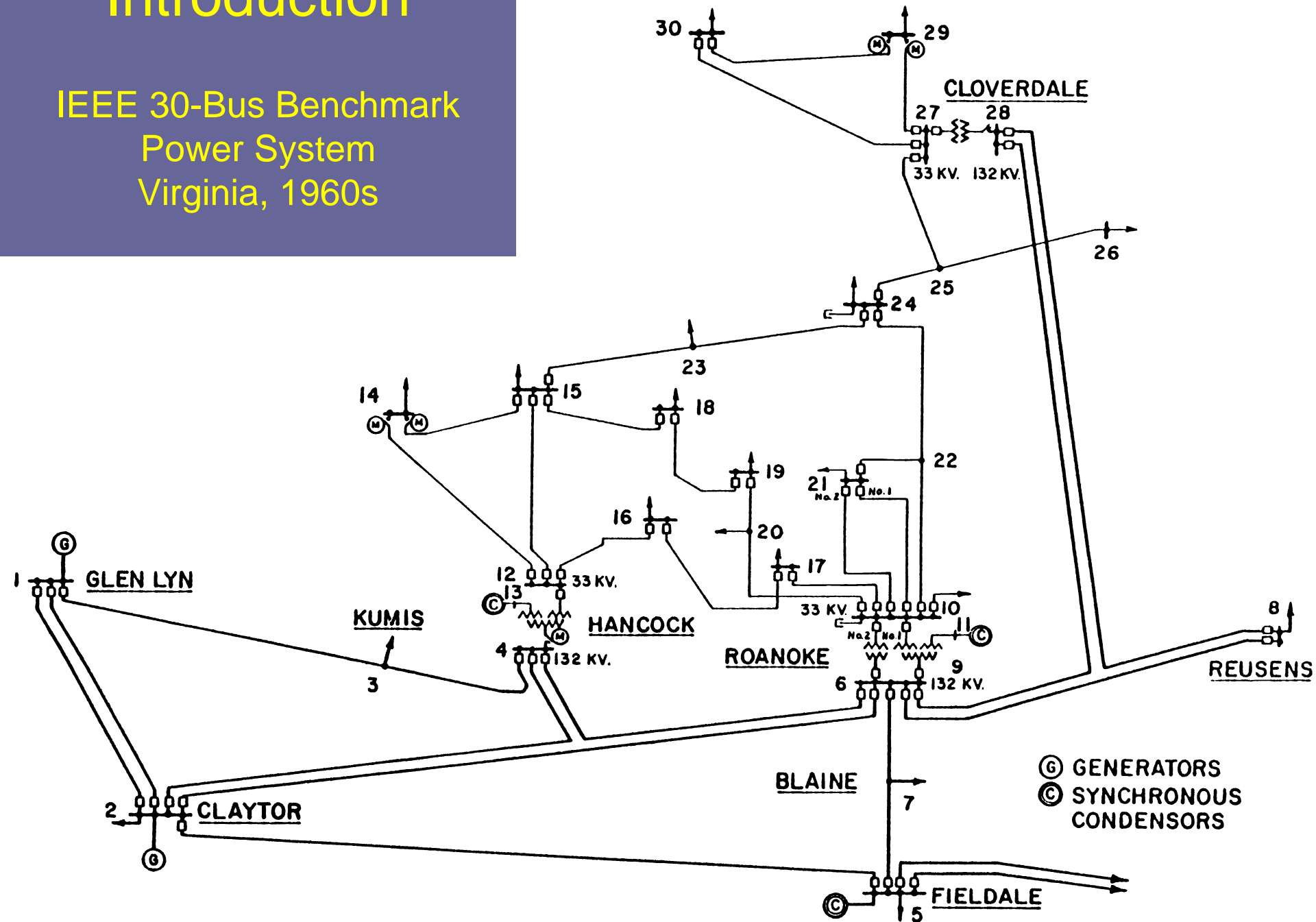
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Outline

- I. Introduction
- II. Power Flow Problem
- III. Simulation Results
- IV. Conclusion

