

Primal-Dual Interior-Point Methods for Thermodynamic Phase Calculations

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Motivations

- Aerosol particles have effects on human health, visibility reduction, acid rain, alteration of the earth's radiation balance, cloud formation, climate warming.
- **Modeling** of the physical and chemical state of atmospheric organic aerosol particles.
- **Thermodynamics** of a particle (global minimization of the Gibbs free energy) treated with a **primal-dual interior-point method**
- **Gas-particle partitioning** modeled by an additional term in the objective function and a sequential quadratic programming approach to decouple the different scales.



Phase Equilibrium

$$\begin{aligned} \min_{\mathbf{n}_\alpha} \quad & \sum_{\alpha=1}^P g(\mathbf{n}_\alpha) \\ \text{s. t.} \quad & \sum_{\alpha=1}^P \mathbf{n}_\alpha = \mathbf{b}, \\ & \mathbf{n}_\alpha \geq 0, \quad \alpha = 1, \dots, P. \end{aligned}$$

- \mathbf{n}_α is the chemical concentration in the phase α .
- Split $\mathbf{n}_\alpha = y_\alpha \mathbf{x}_\alpha$, where $\mathbf{x}_\alpha \in \mathbb{R}^N$ virtual mole-fraction compositions and $y_\alpha \in \mathbb{R}$ total amount of moles in each phase.
- The number of phases P existing at the equilibrium is not known a priori, but a result of the equilibrium computation.
- The function g is homogeneous of degree one.



Phase Equilibrium

$$\begin{aligned} \min_{y_\alpha \mathbf{x}_\alpha} \quad & \sum_{\alpha=1}^P y_\alpha g(\mathbf{x}_\alpha) \\ \text{s. t.} \quad & \sum_{\alpha=1}^P y_\alpha \mathbf{x}_\alpha = \mathbf{b}, \\ & y_\alpha \geq 0, \quad \alpha = 1, \dots, P. \\ & \mathbf{e}^T \mathbf{x}_\alpha = 1, \quad \mathbf{x}_\alpha > 0, \quad \alpha = 1, \dots, N + 1. \end{aligned}$$

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- The number of phases P existing at the equilibrium is not known a priori, but a result of the equilibrium computation.
- Projection on the equality constraints $\mathbf{e}^T \mathbf{x}_\alpha = 1$.



Global Optimization Problem

$$\begin{aligned} \min_{y_\alpha, \mathbf{z}_\alpha} \quad & \sum_{\alpha=1}^P y_\alpha f(\mathbf{z}_\alpha) \\ \text{s. t.} \quad & \sum_{\alpha=1}^P y_\alpha \mathbf{z}_\alpha = \mathbf{b}, \quad \sum_{\alpha=1}^P y_\alpha = 1, \\ & y_\alpha \geq 0, \quad \alpha = 1, \dots, P, \\ & \mathbf{z}_\alpha \in \text{int}(\Delta_N), \quad \alpha = 1, \dots, P. \end{aligned}$$

- $\mathbf{z}_\alpha \in \mathbb{R}^N$ virtual mole-fraction compositions.
- $y_\alpha \in \mathbb{R}$ total amount of moles in each phase. It allows to track of the existence of the phase α .
- $\Delta_N = \text{conv}\{\mathbf{0}, \mathbf{e}_1, \dots, \mathbf{e}_N\}$ is a N -simplex in \mathbb{R}^N .



Convex Hull

- Convex hull of a function $f : \mathbb{R}^N \mapsto \mathbb{R}$: $\text{conv} f$ is the greatest convex function majored by f .

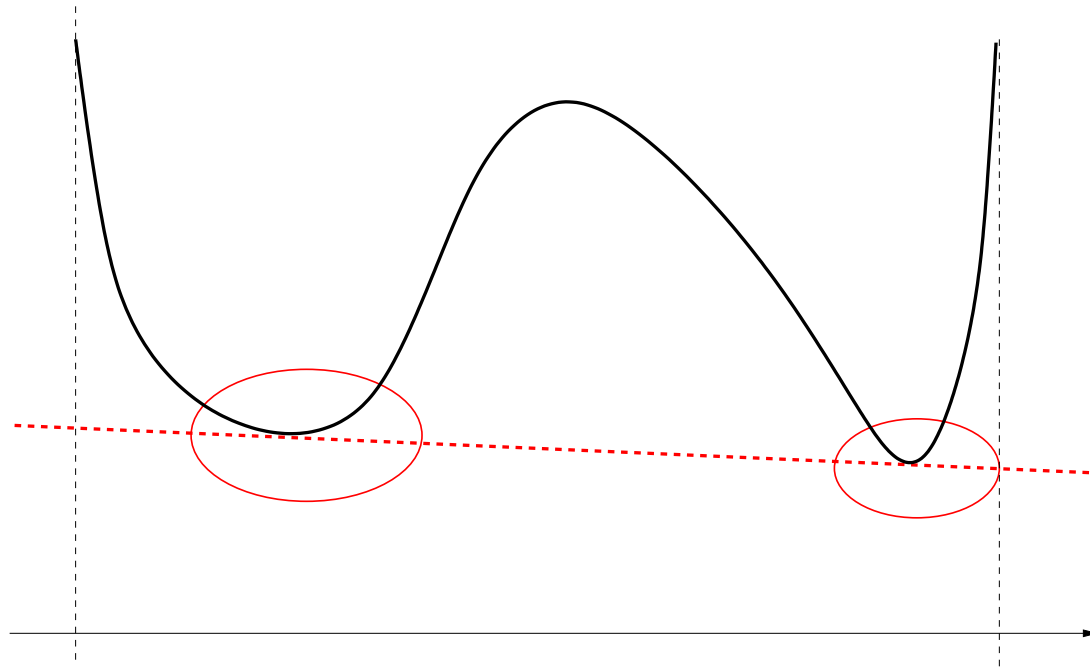
$$(\text{conv} f)(\mathbf{b}) = \inf \left\{ \sum_{\alpha=1}^P y_{\alpha} f(\mathbf{z}_{\alpha}) \mid \sum_{\alpha=1}^P y_{\alpha} \mathbf{z}_{\alpha} = \mathbf{b}, y_{\alpha} \geq 0, \sum_{\alpha=1}^P y_{\alpha} = 1 \right\}.$$

