Number of Simplex Iterations

Each iteration of Simplex can be implemented in polynomial time.

If we use lexicographic pivoting we know that Simplex requires at most $\binom{n}{m}$ iterations, because it will not visit a basis twice.

The input size is $L \cdot n \cdot m$, where n is the number of variables, m is the number of constraints, and L is the length of the binary representation of the largest coefficient in the matrix A.

If we really require $\binom{n}{m}$ iterations then Simplex is not a polynomial time algorithm.

Can we obtain a better analysis?

EADS II

© Harald Räcke

EADS II © Harald Räcke	



7 Klee Minty Cube

Number of Simplex Iterations

Observation Simplex visits every feasible basis at most once.

However, also the number of feasible bases can be very large.

 EADS II
 7 Klee Minty Cube

 © Harald Räcke
 118



However, Simplex may still run quickly as it usually does not visit all feasible bases.

In the following we give an example of a feasible region for which there is a bad Pivoting Rule.

EADS II © Harald Räcke

117

119

Pivoting Rule

A Pivoting Rule defines how to choose the entering and leaving variable for an iteration of Simplex.

In the non-degenerate case after choosing the entering variable the leaving variable is unique.

EADS II ©Harald Räcke 7 Klee Minty Cube

121

Observations

- We have 2n constraints, and 3n variables (after adding slack variables to every constraint).
- Every basis is defined by 2n variables, and n non-basic variables.
- There exist degenerate vertices.
- The degeneracies come from the non-negativity constraints, which are superfluous.
- In the following all variables x_i stay in the basis at all times.
- Then, we can uniquely specify a basis by choosing for each variable whether it should be equal to its lower bound, or equal to its upper bound (the slack variable corresponding to the non-tight constraint is part of the basis).
- We can also simply identify each basis/vertex with the corresponding hypercube vertex obtained by letting $\epsilon \rightarrow 0$.

Klee Minty Cube



Analysis

- In the following we specify a sequence of bases (identified by the corresponding hypercube node) along which the objective function strictly increases.
- The basis $(0, \ldots, 0, 1)$ is the unique optimal basis.
- ► Our sequence S_n starts at (0,...,0) ends with (0,...,0,1) and visits every node of the hypercube.
- An unfortunate Pivoting Rule may choose this sequence, and, hence, require an exponential number of iterations.



Analysis

Lemma 2

The objective value x_n is increasing along path S_n .

Proof by induction:

n = 1: obvious, since $S_1 = 0 \rightarrow 1$, and 1 > 0.

$n-1 \rightarrow n$

- For the first part the value of $x_n = \epsilon x_{n-1}$.
- By induction hypothesis x_{n-1} is increasing along S_{n-1} , hence, also x_n .
- Going from (0, ..., 0, 1, 0) to (0, ..., 0, 1, 1) increases x_n for small enough ϵ .
- For the remaining path S_{n-1}^{rev} we have $x_n = 1 \epsilon x_{n-1}$.
- By induction hypothesis x_{n-1} is increasing along S_{n-1} , hence $-\epsilon x_{n-1}$ is increasing along S_{n-1}^{rev} .

Analysis

The sequence S_n that visits every node of the hypercube is defined recursively



Remarks about Simplex Observation The simplex algorithm takes at most $\binom{n}{m}$ iterations. Each iteration can be implemented in time $\mathcal{O}(mn)$. In practise it usually takes a linear number of iterations. EADS II 7 Klee Minty Cube CHarald Räcke

Remarks about Simplex

Theorem

For almost all known deterministic pivoting rules (rules for choosing entering and leaving variables) there exist lower bounds that require the algorithm to have exponential running time ($\Omega(2^{\Omega(n)})$) (e.g. Klee Minty 1972).

EADS II © Harald Räcke 7 Klee Minty Cube

129

131



Remarks about Simplex

Theorem

For some standard randomized pivoting rules there exist subexponential lower bounds ($\Omega(2^{\Omega(n^{\alpha})})$ for $\alpha > 0$) (Friedmann, Hansen, Zwick 2011).

EADS II © Harald Räcke 7 Klee Minty Cube

130

