

PART 1 Numerical Optimization

- Root finding / Newton's type methods
- Nonlinear Optimisation
- Newton's type methods

(1^{st} + 2^{nd} order
optimality conditions)

Convergence rates
Contraction
Invariance
Conditions for convergence

Equality constrained
Inequality constraints

- (Automatic Differentiation)
- (Parameter estimation)

PART 2 Discrete time Optimal Control

- Formulation, Analysis and Structure
- Dynamic Programming (Discrete state space, LQ, H_{inf}, LQR)
 - ↳ Coordinate of the value function
 - ↳ Minimum principle in DT
 - ↳ Iterative DP
 - ↳ Differential Dynamic Programming.

PART 3 Optimal Control in Continuous-time

- Formulation, overview
- Numerical Integration (Trapezoid Rule, Orthogonal Collocation)
- Hamilton-Jacobi-Bellman (LQR and Riccati Eqn's)
- Pontryagin and Indirect Approach (Bolton, Regular + Singular Arcs)
- Direct Approaches (Single/ Multiple Shooting, Direct Collocation)

PART 4 MPC and moving Prediction horizon

- NMPC
- Parametric NLP
- MHE (State and Parameter estimation)
- Feedback Linearization

CHEM-E7165

- 2h h LECTURES
- 2h h EXERCISES

} HYBRID, AS ALWAYS

THE OBJECTIVE OF THE COURSE IS TO PROVIDE AN INTRODUCTION
TO OPTIMAL CONTROL (MATHEMATICS + NUMERICS)



THERE WILL BE SOME CODING

- PYTHON

- MATLAB

- INTUITIVE ROUTINES FOR
OPTIMAL CONTROL

TUESDAYS - 2/1 SESSIONS (LECTURE + EXERCISES)

WEDNESDAYS - 1 SESSION (LECTURE)

FRIDAYS - 1/2 SESSION (EXERCISES + CLASSES)

EXAM ON FEB 17 - WE WILL NOT HAVE AN EXAM

→ EXERCISES → PROJECT WORK

OUTCOMES

- UNDERSTAND THE BASIC FORMULATION OF NONLINEAR PROGRAMMING (GENERAL CLASS OF NUMERICAL OPTIMIZATION) PROBLEMS
- UNDERSTAND THE BASIC METHODS OF NUMERICAL INTEGRATION OF DYNAMICAL MODELS (ODE SIMULATION)
- UNDERSTAND HOW TO COMBINE THE ABOVE TO FORMULATE AND SOLVE OPTIMAL CONTROL PROBLEMS
 - DISCRETE TIME
 - CONTINUOUS TIME

