

B-SPLINES-BASED SOLUTIONS OF DIFFERENTIAL EQUATIONS IN BIOENGINEERING: MOTION MONITORING AND IMAGE PROCESSING APPLICATIONS

Bachelor's thesis

Degree in Biomedical Engineering

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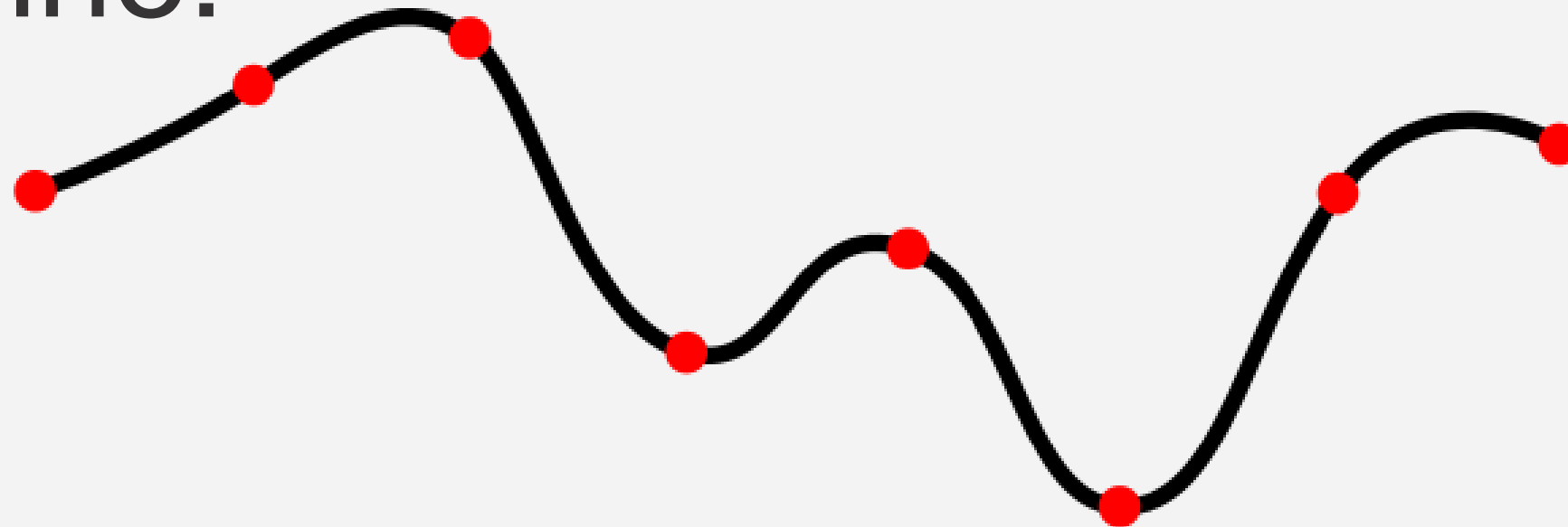


Universidad
Rey Juan Carlos

Introduction

Framework to provide solutions to differential equation in the field of bioengineering.

Spline:

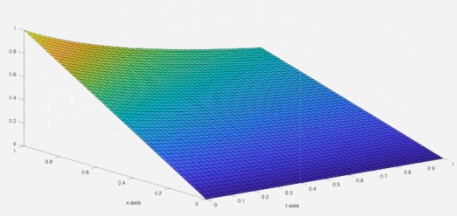


Bioengineering applications:

Estimate human motion in rehabilitation.

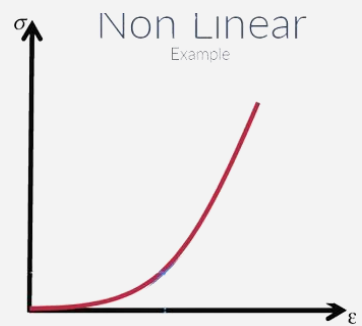
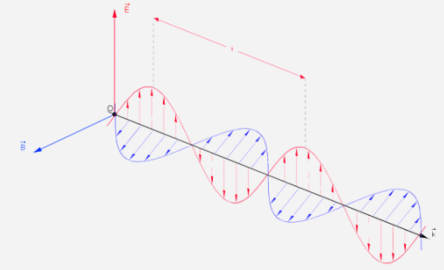
Represent pre-operative models as splines.

Objectives



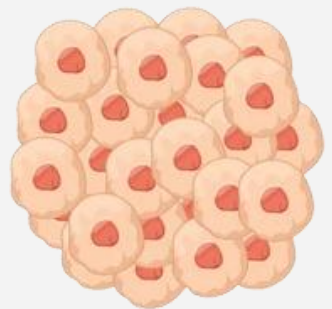
1 The methodology is scalable in term of variate nature.

2 The methodology is scalable to systems of equations.



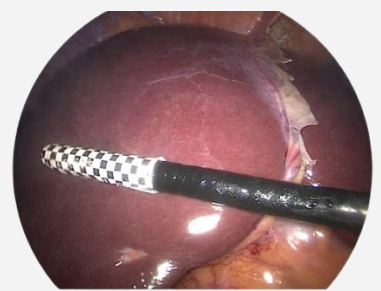
3 The methodology is scalable to non-linear equations.

4 To determine good choice in the process decisions.



5 To estimate tumour cells proliferation.

6 To estimate orientation and human motion from IMUs.



7 To represent a preoperative model as a spline.

About splines

Smoothness

$$B_{j,0} = \begin{cases} 1 & t_j < t < t_{j+1} \\ 0 & \text{otherwise} \end{cases}$$

$$B_{j,k}(x) = \frac{x - t_j}{t_{j+k-1} - t_j} B_{j,k-1}(x) + \frac{t_{j+k} - x}{t_{j+k} - t_{j+1}} B_{j+1,k-1}(x)$$

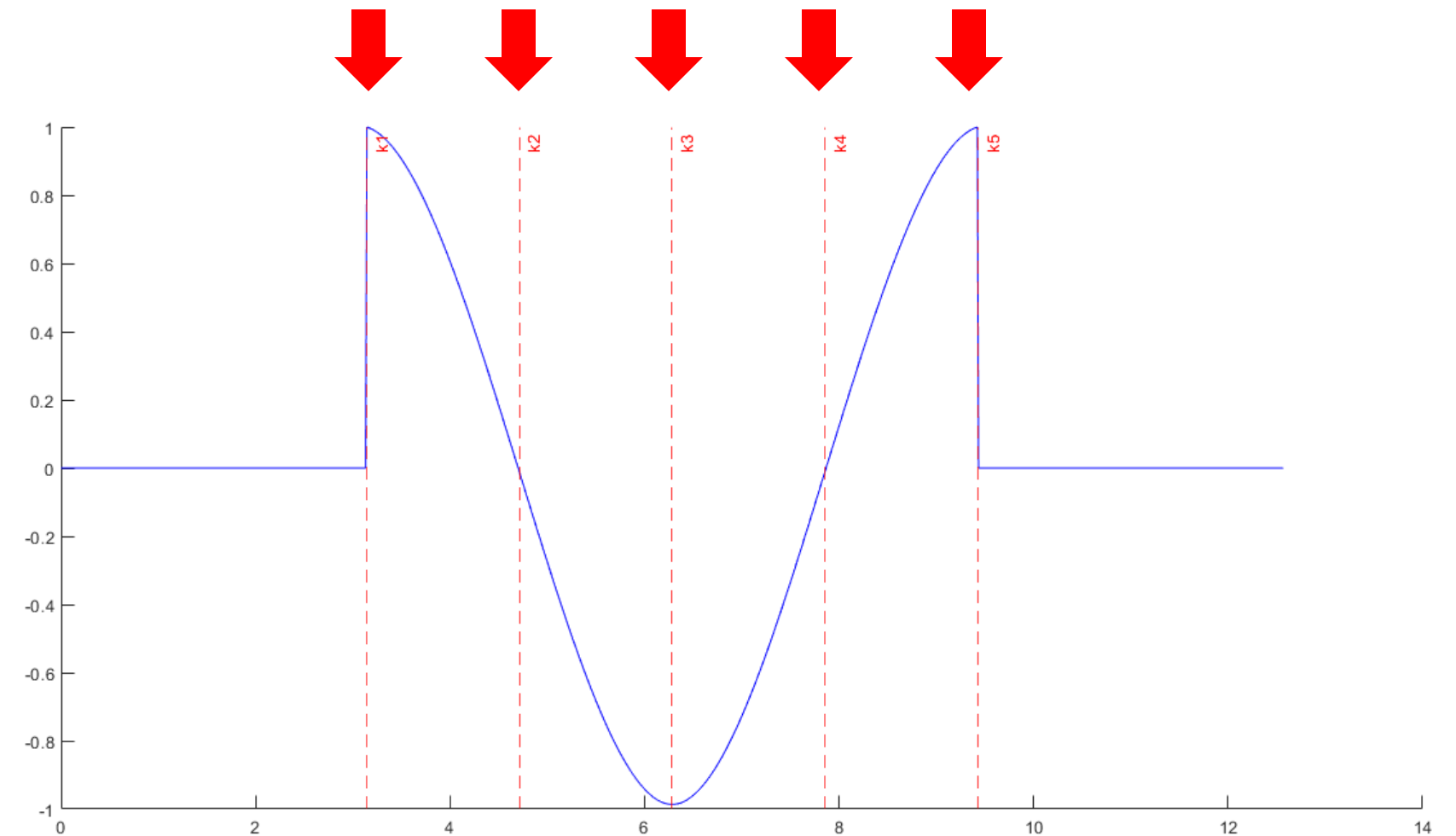
$$f(x) = \sum_{j=1}^n B_{j,k} a_j$$

Tensor product splines

$$f(x_1, \dots, x_v) = \sum_{i=1}^m \dots \sum_{j=1}^n B_{i,k} \dots B_{j,k} a_{i,\dots,j}$$

$$\mathbb{R}^n \Rightarrow \mathbb{R}^d \quad dx_1 x_2 \dots x_{n+1}$$

Knots: (k1, k1, k1, k1, k2, k3, k4, k5, k5, k5, k5)