- **Carathéodory's Theorem:** For a set $C \neq \emptyset$ in \mathbb{R}^N , every point of $\text{conv}(C)$ belongs to some simplex with vertices in C and thus can be expressed as a convex combination of $N + 1$ points of C (not necessarily different). When C is connected, N points suffice.

$$(\text{conv} f)(\mathbf{b}) = \inf \left\{ \sum_{\alpha=1}^{N+1} y_{\alpha} f(\mathbf{z}_{\alpha}) \mid \sum_{\alpha=1}^{N+1} y_{\alpha} \mathbf{z}_{\alpha} = \mathbf{b}, y_{\alpha} \geq 0, \sum_{\alpha=1}^{N+1} y_{\alpha} = 1 \right\}.$$



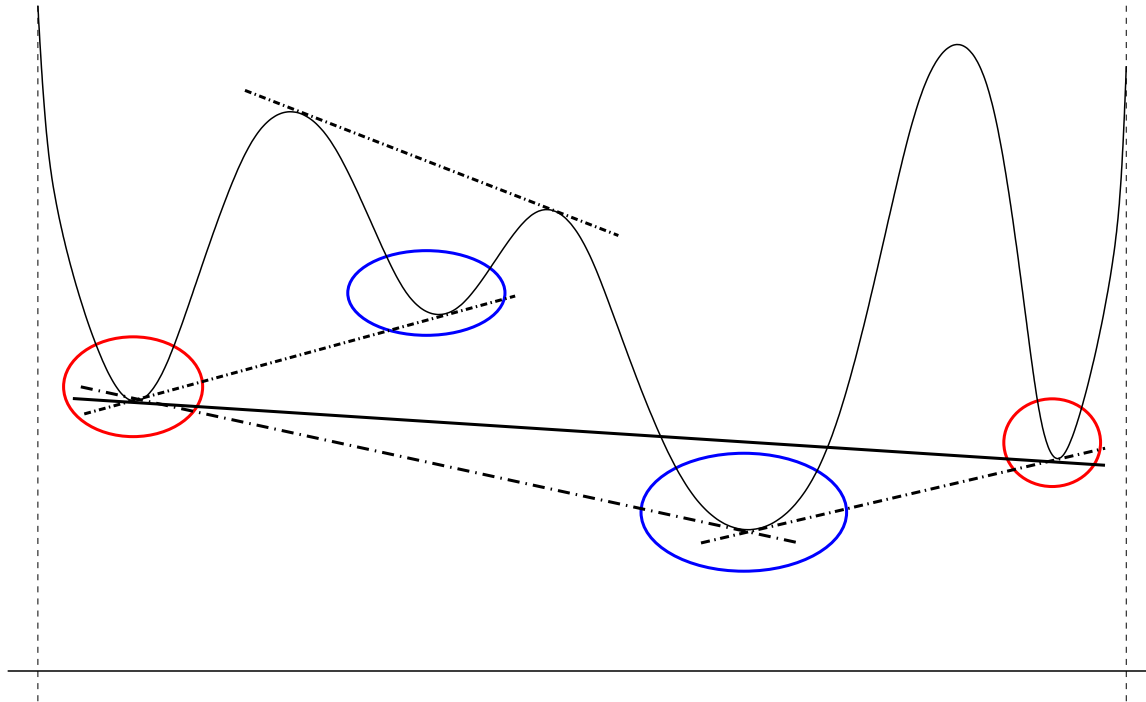
Properties of the Energy Function



- Regularity of the energy function: $f \in C^\infty(\text{int}\Delta_N) \cap C^0(\overline{\Delta_N})$.
- $\forall \mathbf{x}_0 \in \partial\Delta_N, \quad \forall w \text{ proper } \lim_{\mathbf{x} \rightarrow \mathbf{x}_0} \nabla f \cdot w = -\infty$.
- Each vertex of the N -simplex Δ_N (pure components) belongs to a connected convex region of f .
- Determination of a tangent plane to the energy function.



Properties of the Energy Function



- No *isolated connected convex regions* (for global optimum).
- The function f is strictly convex in a neighborhood of a solution (for uniqueness).

Some Results on the Convex Hull

- Application of results from [Rabier, Griewank (1992)].
- The feed \mathbf{b} can be decomposed into a **stable phase splitting**:

$$\mathbf{b} = \sum_{\alpha=1}^P y_{\alpha} \mathbf{z}_{\alpha}, \quad \mathbf{z}_{\alpha} \in \text{int}(\Delta_N), \quad \alpha = 1, \dots, P.$$

The convex hull is continuously differentiable on $\text{int}(\Delta_N)$.

- The following relations hold for all $(k, m) = 1, \dots, P$:

$$\begin{aligned} \nabla f(\mathbf{z}_k) &= \nabla f(\mathbf{z}_m) \\ f(\mathbf{z}_k) - \nabla f(\mathbf{z}_k) \cdot \mathbf{z}_k &= f(\mathbf{z}_m) - \nabla f(\mathbf{z}_m) \cdot \mathbf{z}_m \end{aligned}$$

Moreover the **tangent plane** $\Theta(\mathbf{z}) = f(\mathbf{z}_1) + \nabla f(\mathbf{z}_1) \cdot (\mathbf{z} - \mathbf{z}_1)$ is tangent to f at all active vertices and always below the graph of f .

