### 3.7 Computational efficiency of the simplex method

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The computational efficiency of the simplex method is determined by
(1) the computational effort of each iteration;
(2) the number of iterations.

How many iterations are needed in the worst case?

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## Idea for negative answer (lower bound)

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- a path that visits all vertices and always moves from a vertex to an adjacent one that has lower costs.


## Computational efficiency of the simplex method

## Unit cube

Consider the unit cube in $\mathbb{R}^{n}$, defined by the constraints

$$
0 \leq x_{i} \leq 1, \quad i=1, \ldots, n
$$

The unit cube has

- $2^{n}$ vertices;
- a spanning path, i.e. a path traveling the edges of the cube visiting each vertex exactly once.




## Computational efficiency of the simplex method

## Klee-Minty cube

Consider a perturbation of the unit cube in $\mathbb{R}^{n}$, defined by the constraints

$$
\begin{aligned}
0 & \leq x_{1} \leq 1 \\
\epsilon x_{i-1} & \leq x_{i} \leq 1-\epsilon x_{i-1}, \quad i=2, \ldots, n
\end{aligned}
$$

for some $\epsilon \in(0,1 / 2)$.


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## Theorem

Consider the linear programming problem of minimizing $-x_{n}$ subject to the constraints above. Then
(1) The feasible set has $2^{n}$ vertices.
(2) The vertices can be ordered so that each one is adjacent to and has lower cost than the previous one.
(3) There exists a pivoting rule under which the simplex method requires $2^{n}-1$ changes of basis before it terminates.

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## The diameter of polyhedra

## Definition

- The distance $d(x, y)$ between two vertices $x, y$ is the minimum number of edges required to reach $y$ starting from $x$.
- The diameter $D(P)$ of polyhedron $P$ is the maximum $d(x, y)$ over all pairs of vertices $(x, y)$.
- $\Delta(n, m)$ is the maximum $D(P)$ over all bounded polyhedra in $\mathbb{R}^{n}$ that are represented in terms of $m$ inequality constraints.
- $\Delta_{u}(n, m)$ is the maximum $D(P)$ over all polyhedra in $\mathbb{R}^{n}$ that are represented in terms of $m$ inequality constraints.


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Known bounds

- Lower bounds: $\Delta_{u}(n, m) \geq m-n+\left\lfloor\frac{n}{5}\right\rfloor$
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$$
\Delta(n, m) \leq \Delta_{u}(n, m)<m^{1+\log _{2} n}=(2 n)^{\log _{2} m}
$$

## Average case behavior of the simplex method

## Remark

- Despite the exponential lower bounds on the worst case behavior of the simplex method (Klee-Minty cubes etc.), the simplex method usually behaves well in practice.
- The number of iterations is "typically" $O(m)$.
- There have been several attempts to explain this phenomenon from more a theoretical point of view.
- These results say that "on average" the number of iterations is $O(\cdot)$ (usually polynomial).
- One main difficulty is to come up with a meaningful and, at the same time, manageable definition of the term "on average".