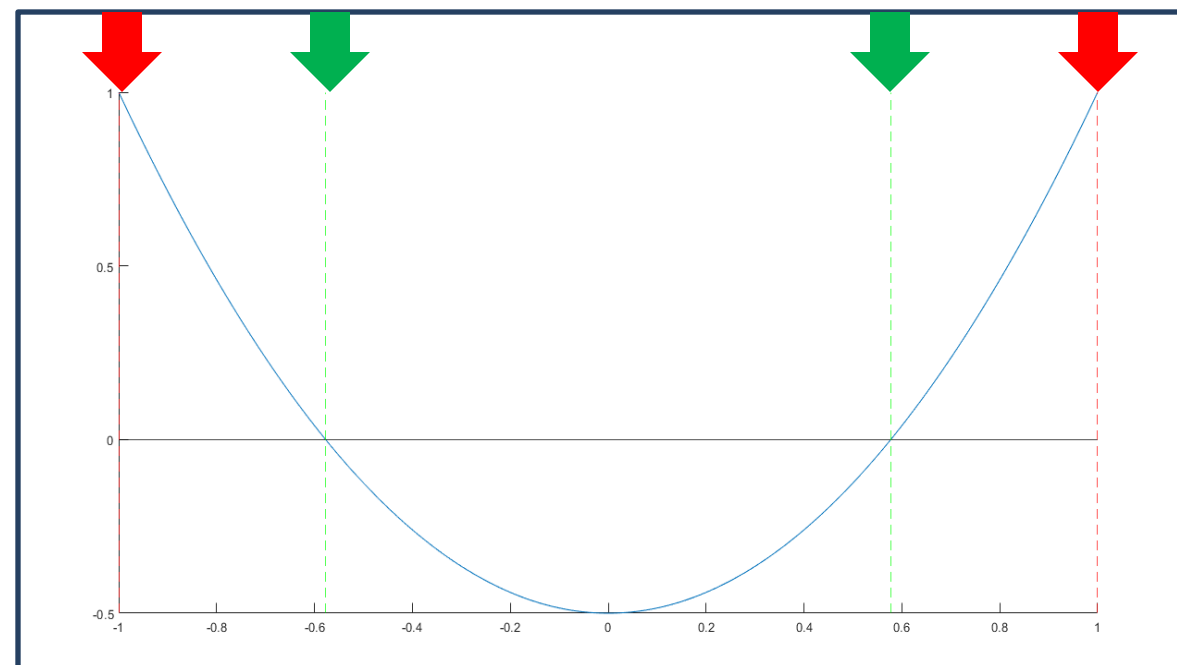


Collocation methods

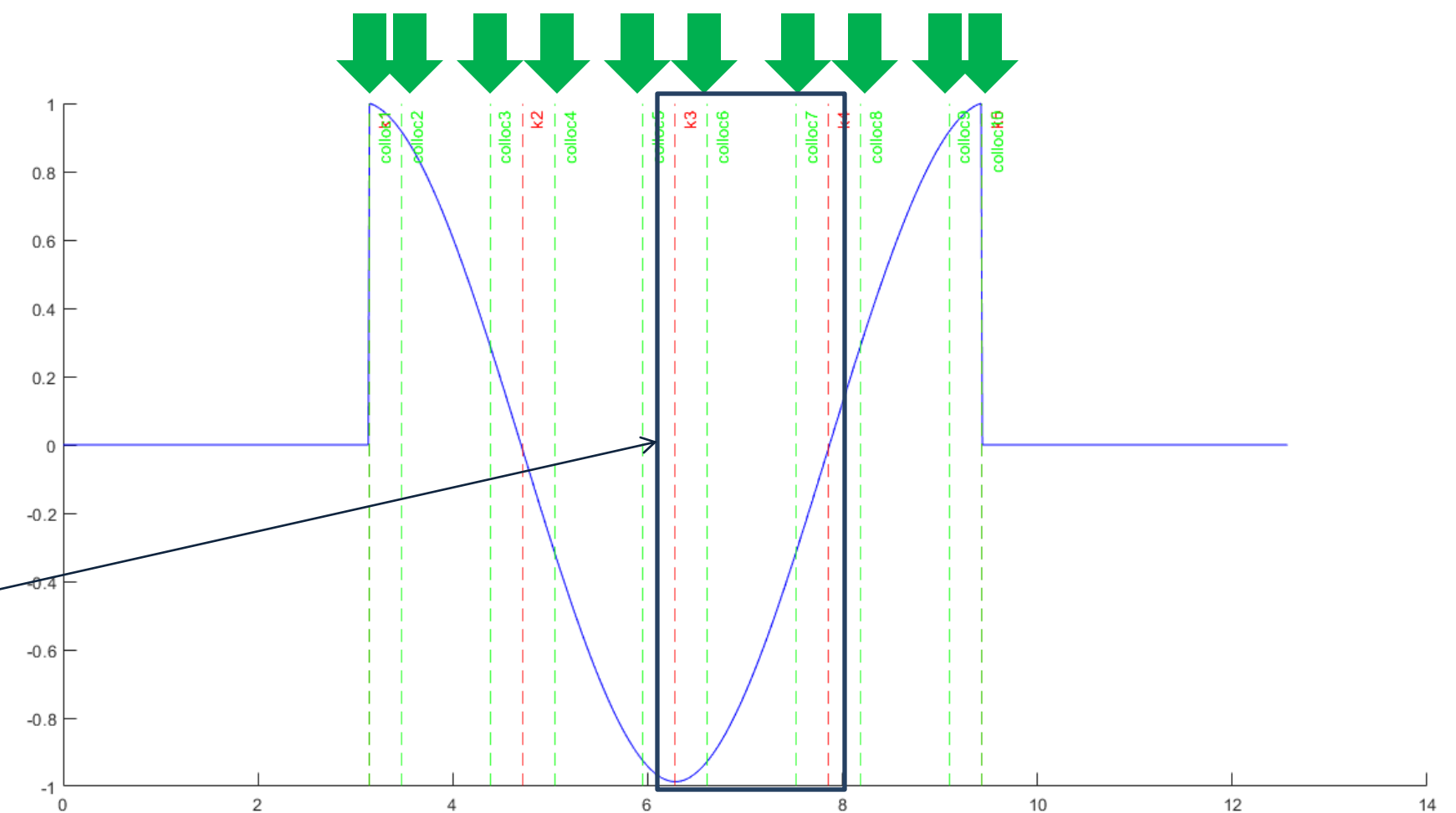
Spline design

- Order
- Knots sequence
 - Domain
 - Smoothness

Legendre polynomial



$$f(x_0, y(c_k|x_0), \frac{dy(c_k|x_0)}{dx}, \dots, \frac{d^n y(c_k|x_0)}{dx^n}) = 0$$