Acknowledgments: A. Anantharam, Ecole Polytechnique de Paris



An Interior-Point Method

- Introduction of a log-barrier penalty function to relax the inequality constraints.

$$\begin{aligned} \min_{y_\alpha, \mathbf{z}_\alpha} \quad & \sum_{\alpha=1}^{N+1} y_\alpha f(\mathbf{z}_\alpha) \\ \text{s. t.} \quad & \sum_{\alpha=1}^{N+1} y_\alpha \mathbf{z}_\alpha = \mathbf{b}, \quad \sum_{\alpha=1}^{N+1} y_\alpha = 1, \\ & \mathbf{z}_\alpha \in \text{int}(\Delta_N), \quad \alpha = 1, \dots, N + 1. \\ & y_\alpha \geq 0, \quad \alpha = 1, \dots, N + 1, \end{aligned}$$

- ν is a penalty parameter, which tends to zero.
- The phase α disappears when $y_\alpha \simeq 0$.
- Initial guess with $N + 1$ vertices and deletion of extra vertices until convergence is reached.



An Interior-Point Method

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$$\begin{aligned} \min_{y_\alpha, \mathbf{z}_\alpha} \quad & \sum_{\alpha=1}^{N+1} y_\alpha f(\mathbf{z}_\alpha) - \nu \sum_{\alpha=1}^{N+1} \ln(y_\alpha) \\ \text{s. t.} \quad & \sum_{\alpha=1}^{N+1} y_\alpha \mathbf{z}_\alpha = \mathbf{b}, \quad \sum_{\alpha=1}^{N+1} y_\alpha = 1, \\ & \mathbf{z}_\alpha \in \text{int}(\Delta_N), \quad \alpha = 1, \dots, N + 1. \end{aligned}$$

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Karush-Kuhn-Tucker Conditions

- KKT conditions (stationary points of the Lagrangian):

$$\begin{aligned}
 y_\alpha (\nabla f(\mathbf{z}_\alpha) + \boldsymbol{\eta}) &= 0, & \alpha = 1, \dots, N + 1, \\
 f(\mathbf{z}_\alpha) + \boldsymbol{\eta}^T \mathbf{z}_\alpha - \frac{\nu}{y_\alpha} + \theta &= 0, & \alpha = 1, \dots, N + 1, \\
 \sum_{\alpha=1}^{N+1} y_\alpha \mathbf{x}_\alpha - \mathbf{b} &= 0, & \sum_{\alpha=1}^{N+1} y_\alpha - 1 = 0.
 \end{aligned}$$

- KKT System (Newton method):

$$\begin{pmatrix}
 y_\alpha \nabla^2 f(\mathbf{z}_\alpha) & \nabla f(\mathbf{z}_\alpha) + \boldsymbol{\eta} & y_\alpha & 0 \\
 (\nabla f(\mathbf{z}_\alpha) + \boldsymbol{\eta})^T & \frac{\nu}{y_\alpha^2} & \mathbf{z}_\alpha^T & \mathbf{e} \\
 y_\alpha^T & \mathbf{z}_\alpha & 0 & 0 \\
 0 & \mathbf{e}^T & 0 & 0
 \end{pmatrix}
 \begin{pmatrix}
 \mathbf{p}_z \\
 \mathbf{p}_y \\
 \mathbf{p}_\eta \\
 \mathbf{p}_\theta
 \end{pmatrix}
 =
 \begin{pmatrix}
 \mathbf{b}_z \\
 \mathbf{b}_y \\
 \mathbf{b}_\eta \\
 \mathbf{b}_\theta
 \end{pmatrix}$$



Global Optimum vs Local Optimum

- $(y_\alpha, \mathbf{z}_\alpha)$ is a global minimum if and only if $(y_\alpha, \mathbf{z}_\alpha, \boldsymbol{\eta}, \theta)$ is a KKT point with $\mathbf{z}_\alpha \in \Delta_n$ such that $\nabla^2 f(\mathbf{z}_\alpha) > 0$ and $y_\alpha > 0$.
- A feasible point $(y_\alpha, \mathbf{z}_\alpha)$ is a global minimum if and only if it is a KKT point and

$$f(\mathbf{z}) \geq \Theta(\mathbf{z}) \quad \forall \mathbf{z} \in \Delta_n,$$

where $\Theta(\mathbf{z})$ is the hyperplane associated with the KKT point.



Computational Issues

- KKT systems for chemical systems are usually **ill-conditioned**.
- **Direct decomposition techniques** of the block-structured system (range-space + null-space).
- Control of the inertia of the matrices arising in the resolution.
- Numerically coupled with active sets procedure.

Assumptions:

- The Hessian $\nabla^2 f(\mathbf{z}_\alpha)$ has to remain positive definite (*second order conditions*)
- The iterates $\mathbf{z}_1, \mathbf{z}_2, \dots, \mathbf{z}_{N+1}$ remain linearly independent (*linear independent constraint qualification*).



Computational Issues

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Algorithm:

- Initialization of the primal and dual variables. Initialization of the penalty parameter ν .
- Resolution of the KKT system to obtain a descent direction.
- Update the penalty parameter ν and compute next iterate.

