# LINEAR PROGRAMMING

[V. CH4]: EFFICIENCY OF THE SIMPLEX METHOD

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## PERFORMANCE MEASURES

WORST-CASE ANALYSIS OF THE SIMPLEX METHOD

EMPIRICAL AVERAGE PERFORMANCE

#### PERFORMANCE MEASURES

We saw that the simplex method (equipped with anti-cycling rules) will solve any linear programming problem for which an optimal solution exists. The question, now, is *how fast* it will solve such a problem.

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### Performance measures are:

- worst case looks at all problems of a given "size" and asks how much effort is needed to solve the hardest.
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## Performance measures are:

- worst case looks at all problems of a given "size" and asks how much effort is needed to solve the hardest.
- average case looks at the average amount of effort needed to solve all.
- → Worst-case analyses are generally easier. It needs to provide an upper bound on how much effort required + a specific example attaing this bound.
- $\leadsto$  On an average case analysis, one needs a stochastic model of random problems and evaluate the amount of effort required to solve every problem in the sample space.

worst-case is more tractable, but less relevant while dealing with real problems

#### PERFORMANCE MEASURES

Before looking at the worst case, we must discuss:

**First:** how do we specify the size of a problem? recall that we have m (constraints), n (variables)

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- the number of bits needed to store all data

Even with the last one, "size" of data might still be ambiguous. (why?)

we shall simply focus on m and n to characterize the size of a problem.

#### PERFORMANCE MEASURES

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### Fortunately, a reasonable substitute:

Algorithms are generally iterative processes. So

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First factor is platform-independent and so a *resasonable suragate* for the acctual time (not when different algorithm are compared!)

We simply count the number of iterations (pivots).

WORST-CASE ANALYSIS OF THE SIMPLEX METHOD

Performance Measures

WORST-CASE ANALYSIS OF THE SIMPLEX METHOD

EMPIRICAL AVERAGE PERFORMANCE

Since the simplex method operates by moving from one dictionary to another (without cycling), an upper bound on the number of iterations is simply the number of possible dictionaries:

$$\binom{n+m}{m}$$

For a fixed value of the sum n + m, this expression is maximized when m = n.

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How big is it? Not hard to show:

$$\frac{1}{2n}2^{2n} \le \binom{2n}{n} \le 2^{2n}$$

For n=25:

$$2^{50} = 1.1259 \times 10^{15}$$

Our best chance for finding a bad example is to look at the case where m = n.

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lets look at the generic example:

Given constants 
$$1 = \beta_1 \ll \beta_2 \ll \cdots \ll \beta_n$$

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$$\max_{x} \sum_{j=1}^{n} 2^{n-j} x_{j} - \frac{1}{2} \sum_{j=1}^{n} 2^{n-j} \beta_{j}$$
s.t. 
$$2 \sum_{j=1}^{i-1} 2^{i-j} x_{j} + x_{i} \leq \sum_{j=1}^{i-1} 2^{i-j} \beta_{j} + \beta_{i} \quad i = 1, 2, \dots, n$$

$$x_{j} \geq 0 \qquad i = 1, 2, \dots, n$$

### For n=3

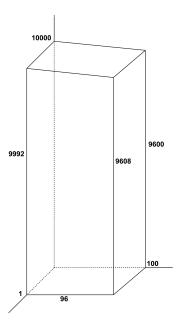
$$\begin{array}{lllll} \max_{x} & 4x_{1} + & 2x_{2} + & 1x_{3} - & 2\beta_{1} - \beta_{2} - \frac{1}{2}\beta_{3} \\ \text{s.t.} & 1x_{1} + & + & \leq \beta_{1} \\ & 4x_{1} + & 1x_{2} + & \leq 2\beta_{1} + \beta_{2} \\ & 8x_{1} + & 4x_{2} + & 1x_{3} \leq 4\beta_{1} + 2\beta_{2} + \beta_{3} \\ & x_{1}, & x_{2}, & x_{3} \geq 0 \end{array}$$

For n=3

$$\begin{array}{lllll} \max & 4x_1 + & 2x_2 + & 1x_3 - & 2\beta_1 - \beta_2 - \frac{1}{2}\beta_3 \\ \text{s.t.} & 1x_1 + & + & \leq \beta_1 \\ & 4x_1 + & 1x_2 + & \leq 2\beta_1 + \beta_2 \\ & 8x_1 + & 4x_2 + & 1x_3 \leq 4\beta_1 + 2\beta_2 + \beta_3 \\ & x_1, & x_2, & x_3 \geq 0 \end{array}$$

Assuming  $\beta_2 = 98, \beta_3 = 9800$ , the feasible region looks like :

$$\begin{array}{lll} 1x_1 & \leq 1 \\ 4x_1 + & 1x_2 & \leq 100 \\ 8x_1 + & 4x_2 + & 1x_3 \leq 10000 \\ x_1, & x_2, & x_3 \geq 0 \end{array}$$



Constraints represent a "minor" distortion to an 3-dimensional hypercube:

$$0 \le x1 \le 1$$

$$0 \leq x2 \leq 100$$

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 $0 \le x3 \le 10000$ 

This region has  $2^{n=3}$  vertices and the idea is to trick the pivot rule so as to visit all of them.