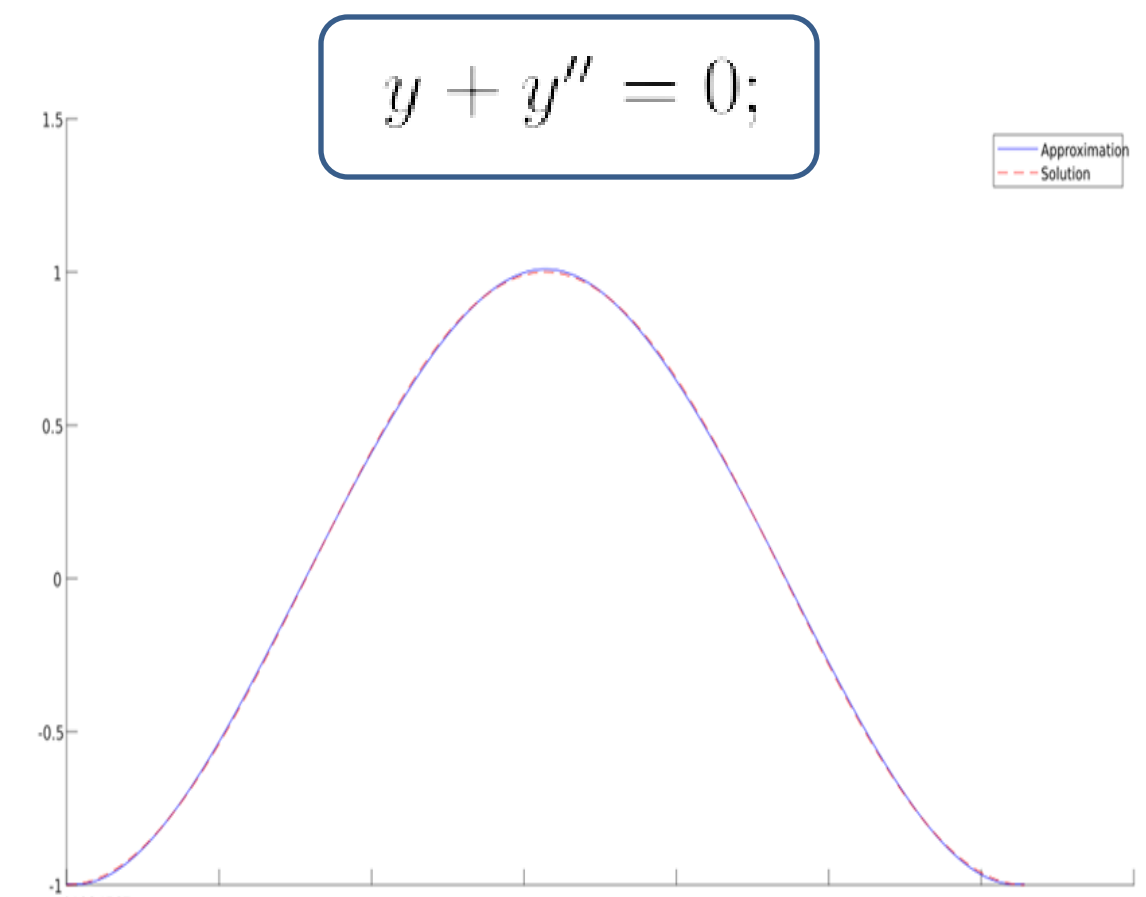
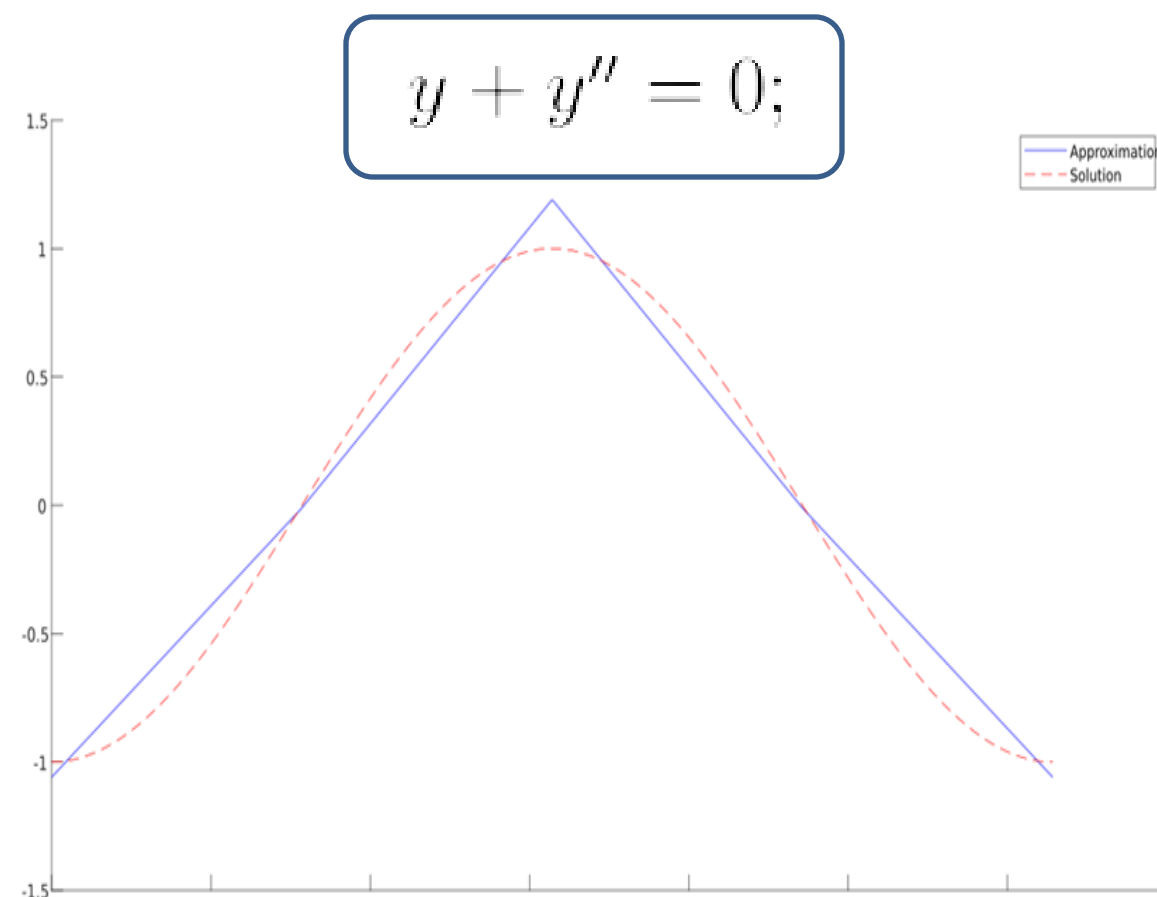
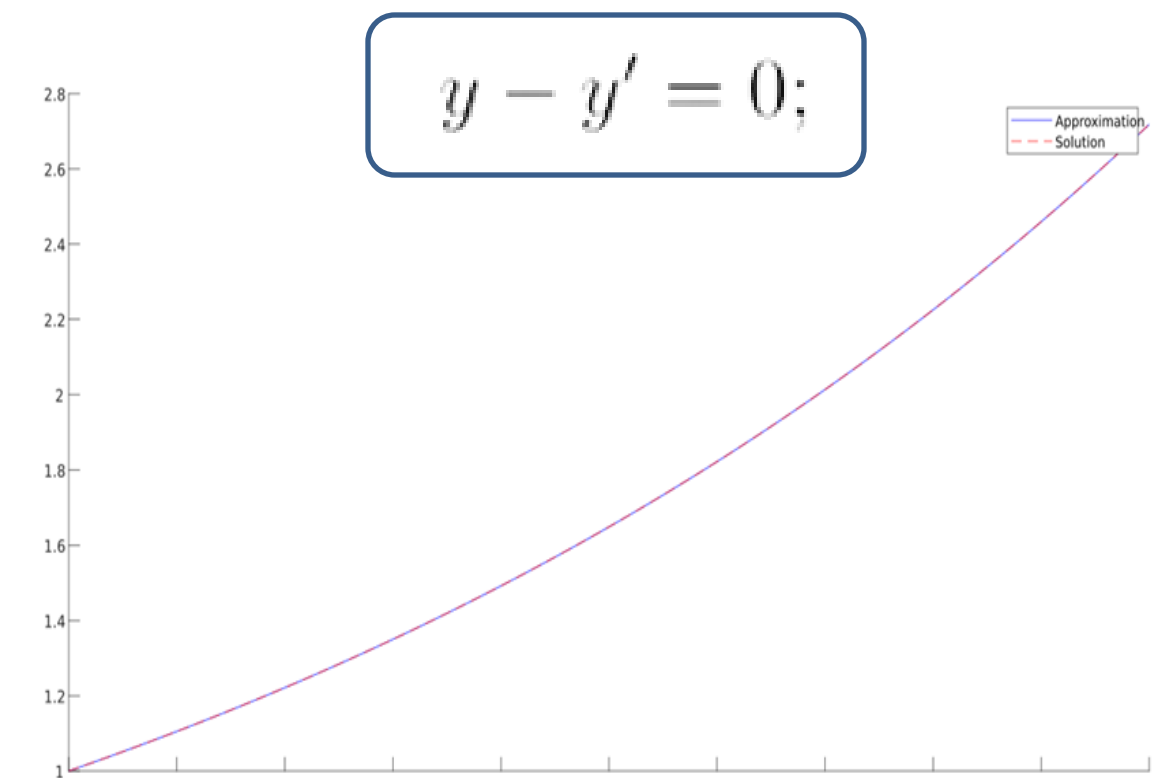
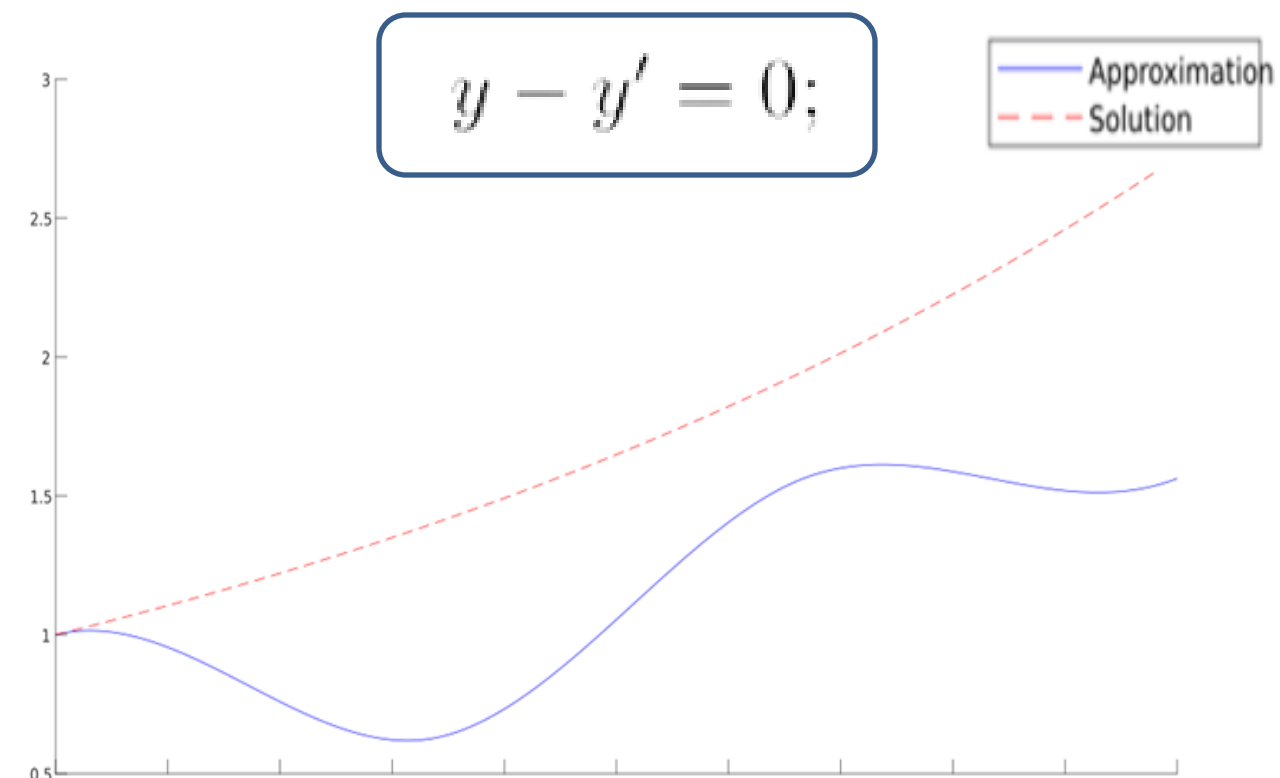
Decisions and impact

