

Why Decision Optimization is important

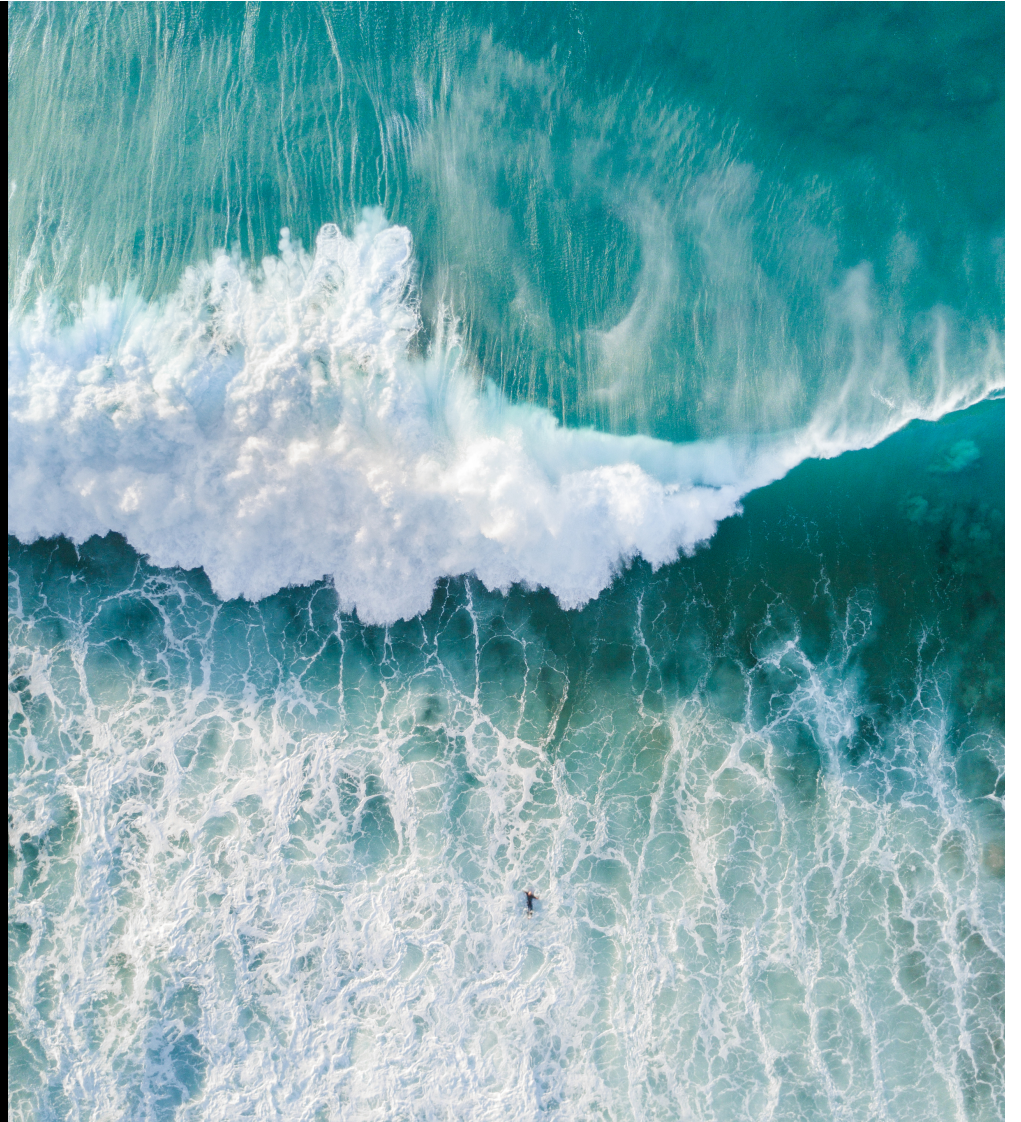
*“Everything should be made
as simple as possible,
but not one bit simpler.”*

Albert Einstein, (attributed)

*“All models are wrong,
but some are useful.”*

George Box

Machine
Learning is
eating the
world...



There is more &
more data.

We draw insights
from that data using
machine learning

to make better
decisions.