Introduction

IEEE 30-Bus Benchmark
Power System
Virginia, 1960s

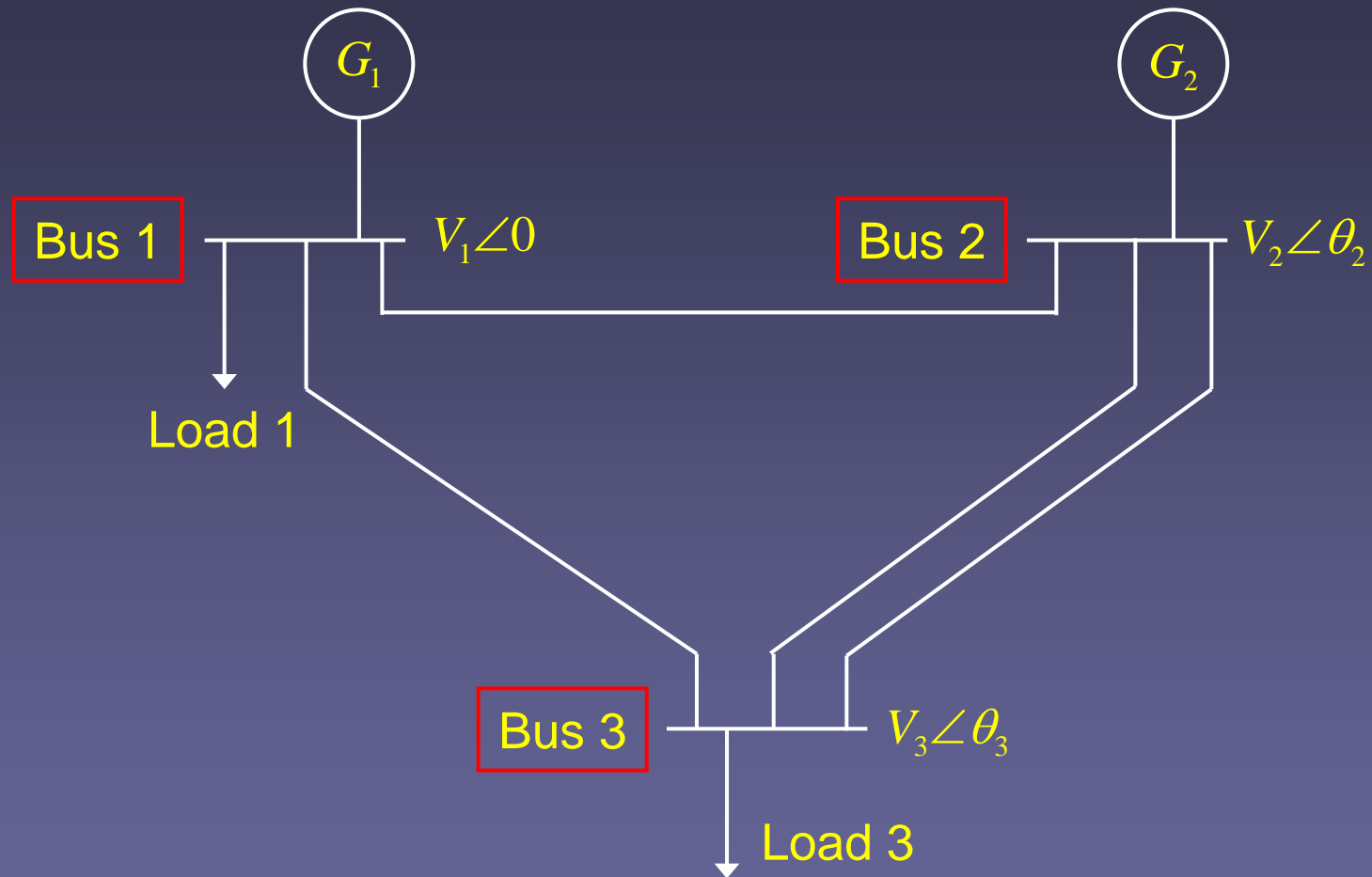


Introduction

Power System Design Problems

- Large number of variables (5,000) and constraints (10,000)
- Nonlinear and non-differentiable
- Multiple local minima
- Numerical methods unreliable under high loads and faults

Power Flow Problem



Power Flow Problem

30-Bus System

Bus voltage magnitude and phase angle vectors:

$$\mathbf{V} = (V_1, V_2, \dots, V_{30}), \quad \boldsymbol{\theta} = (\theta_1, \theta_2, \dots, \theta_{30})$$