$$\begin{cases} y - y' = 0; \\ y(0) = 1 \end{cases}$$

Best choices

- Collocation points at Legendre Quadrature
- 5 breaks per sin cycle
- 4 collocation points per sub polynomial
- Sixth-order spline

$$\begin{cases} y + y'' = 0; \\ y(0) = 1 \\ y(2\pi) = 1 \end{cases}$$

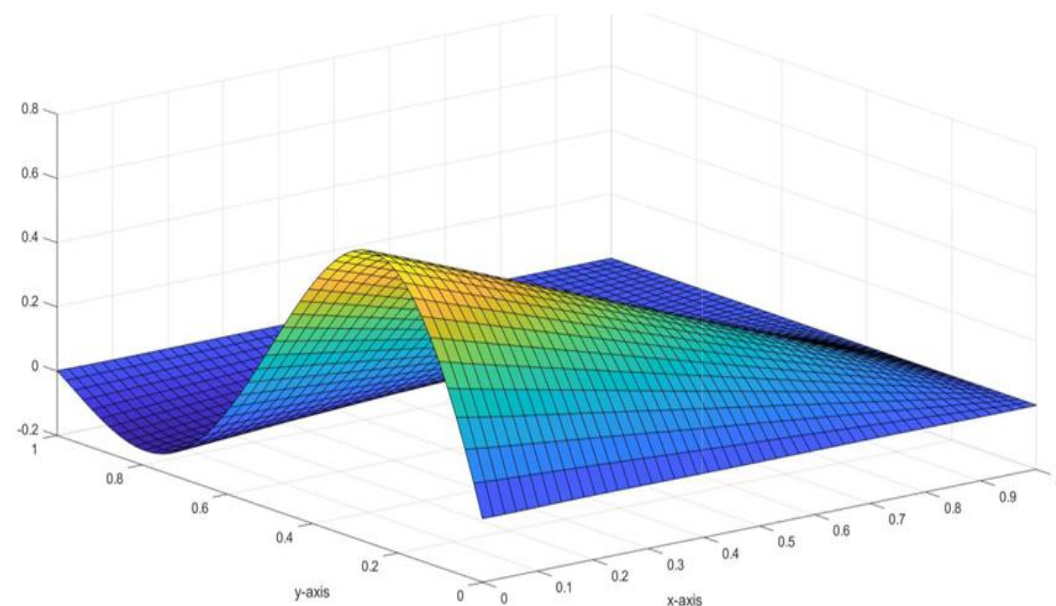


Multivariate splines

Laplace equation

$$\frac{d^2 z}{dx^2} + \frac{d^2 z}{dy^2} = f(x, y)$$

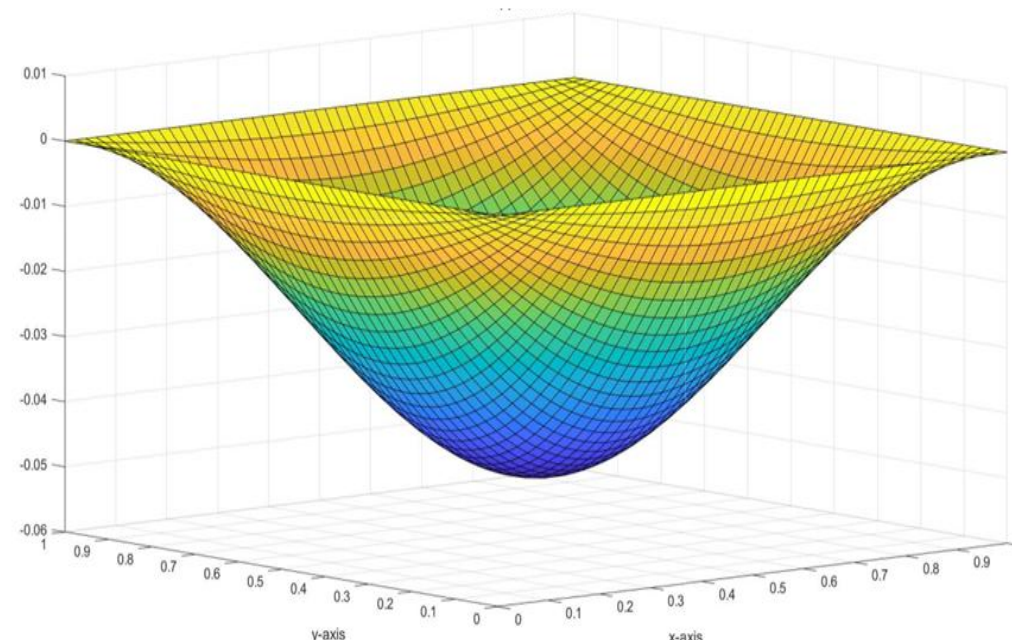
$$\left\{ \begin{array}{l} f(x, y) = 10(x-1)\cos(5y) - 25(x-1)(y-1)\sin(5y) \\ z(0, y) = (1-y)5y \\ z(1, y) = 0 \\ z(x, 0) = 0 \\ z(x, 1) = 0 \end{array} \right.$$



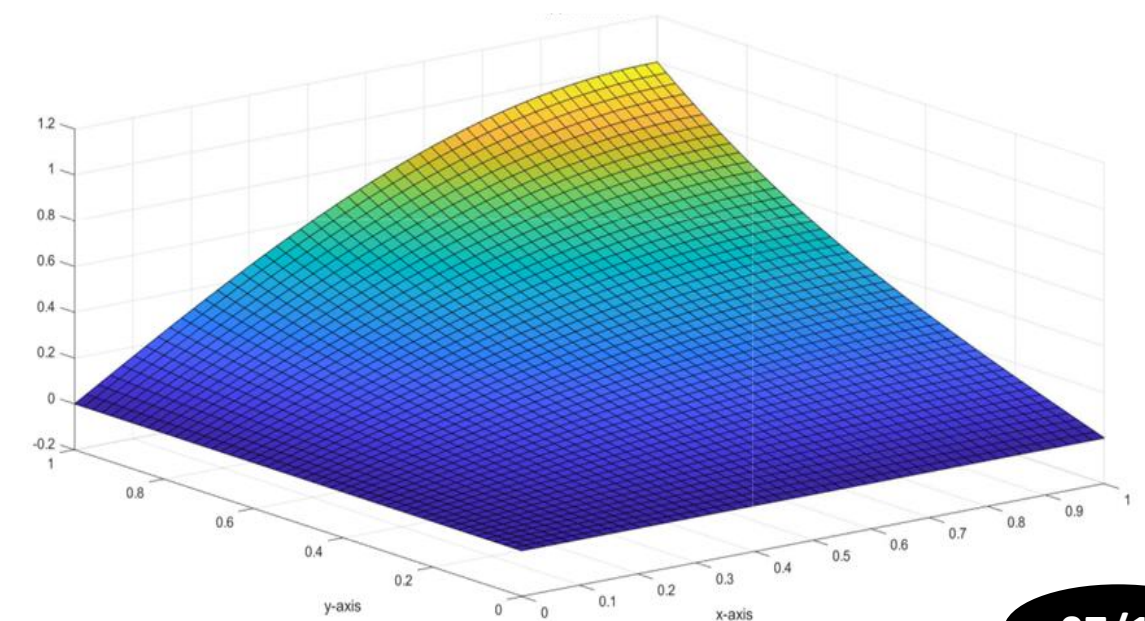
Multivariate collocation

- Cartesian grid of the collocation sites of each variable.
- Collocation as Cartesian product of the univariate collocation.

$$\left\{ \begin{array}{l} f(x, y) = f(x, y) = \sin x\pi \sin y\pi \\ z(0, y) = 0 \\ z(1, y) = 0 \\ z(x, 0) = 0 \\ z(x, 1) = 0 \end{array} \right.$$