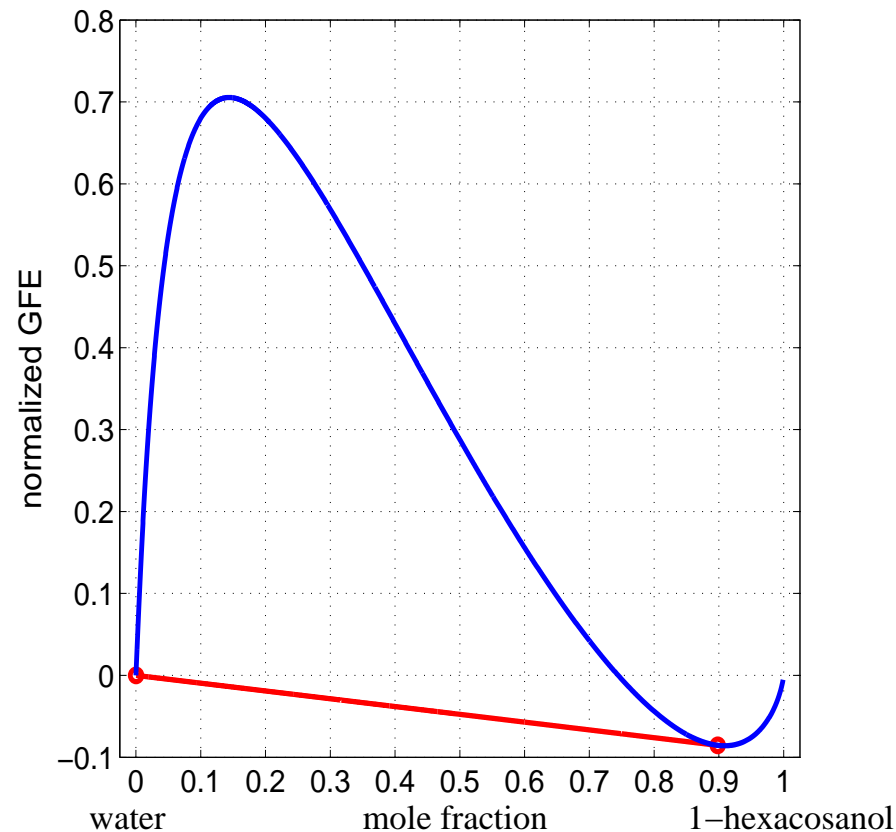
Key Issue:

- Initialization of \mathbf{z}_α in order to guarantee the second order conditions during the whole algorithm.



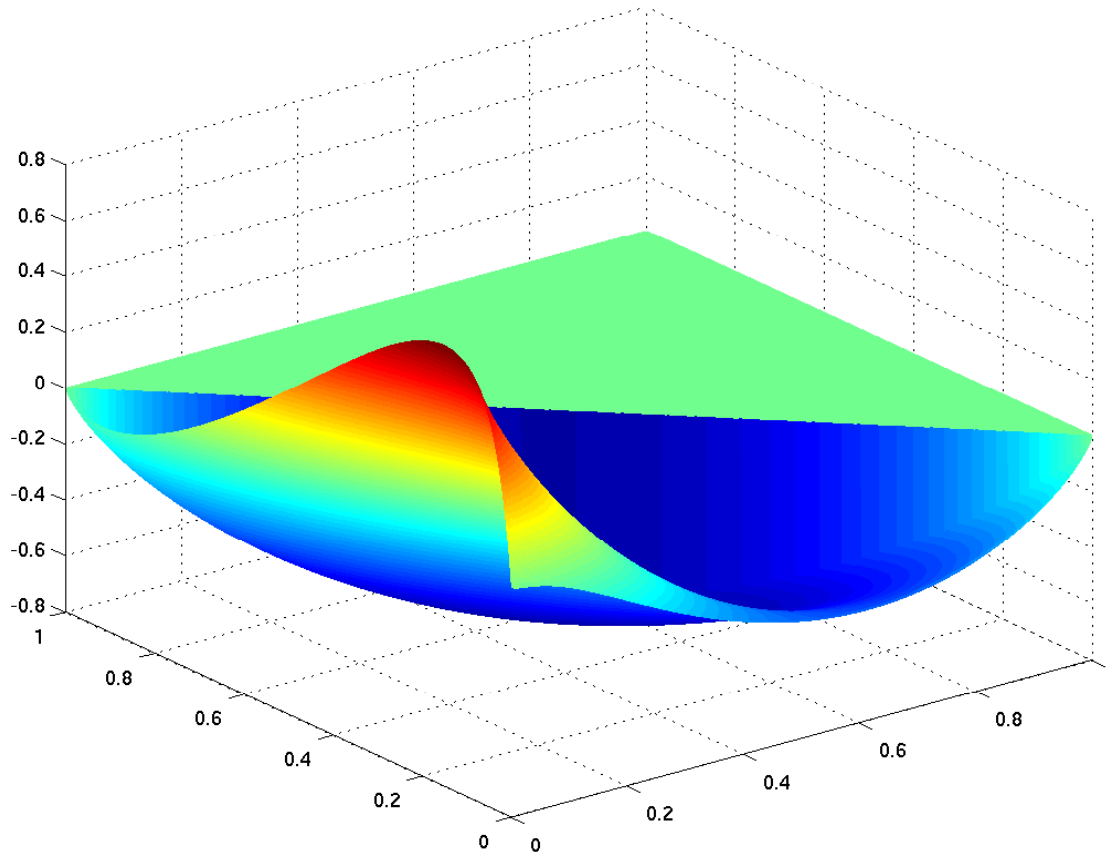
Results in One Dimension

- Energy is defined on the $(0, 1)$ segment. Both extremities of the segment are in a convex region by assumption.
- Convex Hull:



Results in Two Dimensions

- Non-convex energy Function:

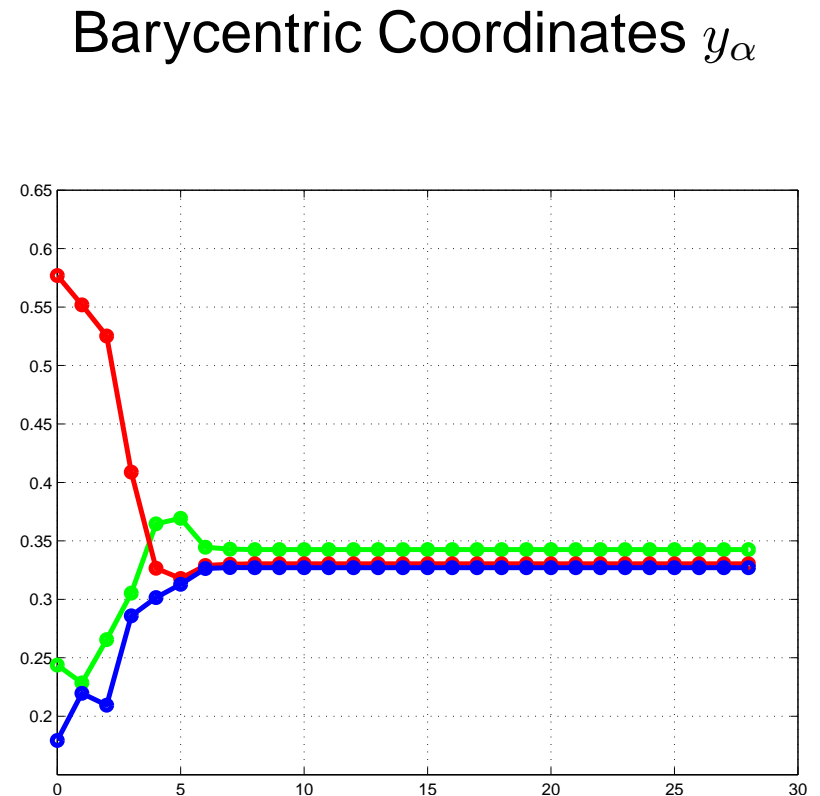
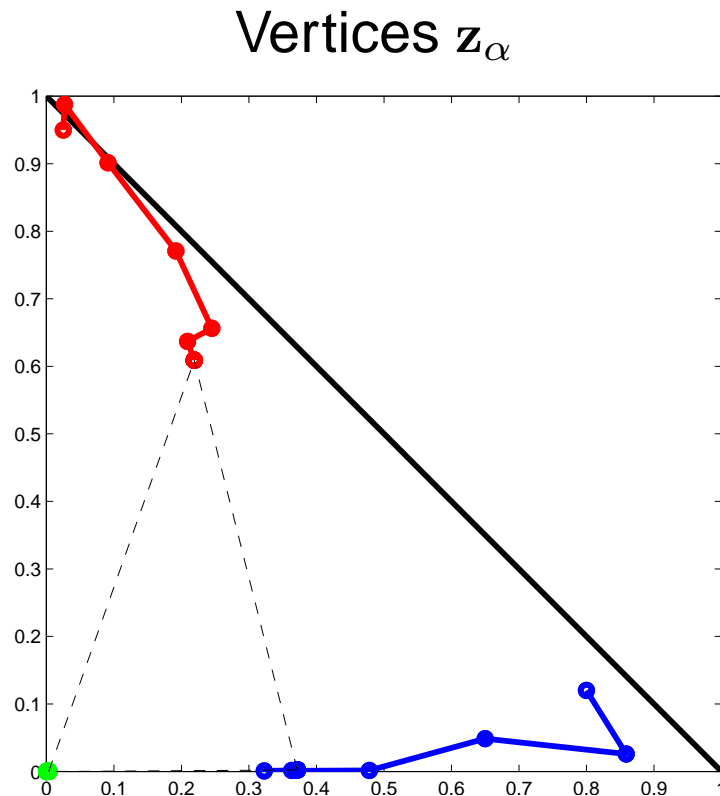


- Determination of a tangent plane.



Convergence of Phase Simplexes

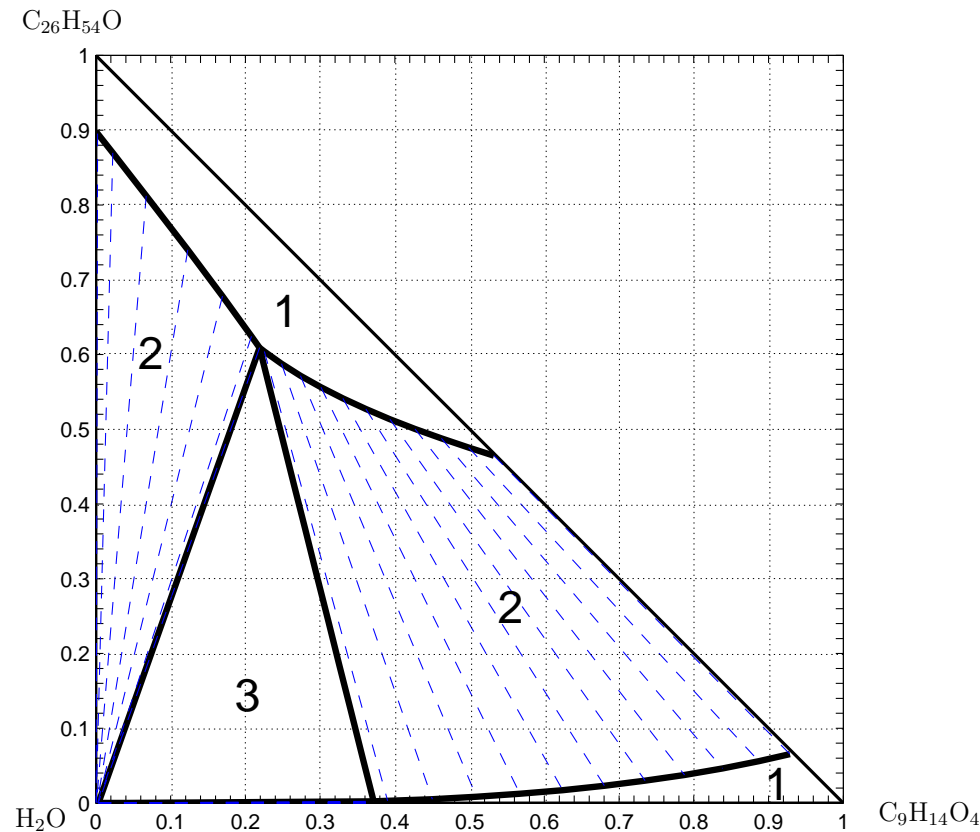
- Convergence towards three active vertices:



- Particular choice of initial guess (near the vertices) and numerical parameters to avoid local minima.