60 unknowns

Bus i
2 Constraint
Eqs

$$P_i(\mathbf{V}, \boldsymbol{\theta}) = V_i \sum_{j \in N_b} V_j (G_{ij} \cos(\theta_{ij}) - B_{ij} \sin(\theta_{ij}))$$
$$Q_i(\mathbf{V}, \boldsymbol{\theta}) = V_i \sum_{j \in N_b} V_j (G_{ij} \sin(\theta_{ij}) - B_{ij} \cos(\theta_{ij}))$$

30x2 = 60
Constraint
Eqs

Power Flow Problem

Cost Function = Power / Voltage Mismatch (Ideally = 0)

$$\Delta = |\text{Estimated} - \text{Specified}|$$



Numerically Heuristically Design

$$\Delta P_i = |P_i^{\text{est}} - P_i^{\text{sp}}|$$

$$\Delta Q_i = |Q_i^{\text{est}} - Q_i^{\text{sp}}|$$

$$\Delta V_i = |V_i^{\text{est}} - V_i^{\text{sp}}|$$

$$\min_{\mathbf{V}, \boldsymbol{\theta}} \left\{ \sum_{i \in \text{Gen}} \Delta P_i^2(\mathbf{V}, \boldsymbol{\theta}) + \lambda_Q \sum_{i \in \text{PQ}} \Delta Q_i^2(\mathbf{V}, \boldsymbol{\theta}) + \lambda_V \sum_{i \in \text{PV}} \Delta V_i^2(\mathbf{V}) \right\}$$