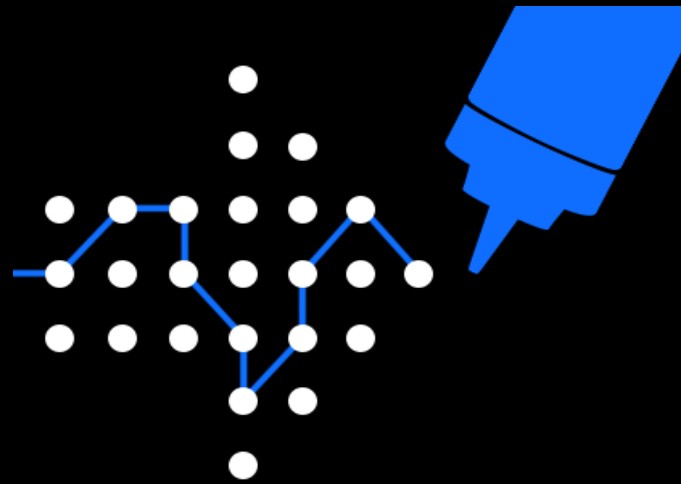


And then...

...we decide based
on our guts : (



Instead, we can leverage the insights and data we got...



... and use **mathematical optimization** to plan for better, data-driven actions.

And the tool for that is **Decision Optimization** feature on Watson Studio.

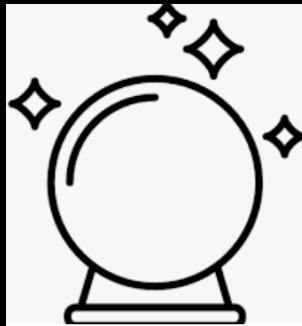
Descriptive

What happened



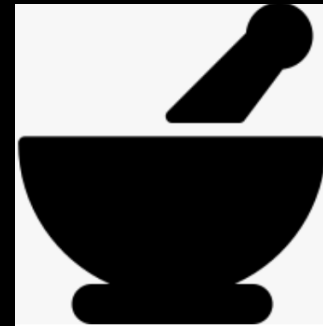
Predictive

What will happen



Prescriptive

What you'll do



Descriptive

Decision

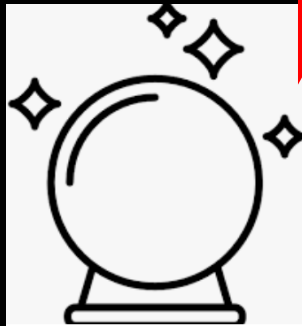
Optimization

What happened



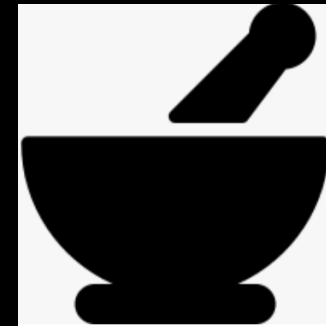
Predictive

What will



Prescriptive

What you'll do



Use cases for Decision Optimization are everywhere

Because we have limitations everywhere:

- time
- number of attempts to client
- number of people
- etc.



naming just few of them:

- product modeling
- worker shifts planning
- production planning
- marketing campaign planning
- managing sales activities
- choosing best offering portfolio for clients
- and more



poll 1

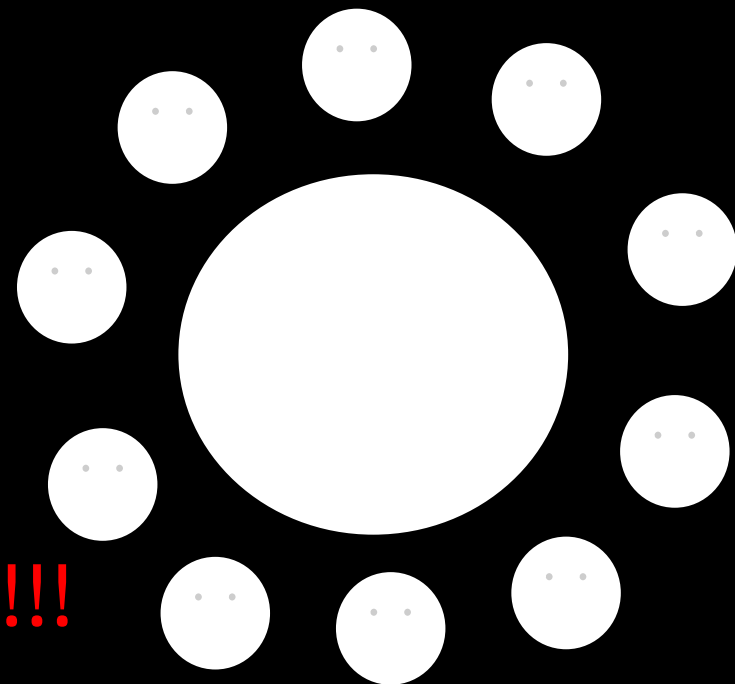


Mathematical Optimization

basic concepts

Optimization is hard:

- It tricks intuition
- It is hard to compute



10 factorial !!!