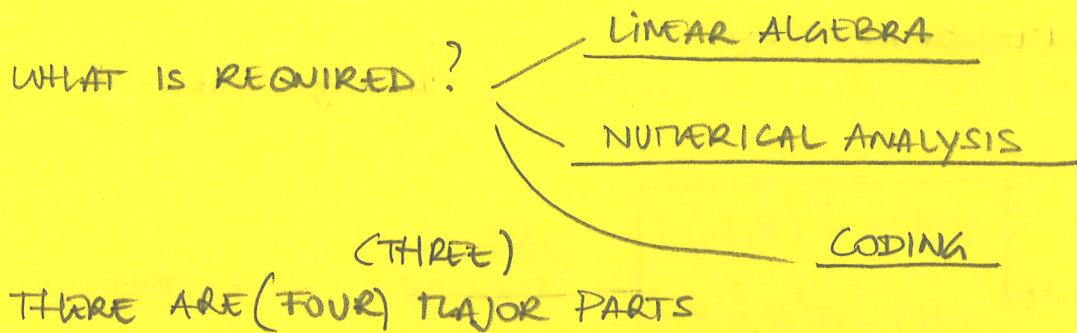
OPTIMAL CONTROL CONSIDERS THE OPTIMIZATION OF
DYNAMICAL SYSTEMS

IT BRIDGES TWO LARGE FIELDS OF APPLIED MATHEMATICS

* NUMERICAL OPTIMIZATION

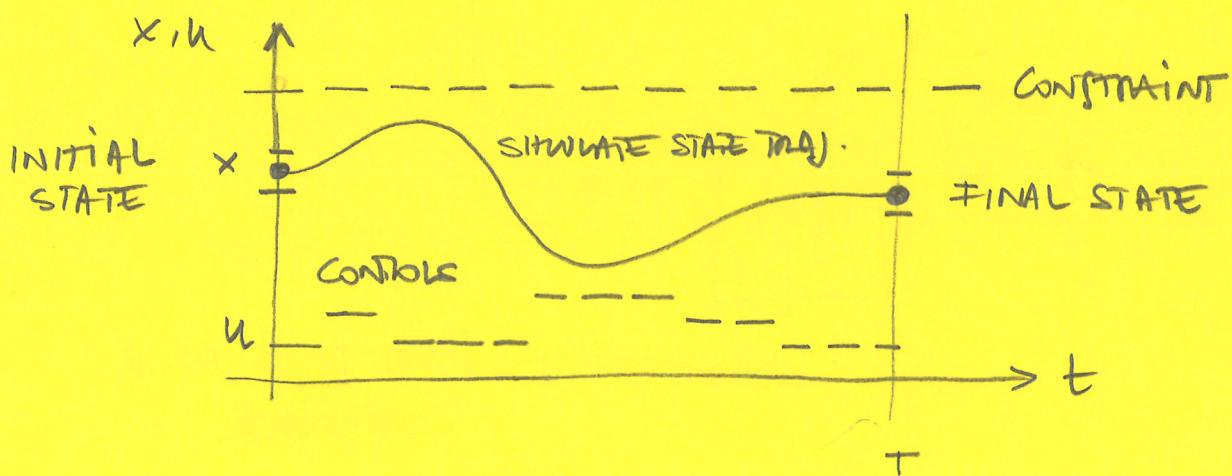
* SYSTEM THEORY (NUMERICAL SIMULATION + ANALYSIS)
OR INTRODUCE

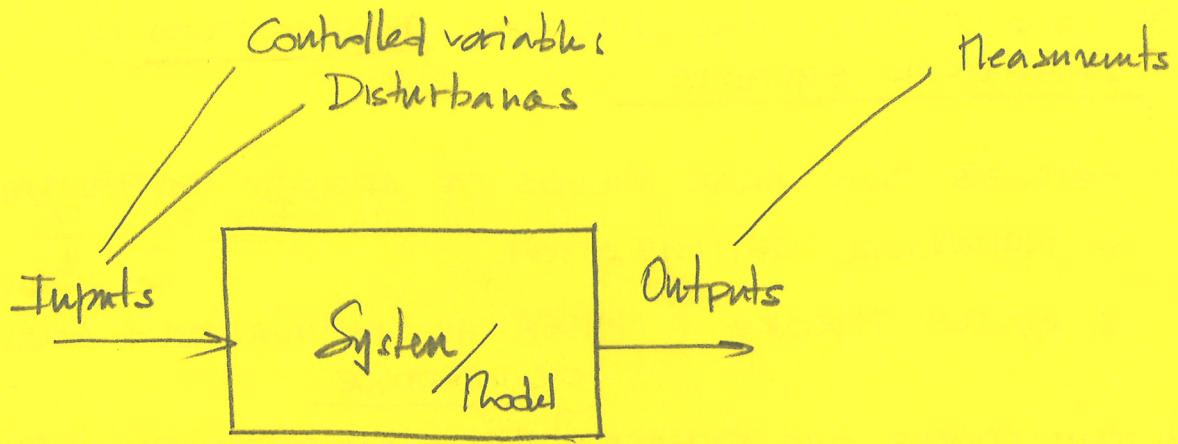
IN THIS COURSE WE REFRESH BASIC CONCEPTS FROM BOTH FIELDS
WITH SOME MORE WEIGHT TO NUMERICAL OPTIMIZATION