Simulation

IEEE 30-Bus

Each Chromosome/Island - 60 genes

$(V_1, V_2, \dots, V_{30}, \theta_1, \theta_2, \dots, \theta_{30})$

$V_1, V_2, V_5, V_8, V_{13}, \theta_1$ --- Fixed

Population = 100

Generations = 100

Monte Carlo Trials = 1000

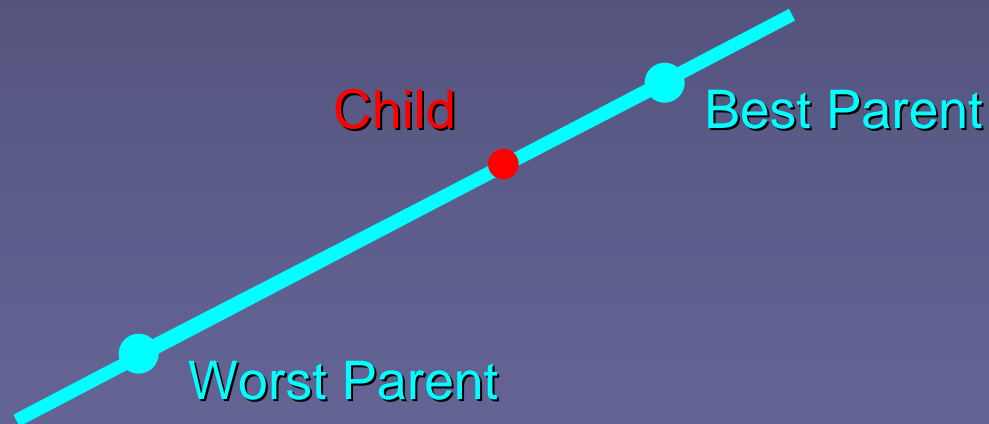
Probability of Mutation = 0.001

Simulation

Heuristic Crossover for GA

$$\alpha \sim U(0,1)$$

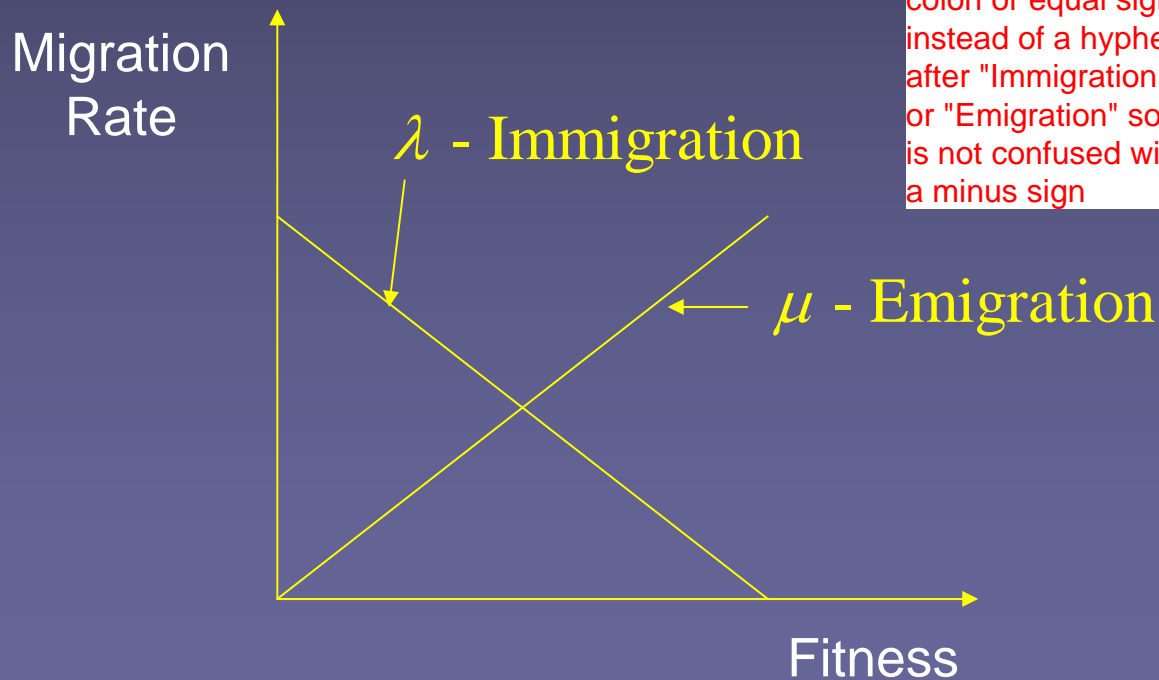
$$\text{Child} = \text{Best Parent} + \alpha(\text{Best Parent} - \text{Worst Parent})$$



Simulation

Migration Scheme for BBO

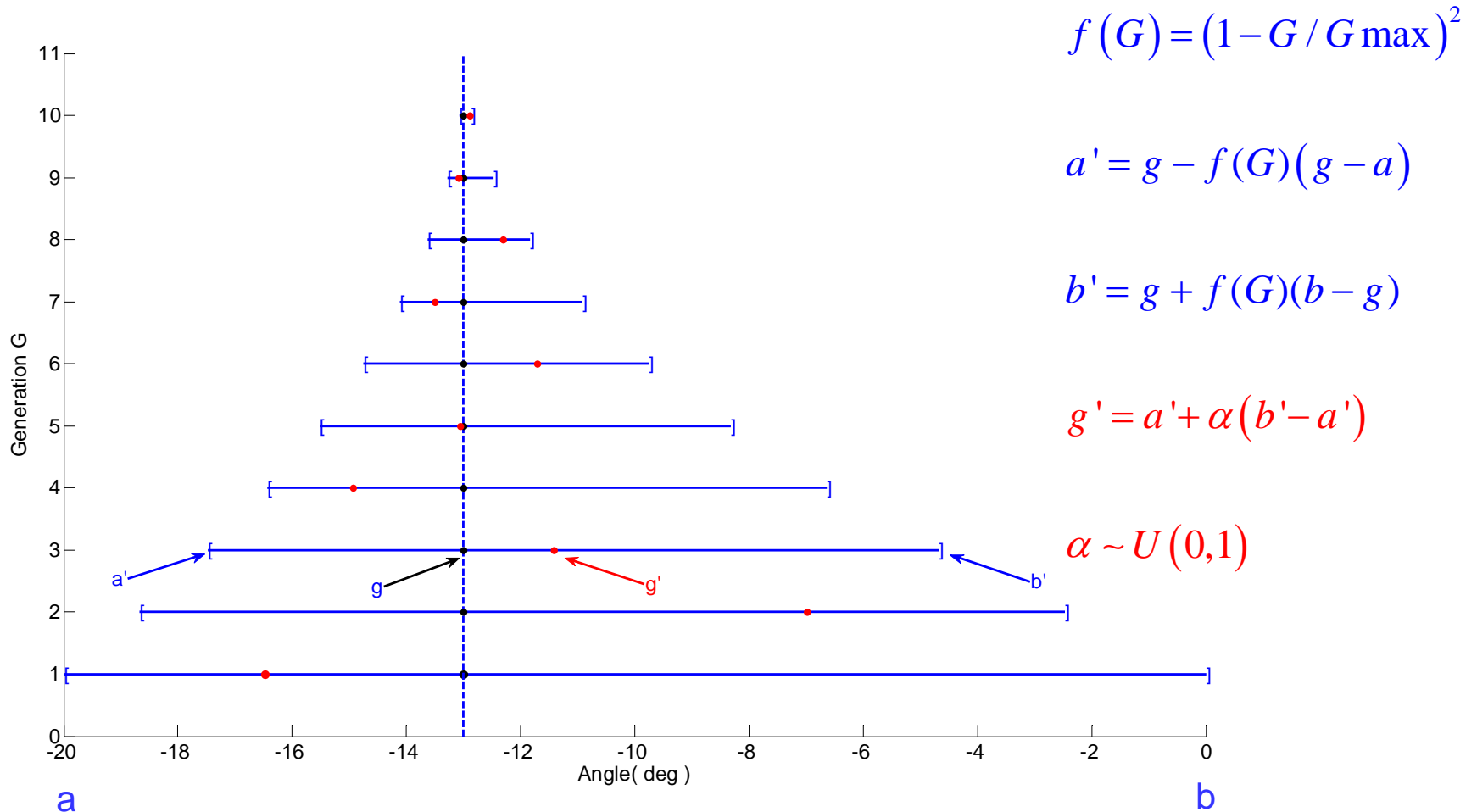
Immigration and Emigration:
Probabilistically determined by lambda, mu, and fitness