Convex Hull and Active Constraints



- Computational cost: 20 s. for $\frac{1}{2}100 \times 100$ grid points, 25 iterations in average (tolerance = 10^{-7}).



Acknowledgments: C. Landry, Graduate Student, EPFL, Switzerland

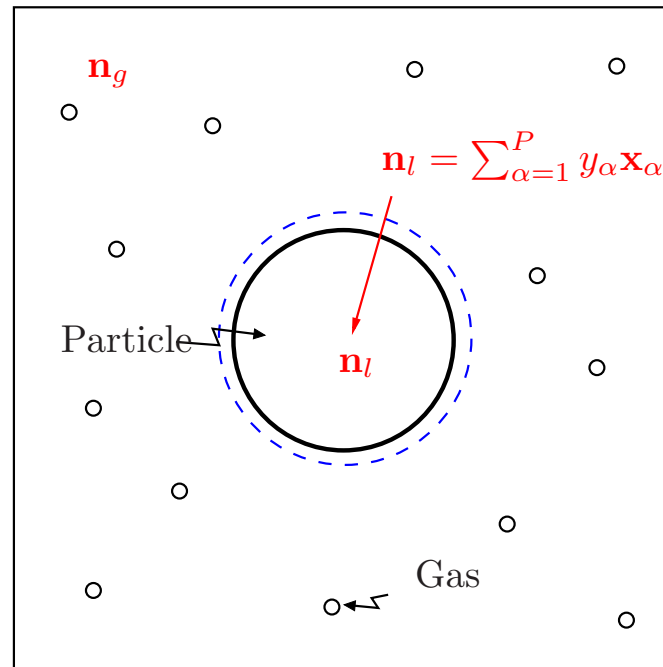
Results in Higher Dimensions

| Size $N + 1$ | # Iterations | CPU time |
|--------------|--------------|----------|
| 3 | 25 | 0.024 |
| 4 | 40 | 0.0369 |
| 18 | 41 | 0.1799 |



Gas-Particle Partitioning

- Single aerosol particle:



- Find the equilibrium state of the particle and the repartition between solid, liquid and gas phases.
- Find the gas-particle partitioning/thermodynamics of the system gas+particle.



Energy Model

$$\begin{aligned} \min_{y_\alpha, \mathbf{x}_\alpha, \mathbf{n}_g} \quad & \sum_{\alpha=1}^P y_\alpha g(\mathbf{x}_\alpha) + \mathbf{n}_g^T \boldsymbol{\mu}_g(\mathbf{n}_g) \\ \text{s. t.} \quad & \sum_{\alpha=1}^P y_\alpha \mathbf{x}_\alpha + \mathbf{n}_g = \tilde{\mathbf{b}}, \\ & y_\alpha \geq 0, \quad \mathbf{e}^T \mathbf{x}_\alpha = 1, \quad \mathbf{x}_\alpha > \mathbf{0}, \quad \alpha = 1, \dots, P, \\ & \mathbf{n}_g > \mathbf{0}. \end{aligned}$$

- \mathbf{n}_g are the concentrations in the gas, supposed to be positive.
- $\boldsymbol{\mu}_g = \boldsymbol{\mu}_g(\mathbf{n}_g) = \ln \left(\mathbf{n}_g \frac{RT}{p_{\text{vapor}}} \right)$ is chemical potentials vector.



Global Optimization Problem

- Projected optimization problem:

$$\begin{aligned} \min_{y_\alpha, \mathbf{z}_\alpha, \mathbf{n}_g} \quad & \sum_{\alpha=1}^P y_\alpha f(\mathbf{z}_\alpha) + \mathbf{n}_g^T \boldsymbol{\mu}_g(\mathbf{n}_g) \\ \text{s. t.} \quad & \sum_{\alpha=1}^P y_\alpha \mathbf{z}_\alpha + \mathcal{P} \mathbf{n}_g = \mathbf{b}, \quad \sum_{\alpha=1}^P y_\alpha + \mathbf{e}^T \mathbf{n}_g = \tilde{b}, \\ & y_\alpha \geq 0, \quad \alpha = 1, \dots, P, \\ & \mathbf{z}_\alpha \in \text{int}(\Delta_N), \quad \alpha = 1, \dots, P, \\ & \mathbf{n}_g > 0. \end{aligned}$$

where $\mathcal{P} : \mathbb{R}^{n+1} \rightarrow \mathbb{R}^n$ is the projection on $\mathbf{e}^T \mathbf{x} = 1$.

- **Result:** The extended problem admits a unique global minimizer (additional term is convex).



Interior-Point Method and KKT Conditions

- Interior-point method and first-order conditions:

$$\begin{aligned}\mu_g(\mathbf{n}_g) + \boldsymbol{\lambda} &= \mathbf{0}, \\ y_\alpha (\nabla f(\mathbf{z}_\alpha) + \boldsymbol{\eta}) &= \mathbf{0}, \quad \alpha = 1, \dots, P, \\ f(\mathbf{z}_\alpha) + \boldsymbol{\eta}^T \mathbf{z}_\alpha + \theta - \frac{\nu}{y_\alpha} &= 0, \quad \alpha = 1, \dots, P, \\ \sum_{\alpha=1}^P y_\alpha \mathbf{z}_\alpha + \mathcal{P} \mathbf{n}_g &= \mathbf{b}, \\ \sum_{\alpha=1}^P y_\alpha + \mathbf{e}^T \mathbf{n}_g &= \tilde{b}.\end{aligned}$$

- **Result:** $(y_\alpha, \mathbf{z}_\alpha, \mathbf{n}_g)$ is a global minimum if and only if $(y_\alpha, \mathbf{z}_\alpha, \mathbf{n}_g, \boldsymbol{\eta}, \theta)$ is a KKT point with $\mathbf{z}_\alpha \in \Delta_n$ such that $\nabla^2 f(\mathbf{z}_\alpha) > 0$ and $y_\alpha > 0$.
- KKT system is obtained by using one Newton iteration at each iteration of the interior-point method.