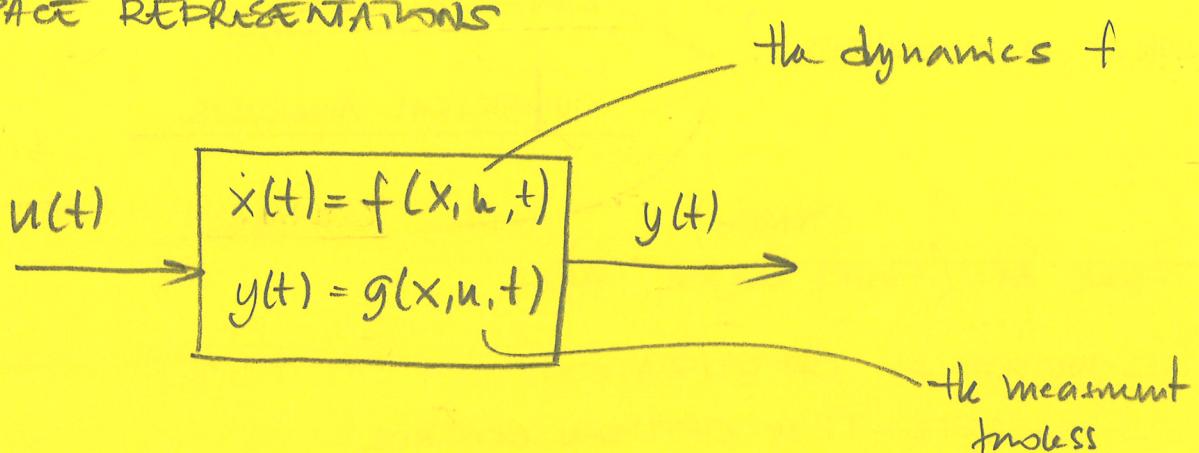
- NUMERICAL OPTIMIZATION
- DISCRETE-TIME OPTIMAL CONTROL
- CONTINUOUS-TIME OPTIMAL CONTROL
- (ONLINE OPTIMAL CONTROL)

FOR A GIVEN SYSTEM STATE x , WHICH CONTROLS u LEAD TO
THE BEST (IN SOME SENSE) OBJECTIVE, WITHOUT VIOLATING
THE CONSTRAINTS?

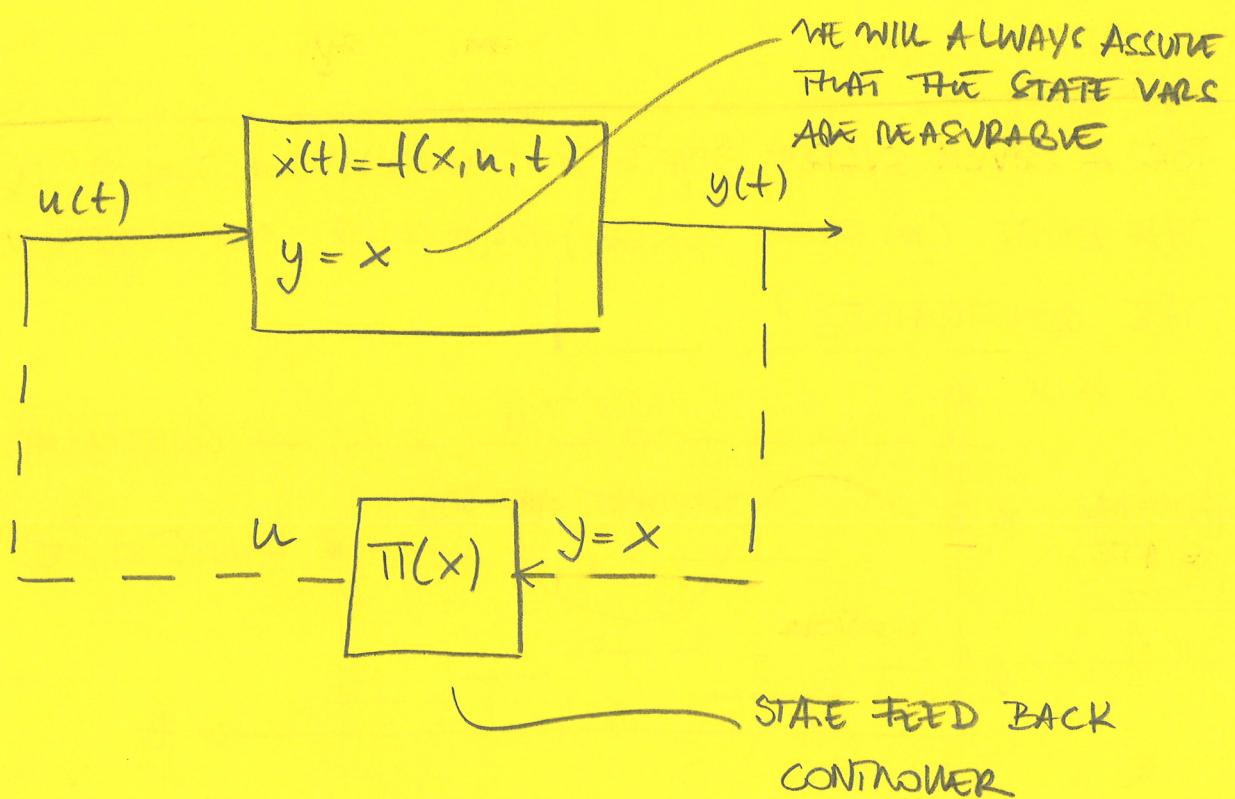




STATE SPACE REPRESENTATIONS



x are the state variables \rightarrow they evolve in time.