more clear to use a colon or equal sign instead of a hyphen after "Immigration" or "Emigration" so it is not confused with a minus sign

Simulation

Non-Uniform Mutation for GA and BBO

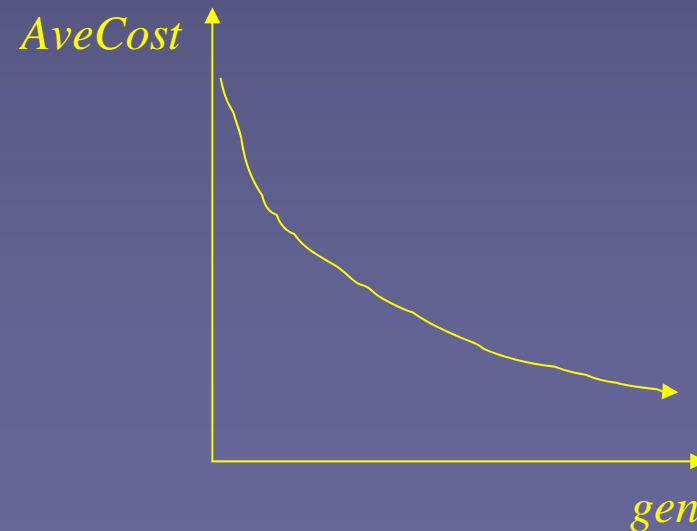
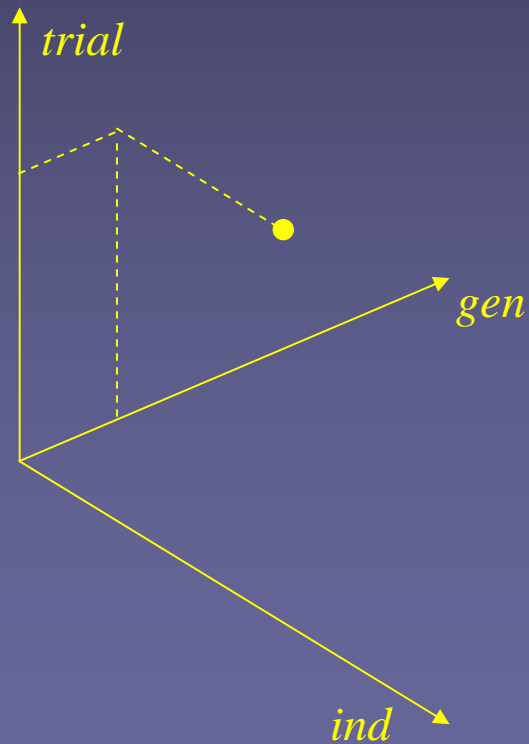