Optimization is Everywhere

Personal choices

- best career choices,
- best use of your time
- best strategies,
- best value for the dollar

Company choices

- maximize value to shareholders
- determine optimal mix of products or services
- minimize production costs
- minimize cost of getting product to customers
- maximize value of advertising
- hire the best workers

poll 2



(data)

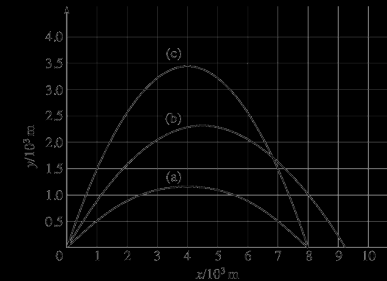


You will see it everywhere but not instantly!

(algorithm)

$$h = -16t^2 + vt + s$$

t- time
s – init height
v – init velocity
h - height



(model)

```
import gym
import numpy as np
import random

env = gym.make("MountainCar-v0")
env.reset()

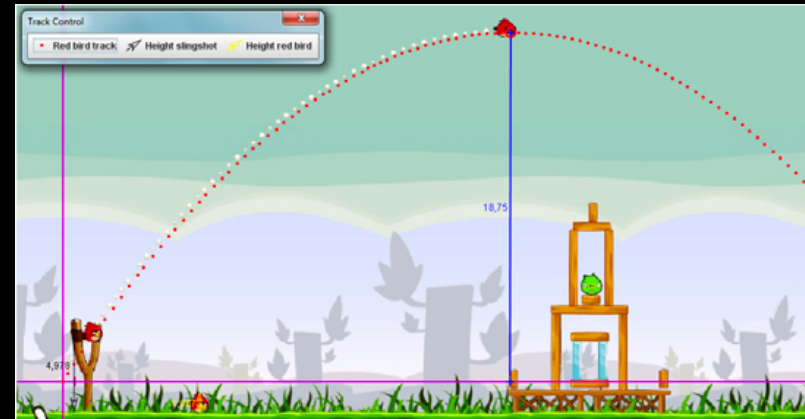
# print(env.observation_space.high)
# print(env.observation_space.low)
# print(env.action_space.n)

DISCRETE_OS_SIZE = [20] * len(env.observation_space.high)
discrete_os_win_size = (env.observation_space.high - env.observation_space.low) / DISCRETE_OS_SIZE
# print(discrete_os_win_size)

q_table = np.random.uniform(low=-2, high=0, size=(DISCRETE_OS_SIZE * [env.action_space.n]))

done = False
actioninput = [0,1,2]
while not done:
    # action 0 = turn left
    # action 1 = do nothing
    # action 2 = turn right
    action = random.choice(actioninput)
    new_state, reward, done, _ = env.step(action)
    print(reward, new_state)
    env.render()

env.close()
```



Basic concepts in Optimization

Linear Programming

Simplex algorithm

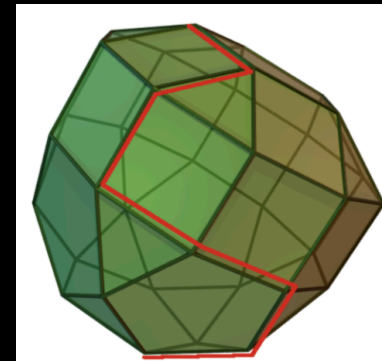
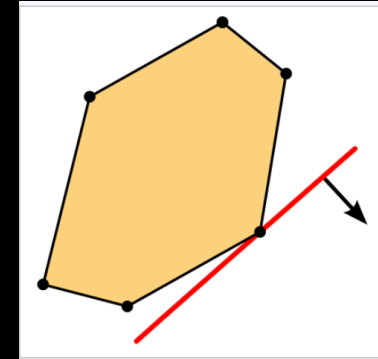
Linear Programming problem:

- Decision variables
- Constraints
- Objective function

Linear Programming

(LP, also called **linear optimization**) is a method to achieve the **best outcome** (such as **maximum profit** or **lowest cost**) in a mathematical model whose requirements are represented by linear relationships

- minimize or maximize a linear objective
- subject to linear equalities and inequalities



The optimization paradigm

1# Decision variables - sth under your control

- The work schedules of each employee
- The level of investments in a portfolio
- what subjects a student should take in each semester

2# Objective function (of the decision variables):

- minimize cost or ...
- maximize expected return or ...
- make the last semester as enjoyable as possible or ...