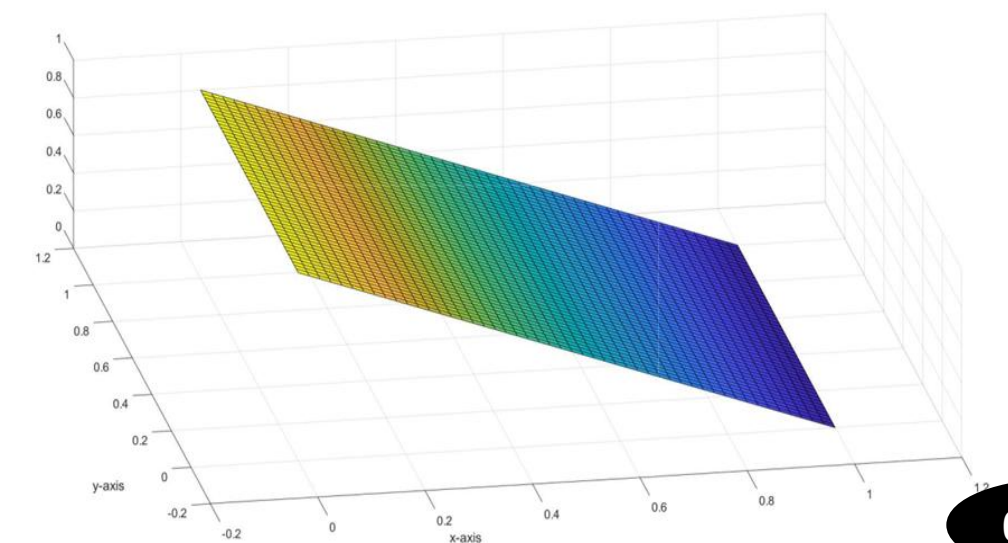
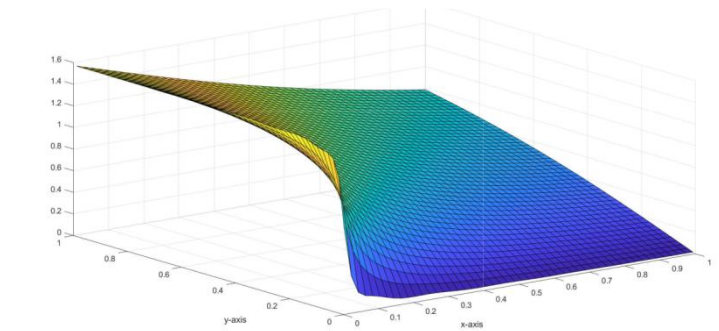
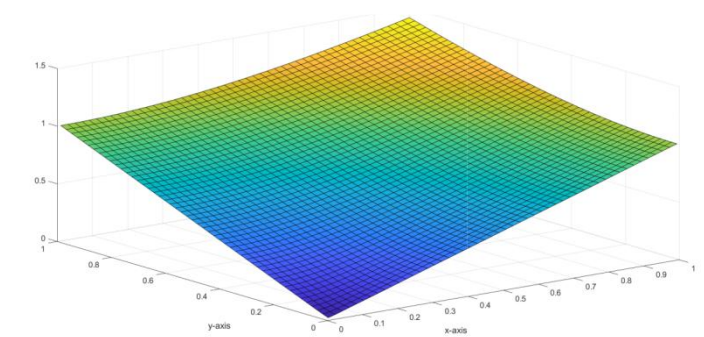
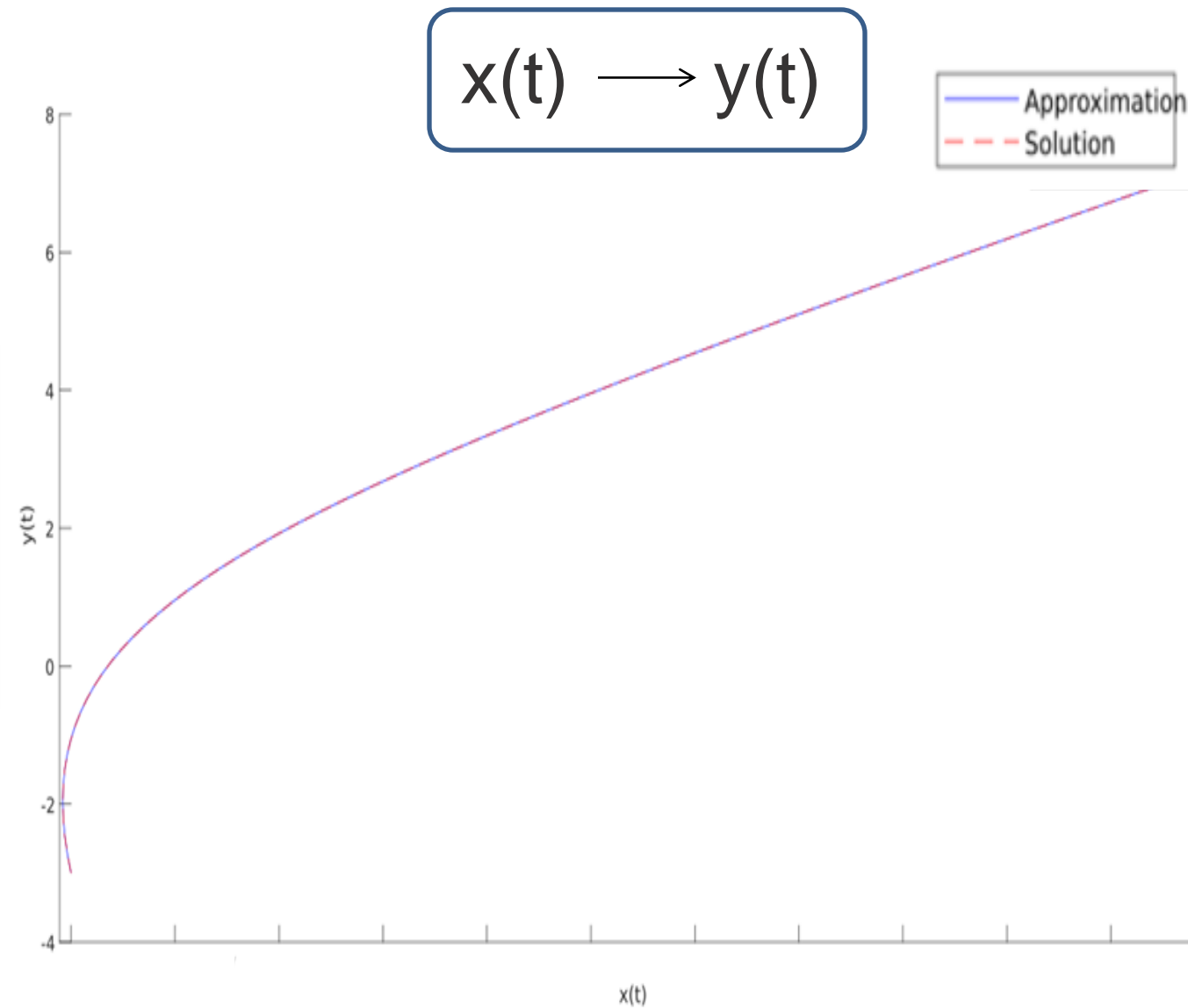
$$\left\{ \begin{array}{l} f(x, y) = 0 \\ z(0, y) = 0 \\ \frac{dz}{dx}(1, y) = 0 \\ z(x, 0) = 0 \\ z(x, 1) = \sin\left(\frac{x\pi}{2}\right) \end{array} \right.$$



Multidimensional splines

System of equations and domain outcome

$$\begin{cases} x' = x + y \\ y' = 4x - 2y \\ x(0) = 2 \\ y(0) = -3 \end{cases}$$



Non-linear DE

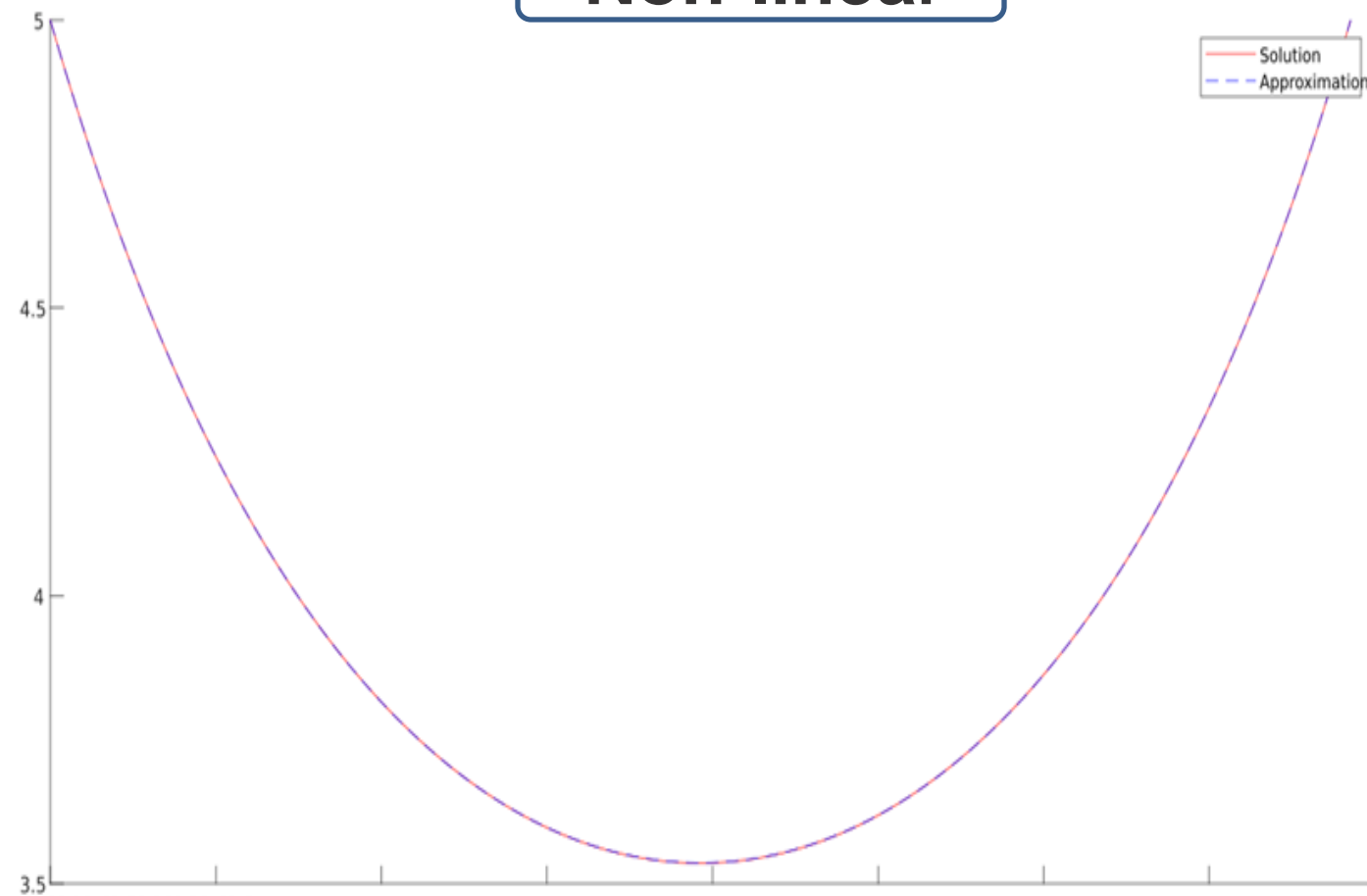
- A linear term that is one order higher than the order of the differential equation is introduced to it.
- The method consists on applying the Newton method to that equation.