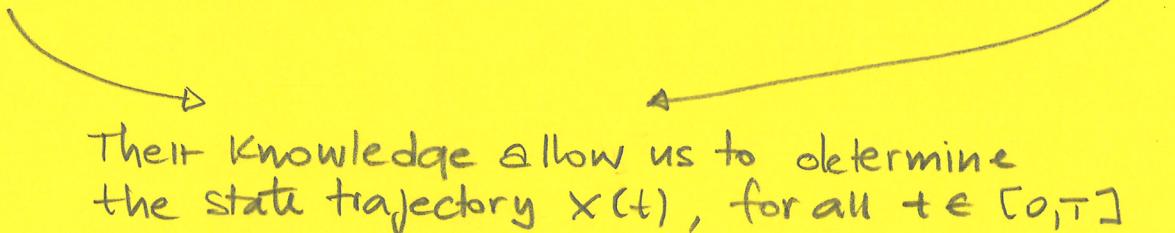
- WE IDENTIFY DYNAMICAL SYSTEMS AS PROCESSES THAT EVOLVE IN TIME
 - They can be characterized by STATES x that allow us to predict the future behaviour of the system
 - Often the dynamics can be controlled by a suitable choice of inputs, we denote them as CONTROLS, u

TYPICALLY THE CONTROLS SHOULD BE OPTIMALLY CHOSEN

- In such a way that an OBJECTIVE FUNCTION is optimized subjected to some CONSTRAINTS

TECHNIQUES FOR NUMERICALLY CHOOSING OPTIMAL CONTROLS IS THE MAIN TOPIC OF THIS COURSE

An important property of a dynamical system is the INITIAL VALUE OF THE STATE VARIABLES, x_0 and the CONTROL INPUT TRAJECTORY $u(t)$ for all $t \in [0, T]$



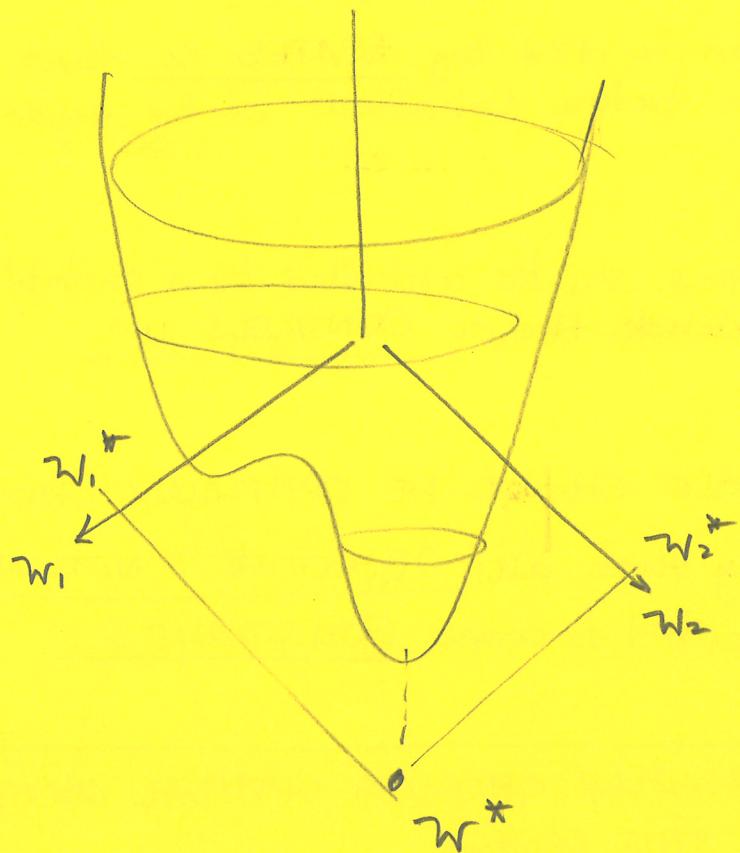
WE DISCUSS IMPORTANT CLASSES OF DYNAMICAL MODELS