Simulation

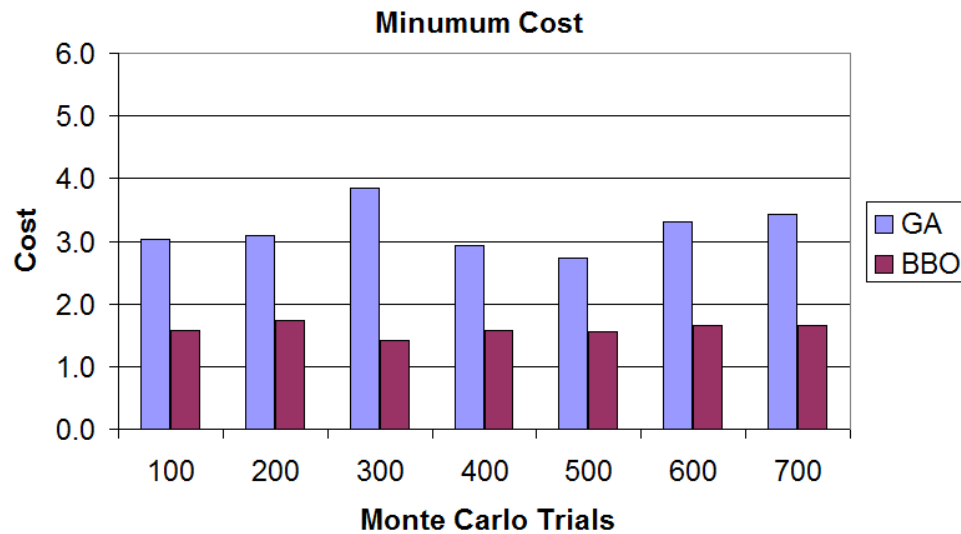
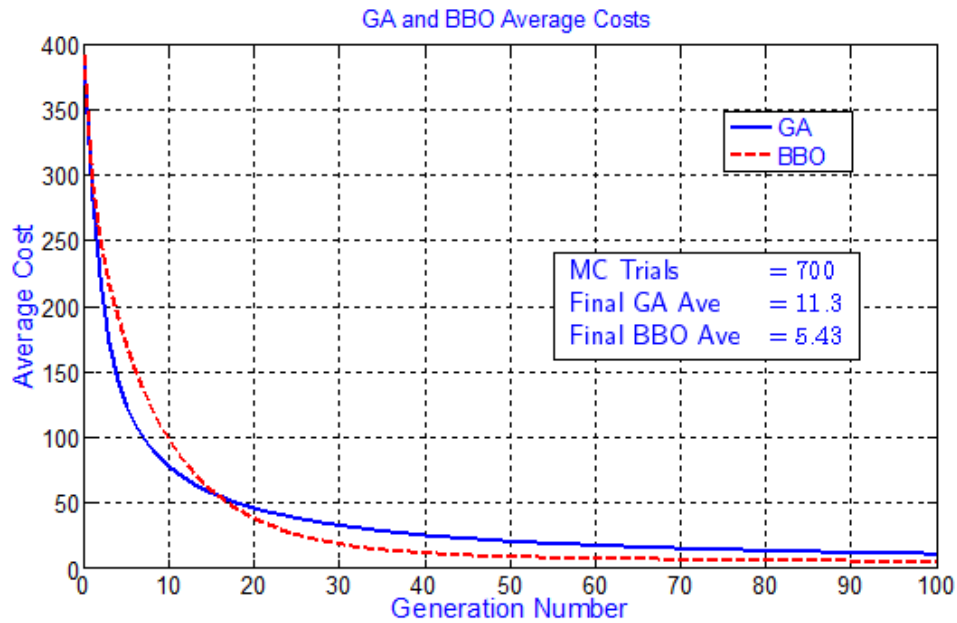
GA and BBO Performance

$$\text{cost} = \text{cost}(\text{individual}, \text{generation}, \text{trial})$$

$$\text{AveCost}(\text{gen}) = \text{ave}_{\text{trials}} \left(\text{ave}_{\text{ind}} [\text{cost}(\text{ind}, \text{gen}, \text{trial})] \right)$$



Simulation



Conclusion

- BBO performs better than GA on Power Flow problem
- First application of BBO to engineering problem
- Not restricted to differentiable system equations
- Need statistical significance of result
- Need to apply to more difficult Optimal Power Flow problem

Thank you !