

## **Module 1: Formulations (Nonlinear Models)**

## So far ...

- Linear programs, and

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**Now:** Nonlinear generalization!

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# Nonlinear Programs

A non-linear program (NLP) is of the form

$$\begin{array}{ll} \min & f(x) \\ \text{s.t.} & g_1(x) \leq 0 \\ & g_2(x) \leq 0 \\ & \dots \\ & g_m(x) \leq 0, \end{array}$$

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**Note:** Linear programs are NLPs!

Example 1: Finding Close Points in an LP

## Finding Close Points in an LP

**Problem:** we are given an LP (P), and an **infeasible** point  $\bar{x}$ .

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Example 2: Binary IP via NLP

# NLPs and Binary IPs

Suppose we are given a **binary IP** (i.e., an integer program all of whose variables are binary).

$$\begin{aligned} \max \quad & c^T x \\ \text{s.t.} \quad & Ax \leq b \\ & x \geq 0 \\ & x_j \in \{0, 1\} \quad (j \in \{1, \dots, n\}) \end{aligned}$$

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$$\begin{aligned} \max \quad & c^T x \\ \text{s.t.} \quad & Ax \leq b \\ & x \geq 0 \\ & \sin(\pi x_j) = 0 \quad (j \in [n]) \quad (*) \end{aligned}$$

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**Q:** Can you change the NLP to express the fact that  $x_j$  is **any non-negative integer** instead of binary?

**Correctness:** note that  $\sin(\pi x_j) = 0$  only if  $x_j$  is an integer.

### Example 3: Fermat's Last Theorem

# Fermat's Last Theorem

## Conjecture [Fermat, 1637]

There are **no integers**  $x, y, z \geq 1$  and  $n \geq 3$  such that

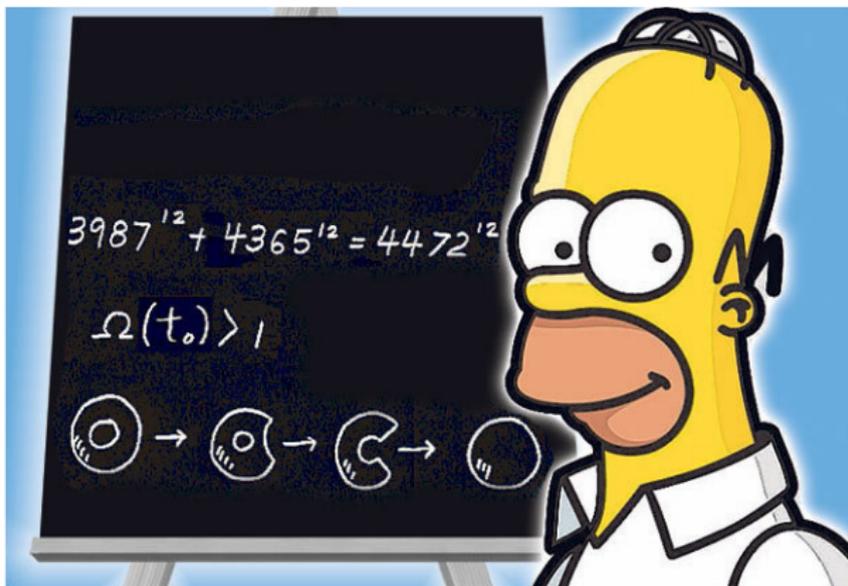
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... doh!

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# NLP for Fermat's Last Theorem

$$\min (x_1^{x_4} + x_2^{x_4} - x_3^{x_4})^2 \\ + (\sin \pi x_1)^2 + (\sin \pi x_2)^2 + (\sin \pi x_3)^2 + (\sin \pi x_4)^2$$

$$\text{s.t. } x_i \geq 1 \quad (i = 1 \dots 3)$$

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## Remark

Fermat's Last Theorem is true iff the NLP has optimal value **greater than** 0.

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Proving Fermat's Last Theorem amounts to **showing that the value 0 can not be attained!**

## Recap

- Non-linear programs are of the form

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- Non-linear programs are strictly more general than integer programs, and thus likely difficult to solve.
- Some famous questions in Math can easily be reduced to solving certain NLPs