- ② DISCRETE-TIME
- ① CONTINUOUS-TIME

THEN WE OVERVIEW IMPORTANT CLASSES OF OPTIMIZATION

PROBLEMS

$$f(w_1, w_2) = f(w)$$

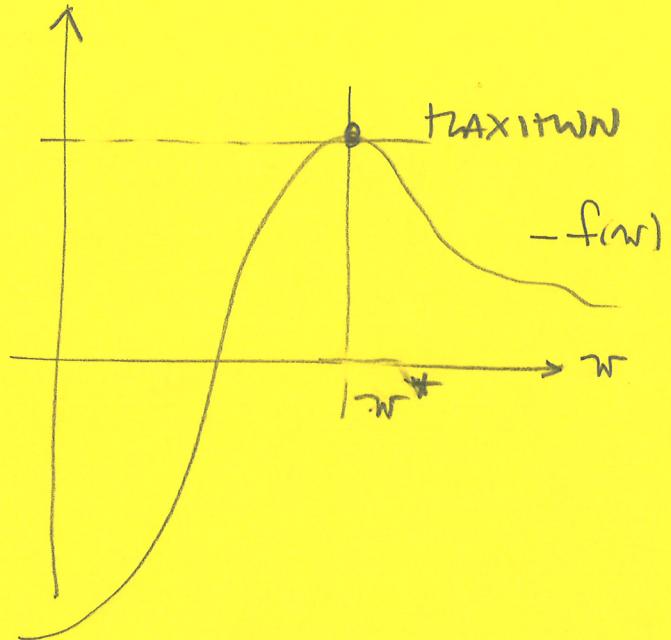
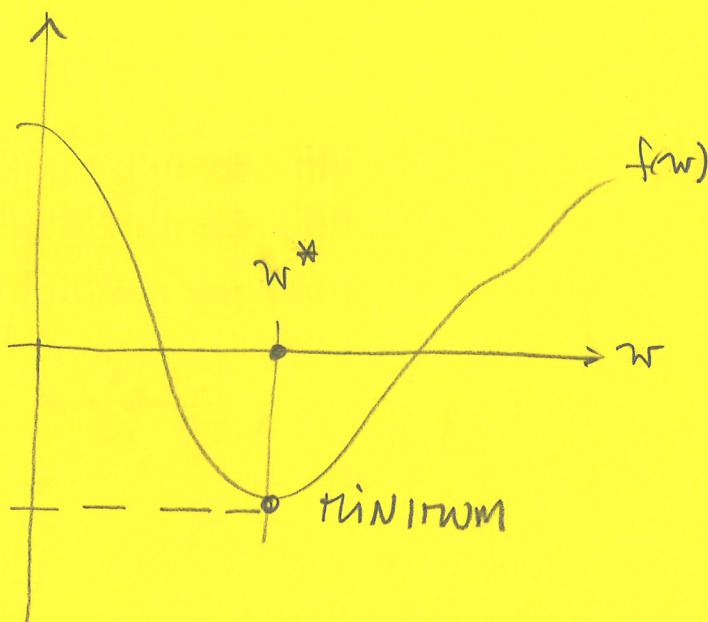


WHAT IS OPTIMISATION?

OPTIMISATION MEANS 'SEARCH FOR THE BEST SOLUTION'
OPTIMAL
IN SOME SENSE

MATHEMATICALLY, THIS MEANS 'MINIMISATION' OR 'MAXIMISATION' OF SOME OBJECTIVE FUNCTION

- $f(w)$ IS USED TO DENOTE THE OBJECTIVE FUNCTION
- w ARE THE 'DECISION VARIABLES'

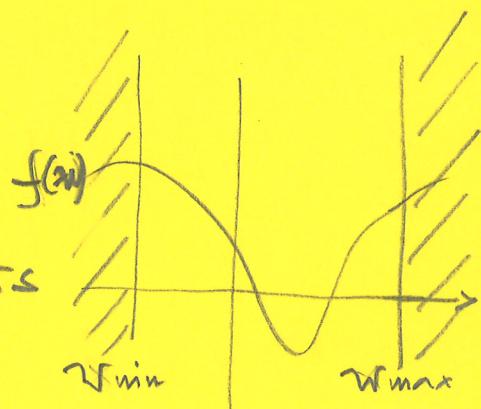


SOLVING MAXIMISATION PROBLEMS IS EQUIVALENT TO SOLVING MINIMISATION PROBLEMS

→ SAME w^*

FROM NOW ON, MINIMISATION

VARIABLES w ARE SUBJECT TO CONSTRAINTS



PYTHON (or MATLAB) + CASADI