Back to the generic Klee-Minty example (n = 3), the first dictionary is

$\zeta = -$	$2\beta_1$ –	$1\beta_2$ —	$0.5\beta_3 +$	$4x_1 +$	$\frac{2}{x_2}$ +	$1 x_3$
$\overline{w_1 = +}$	$1\beta_1$		_	$1x_1$		
$w_2 = +$	$2\beta_1$ +	$1\beta_2$	_	$4x_1 -$	$1x_2$	
$w_3 = +$	$4\beta_1$ +	$2\beta_2$ +	$1\beta_3$ –	$8x_1$ -	$4x_2 -$	$1x_3$

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which is feasible.

Using the largest-coefficient rule:  $(x_1, w_1)$  are the entering, leaving pair.

Corresponding vertex on the Klee-Minty cube is: (0,0,0)

# After the first pivot

$\zeta = +$	$2\beta_1$ –	$1\beta_2$ –	$0.5\beta_3$ –	$4w_1 + $	$2 x_2 +$	$1 x_3$
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## After the first pivot

for the next pivot:  $(x_2, w_2)$ 

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Corresponding vertex on the Klee-Minty cube is:  $(\beta_1, -2\beta_1 + 1\beta_2, 0)$ 

## After the third pivot

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for the next pivot:  $(x_3, w_3)$ 

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for the next pivot:  $(x_3, w_3)$ 

Corresponding vertex on the Klee-Minty cube is:  $(0, 2\beta_1 + 1\beta_2, 0)$ 

## After the fourth pivot

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for the next pivot:  $(x_1, w_1)$ 

#### After the fourth pivot

for the next pivot:  $(x_1, w_1)$ 

Corresponding vertex on the Klee-Minty cube is:  $(0, 2\beta_1 + 1\beta_2, -4\beta_1 - 2\beta_2 + 1\beta_3)$ 

# After the fifth pivot

$\zeta = +$	$2\beta_1$ –	$1\beta_2$ +	$0.5\beta_3$ -	$4w_1 + $	$2 w_2 -$	$1w_3$
$x_1 = +$	$1\beta_1$		_	$1w_1$		
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for the next pivot:  $(w_2, x_2)$ 

## After the fifth pivot

for the next pivot:  $(w_2, x_2)$ 

Corresponding vertex on the Klee-Minty cube is:  $(1\beta_1, -2\beta_1 + 1\beta_2, +4\beta_1 - 2\beta_2 + 1\beta_3)$ 

## After the sixth pivot

## After the sixth pivot

for the next pivot:  $(w_1, x_1)$ 

## After the sixth pivot

for the next pivot:  $(w_1, x_1)$ 

Corresponding vertex on the Klee-Minty cube is:  $(1\beta_1, 0, -4\beta_1 + 2\beta_2 + 1\beta_3)$ 

## And finally after the seventh pivot

$\zeta = +$	$2\beta_1$ +	$1\beta_2$ +	$0.5\beta_3$ $-$	$4x_1 -$	$2x_2 - $	$1w_3$
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Corresponding vertex on the Klee-Minty cube is:  $(0,0,+4\beta_1+2\beta_2+1\beta_3)$ 

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On a machin with 1000 ips. It will take 40 billion years to solve! (The age of the universe is estimated to be 13.7 billion years)

Yet, problems with 10,000 to 100,000 variables/constraints are solved routinely every day. (Worst case analysis!)

• Note that the final dictionary could have been reached from the initial dictionary in just one pivot if we had selected  $x_3$  to be the entering variable. But the largest-coefficient rule dictated selecting  $x_1$  and the exponential worst case running time.

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Does there exist a varient of the Simplex method with polynomial worst case performance?

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- Indeed this is an open question:

Does there exist a varient of the Simplex method with polynomial worst case performance?

• However, For linear programs, we have other algorithm (called *interior point* methods) with *polynomial worstcase performance*. In contrast to the simplex algorithm, which finds an optimal solution by traversing the edges between vertices on a polyhedral set, interior-point methods move through the interior of the feasible region.

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These are the necessary steps needed to take toward a proof of worst case exaponantial time of the generic Klee-Minty example. (Left as part of to the second HW)

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Lets again consider the simple simplex implementation discussed previous week.