Non-viscous approach

- The viscosity term is set to zero.

$$\varepsilon y'' = \frac{50}{(1 + \sin 2x)^2} - (y')^2 - y^2$$

Non-linear

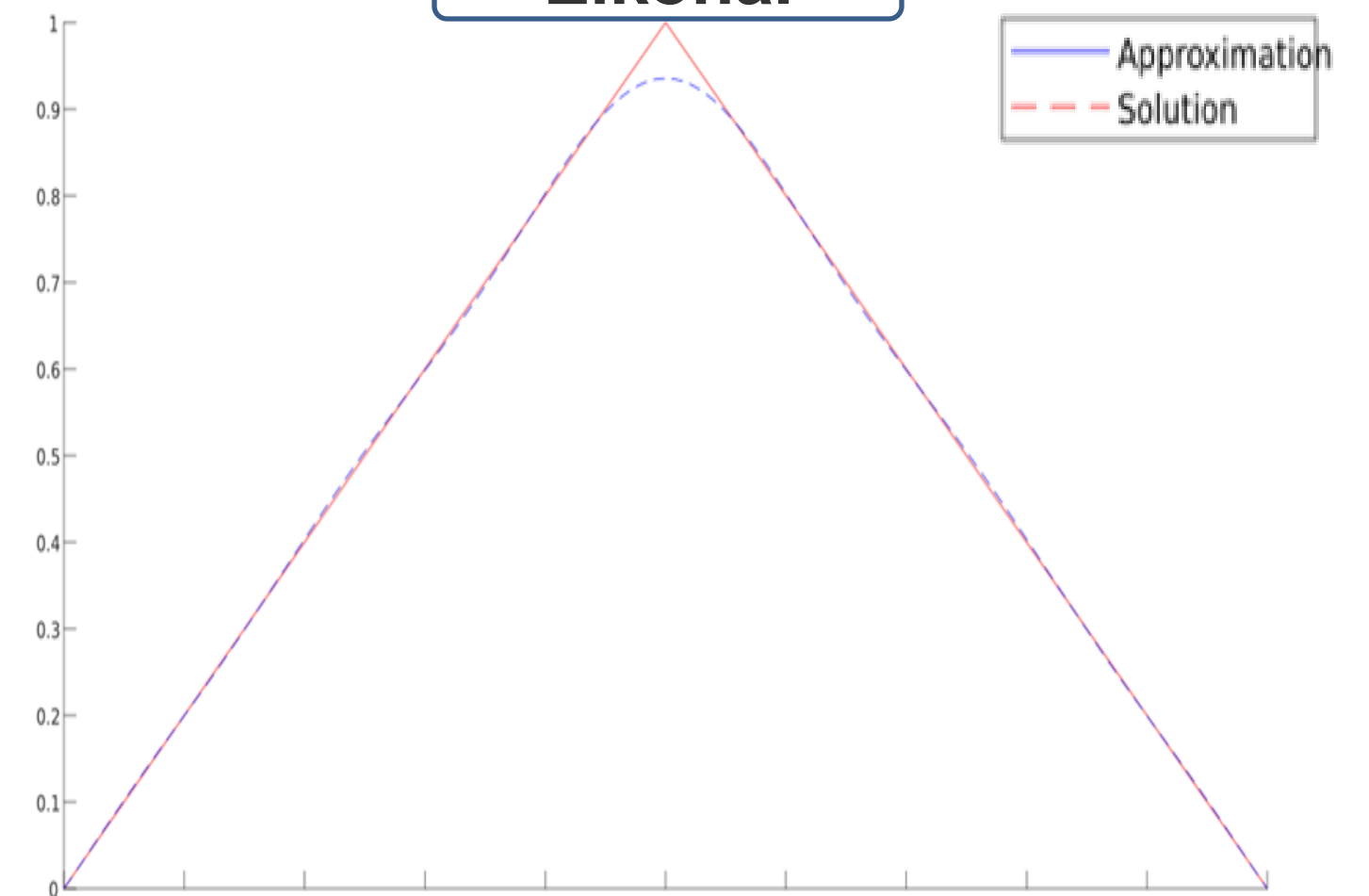


Viscous approach

- The viscosity term is set.

$$\varepsilon y'' = (y')^2 - 1$$

Eikonal



Gompertz model

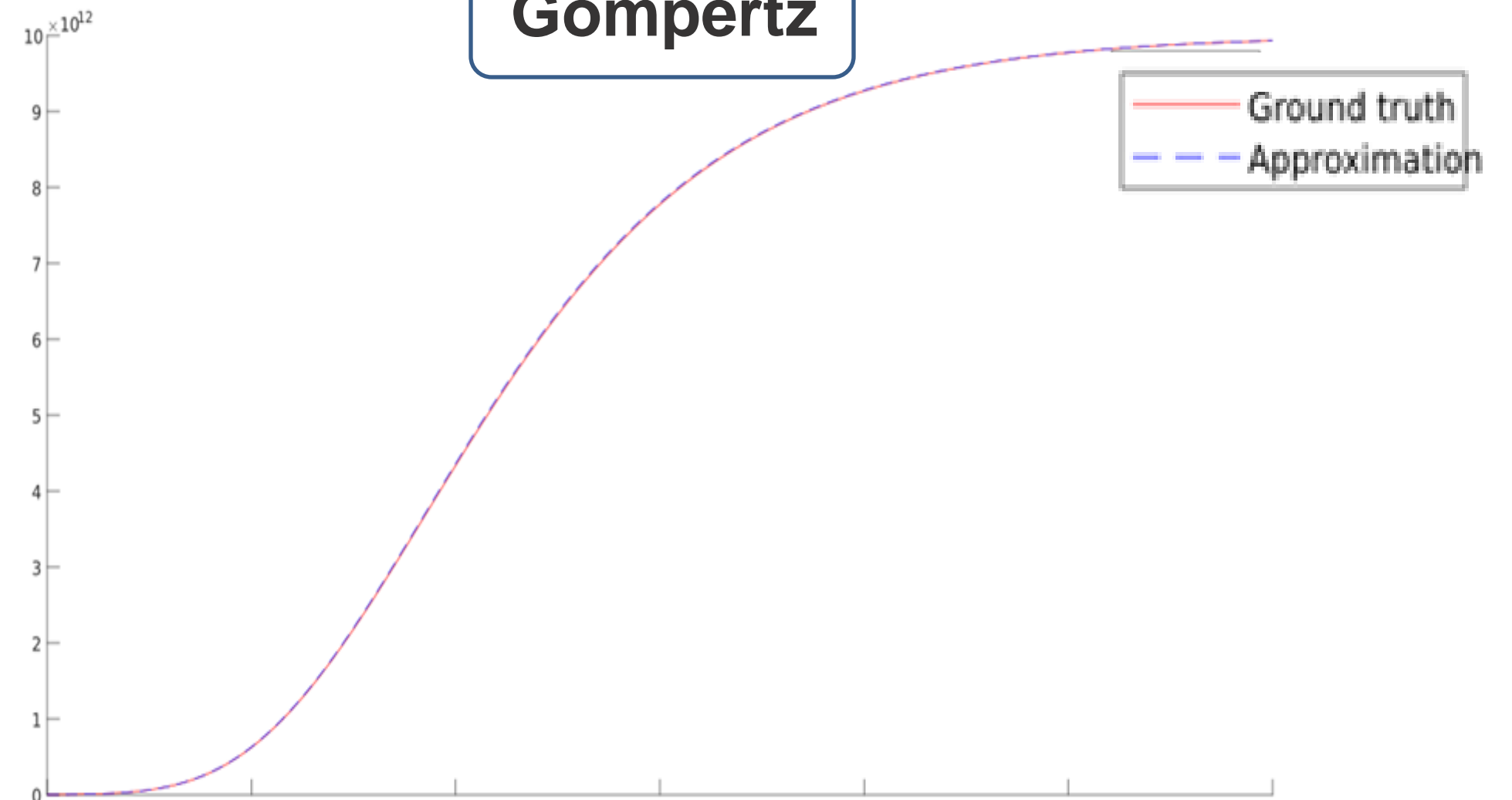
Estimate the number of cells depending on the time of a tumour in angiogenesis stage.