Initialization Techniques

- Geometric properties of the energy function f : the variables z_α are initialized in a neighborhood \mathcal{N} of the vertices of Δ_n such that the function f is convex on $\Delta_n \cap \mathcal{N}$.
- The variables y_α are the barycentric coordinates of \mathbf{b} in terms of z_α and λ are obtained in a least-squares sense.
- Interior-point method to obtain a solution for $\mu_g = \mathbf{n}_g = 0$.
- Define \mathbf{n}_g with

$$\mu_g(\mathbf{n}_g) = \ln \frac{\mathbf{n}_g RT}{p_{\text{vapor}}} = -\lambda.$$

- Scale the liquid quantities by:

$$\mathbf{n}_g + \beta^0 \sum_{\alpha=1}^P y_\alpha \mathbf{x}_\alpha = \mathbf{b}.$$



Extended KKT System

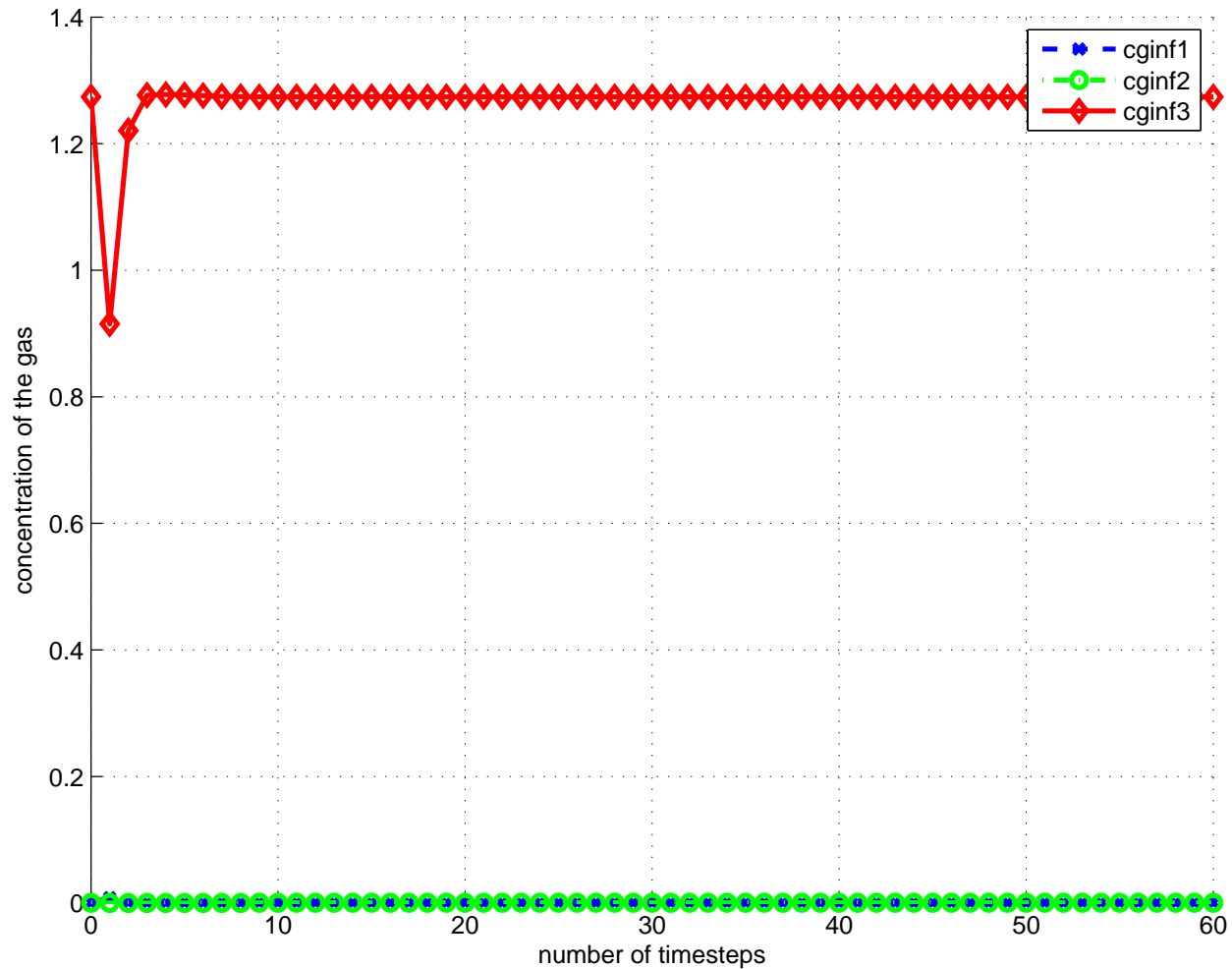
$$\begin{bmatrix}
 \mathbf{H}_b & 0 & 0 & \mathbf{I} & \mathbf{e} \\
 0 & y_\alpha \nabla^2 f(\mathbf{z}_\alpha) & \nabla f(\mathbf{z}_\alpha) + \boldsymbol{\eta} & y_\alpha & 0 \\
 0 & (\nabla f(\mathbf{z}_\alpha) + \boldsymbol{\eta})^T & \frac{\nu}{y_\alpha^2} & \mathbf{z}_\alpha^T & \mathbf{e} \\
 \mathbf{I} & y_\alpha^T & \mathbf{z}_\alpha & 0 & 0 \\
 \mathbf{e}^T & 0 & \mathbf{e}^T & 0 & 0
 \end{bmatrix}
 \begin{bmatrix}
 \mathbf{p}_b \\
 \mathbf{p}_z \\
 \mathbf{p}_y \\
 \mathbf{p}_\eta \\
 \mathbf{p}_\theta
 \end{bmatrix}
 =
 \begin{bmatrix}
 \mathbf{r}_b \\
 \mathbf{r}_z \\
 \mathbf{r}_y \\
 \mathbf{r}_\eta \\
 \mathbf{r}_\theta
 \end{bmatrix}$$