The optimization paradigm

3# Constraints: restrictions on the decision variables:

– “Business rules”

- no worker can work more than 5 consecutive days
- There is at most 2% investment in any stock in the portfolio
- students must take a prerequisite of a subject before taking the subject

– “Physical laws”

- No worker can work a negative amount of time
- The amount of a goods in inventory at the end of period j is the amount of goods arriving during period j plus the amount of goods in inventory in period $j-1$ minus the amount of goods that are sold in the period.

Generic optimization algorithm

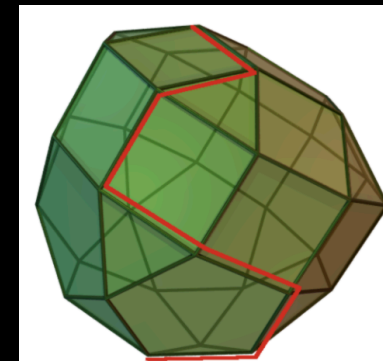
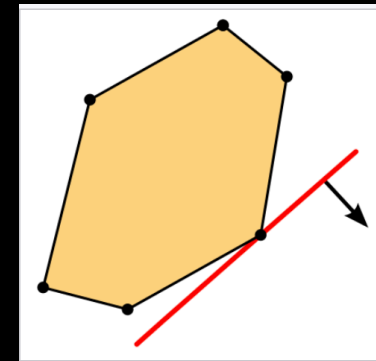
Let x be the vector of decision variables:
Suppose f, g_1, g_2, \dots, g_m are functions

$\max f(x)$
Maximize the objective

Subject to $g_i(x) \geq b_i$

for each $i = 1$ to m
Satisfy the constraints

$x \geq 0$
typically but not always the case.



poll 3



Putting this all together...

...we define 3 things

1# **Decision Variables** – we can manipulate them to achieve different results

2# **Constraints** – they put boundaries to our max or min function

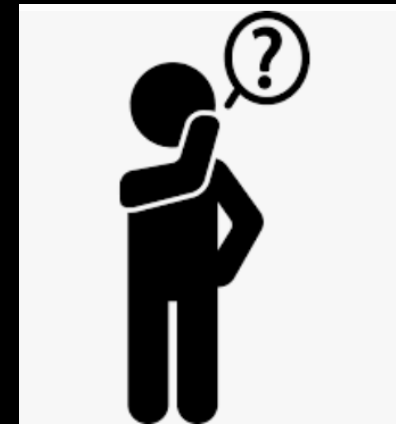
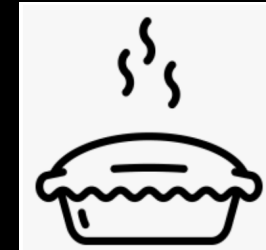
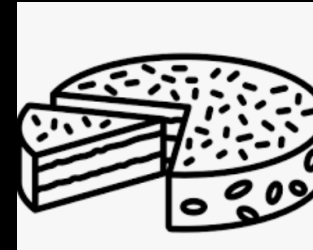
3# **Objective function** – the goal we want to achieve

Example

- minimize or maximize a linear objective
- subject to linear equalities and inequalities

Max is in a pie eating contest that lasts **1 hour**.
Each torte that he eats takes **2 minutes**.
Each apple pie that he eats takes **3 minutes**.
He receives **4 points** for each **torte** and **5 points** for each **pie**.

What should Max eat to get the most points?



Step 1. Determine the decision variables

- Let x be the number of **tortes** eaten by Max.
- Let y be the number of **pies** eaten by Max.

Step 2. Determine the objective function

Maximize $z = 4x + 5y$ (**objective function**)

Step 3. Determine constraints

subject to $2x + 3y \leq 60$ (**constraint**)

$x \geq 0 ; y \geq 0$ (**non-negativity constraints**)



Simplex algorithm

Solution.

A feasible solution satisfies all of the constraints.

$x = 10, y = 10$ is feasible; $x = 10, y = 15$ is infeasible.

An optimal solution is the best feasible solution.

The optimal solution is $x = 30, y = 0, z = 120$

Demo time



Thank you!