- \mathbf{H}_b is positive definite.
- Design of decomposition methods for the Newton system with control of the inertia.
 - Schur complement techniques induce modifications of the lower right matrix.
 - Sequential quadratic programming techniques.



Numerical Results

- Gas-Partitioning for one given b:

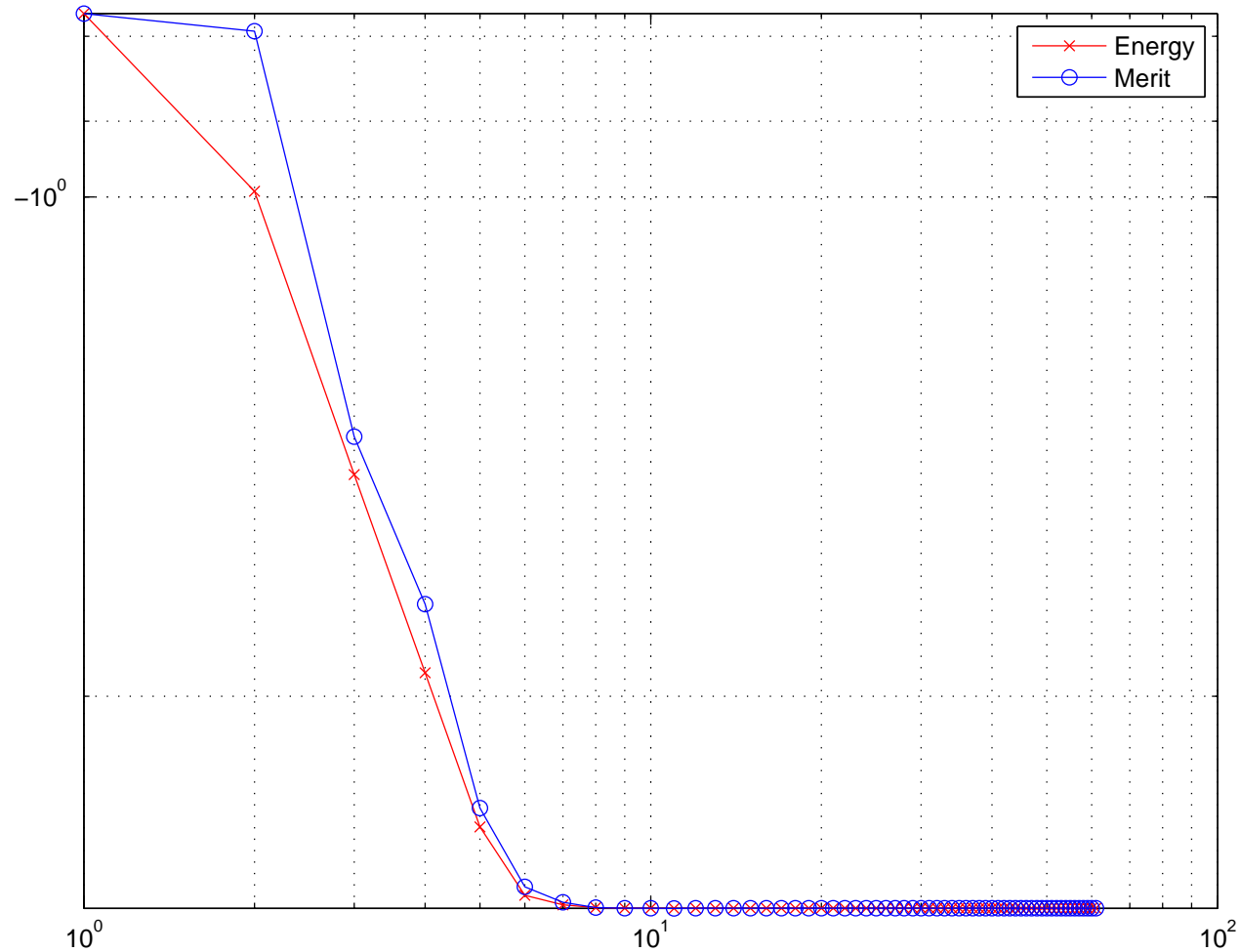


Concentrations n_g



Numerical Results

- Gas-Partitioning for one given b:



Merit function and objective function



Numerical Results (2)

- Sensitivity with respect to the size of the system.

| Size $N + 1$ | # Iterations (init.) | # Iterations | CPU time |
|--------------|----------------------|--------------|----------|
| 2 | 16 | 2 | 0.06 |
| 4 | 31 | 30 | 0.19 |
| 8 | 33 | 29 | 0.45 |
| 18 | 31 | 45 | 0.66 |

- Increase of the number of iterations.
- Increase of CPU time mainly due to the size of the linear systems.



Current and Future Work

1. Determination of the convex hull of a finite family of different energy functions (mixtures of aerosols).
2. Dynamic optimization (coupling with differential equations, time detection of activation/deactivation of constraints).
3. Dynamic optimization with a sequence of optimization problems (population of particles).
4. Extension to the determination of the convex hull with multiple local minima.