It depends on the initial number of cells, the intrinsic growth rate and the carrying capacity.

$$N'(t) = rN \log\left(\frac{C}{N}\right)$$

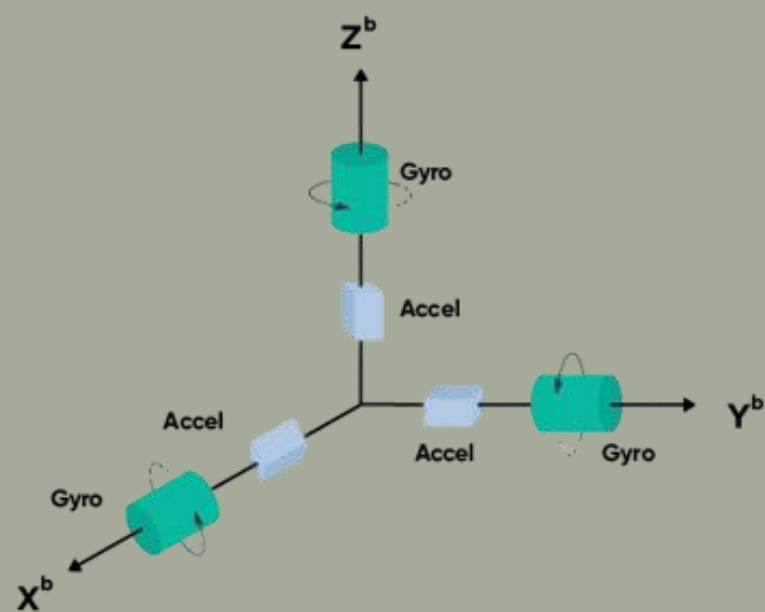
Parameters

- $N(0) = 1_{10}9$ cells
- $r = 0.006$ cells/t
- $C = 1_{10}13$ cells



Motion monitoring

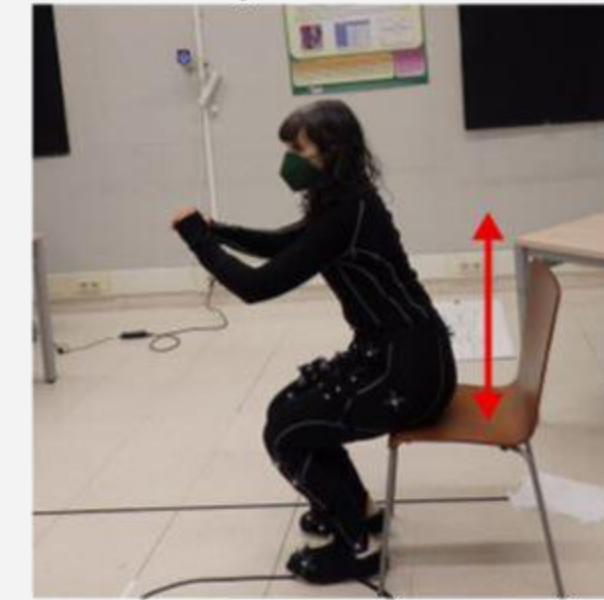
Inertial Measurement Unit (IMU):



Leg exercises



KFE

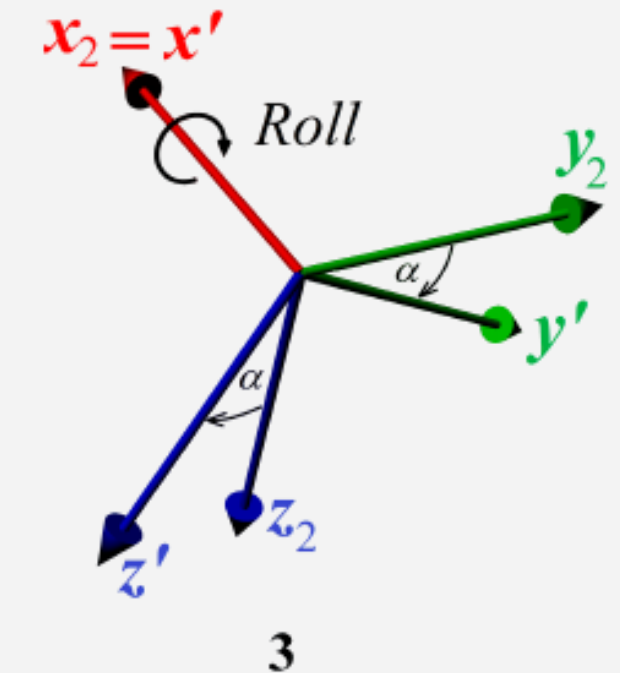
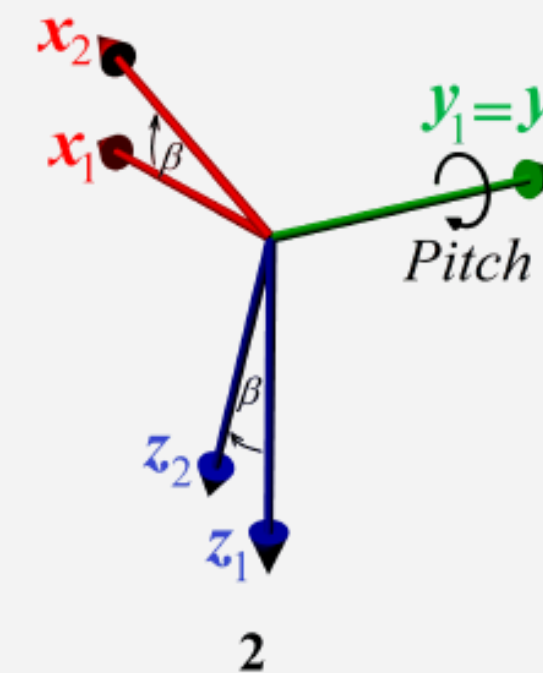
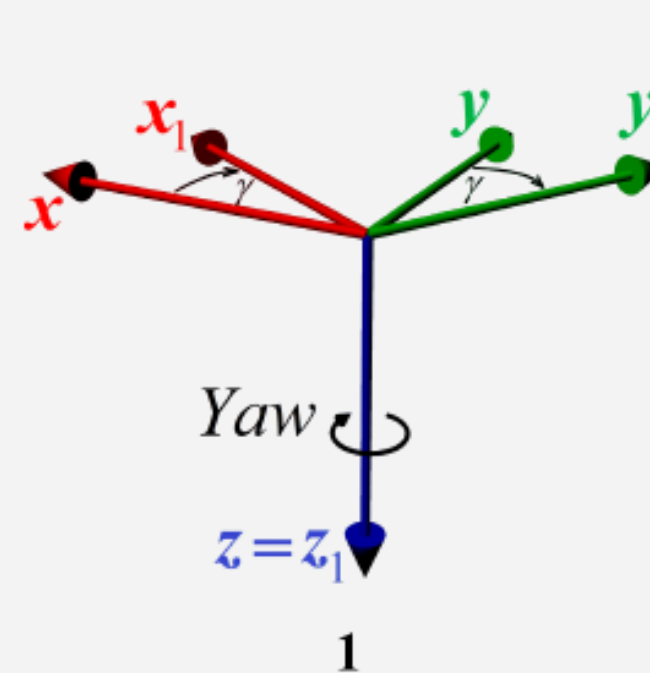


SQT



HAA

Euler angles:



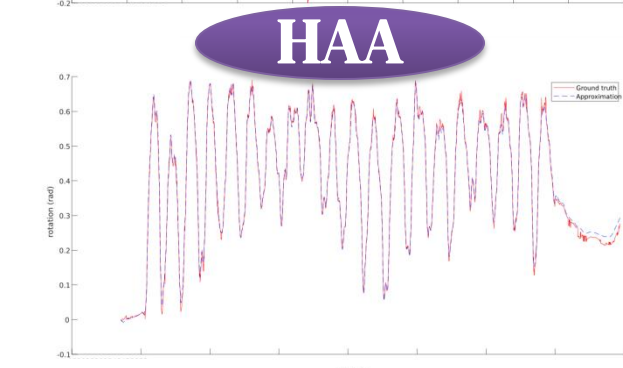
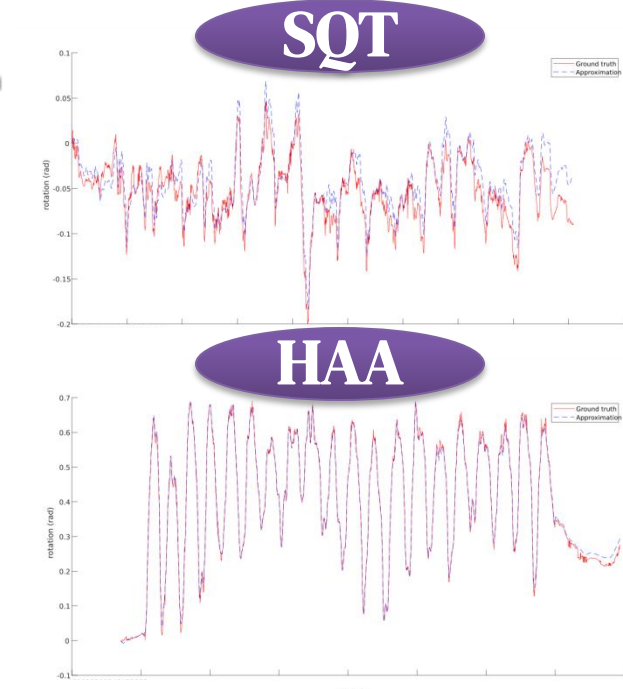
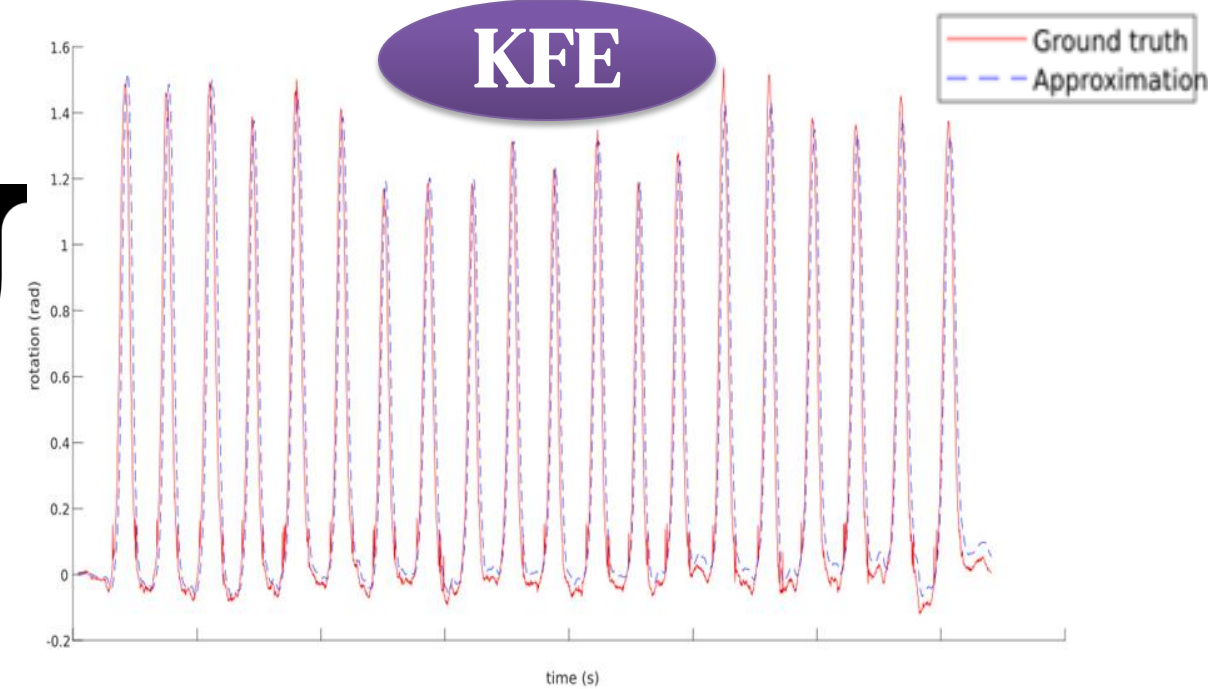
Orientation results

$$\vec{\omega} = \begin{bmatrix} 1 & 0 & -\sin \theta \\ 0 & \cos \psi & \sin \psi \cos \theta \\ 0 & -\sin \psi & \cos \psi \cos \theta \end{bmatrix} \begin{pmatrix} \psi' \\ \theta' \\ \phi' \end{pmatrix}$$

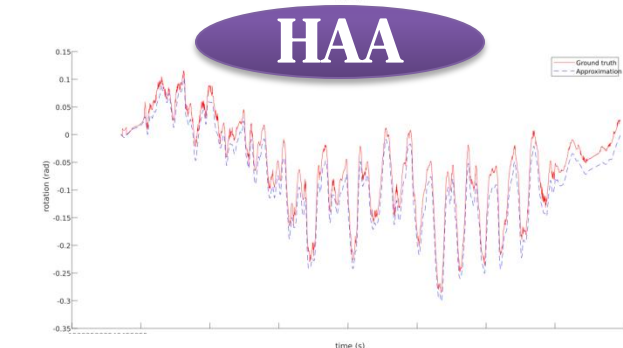
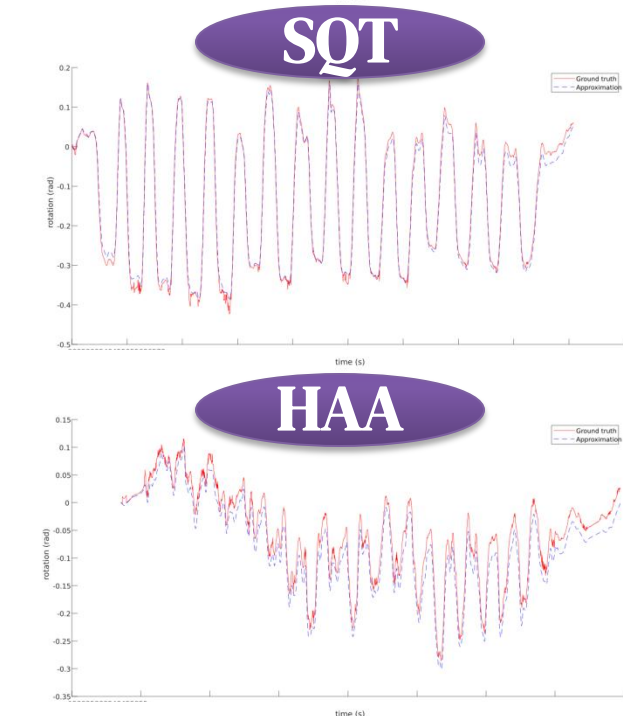
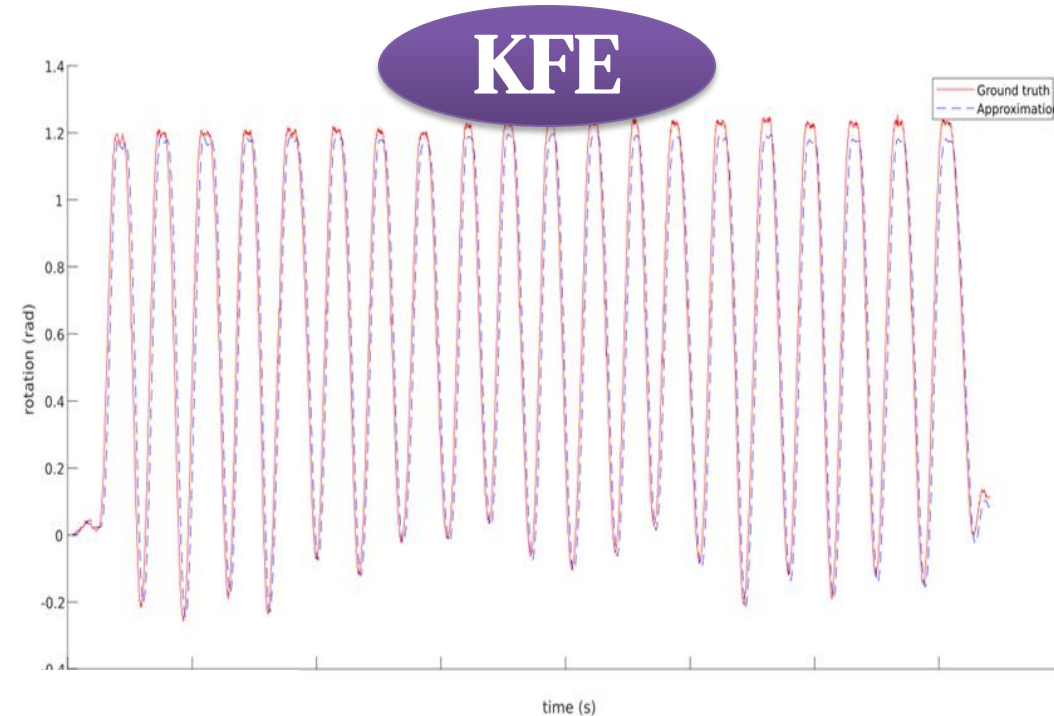
Table 5.18: Error metrics of KFE orientation.

Angle (rad)	Mean error	Max error	Std error
Yaw (ψ)	0.12	0.57	0.12
Pitch (θ)	0.11	0.30	0.07
Roll (ϕ)	0.16	0.55	0.12

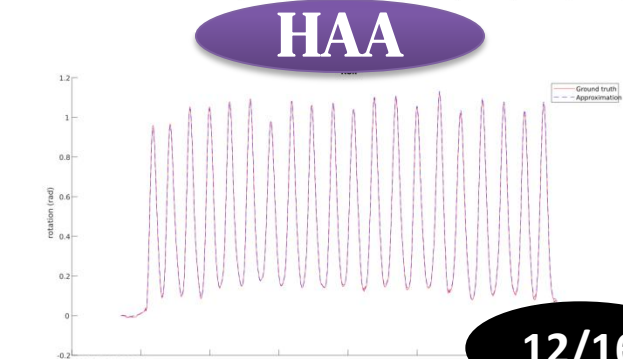
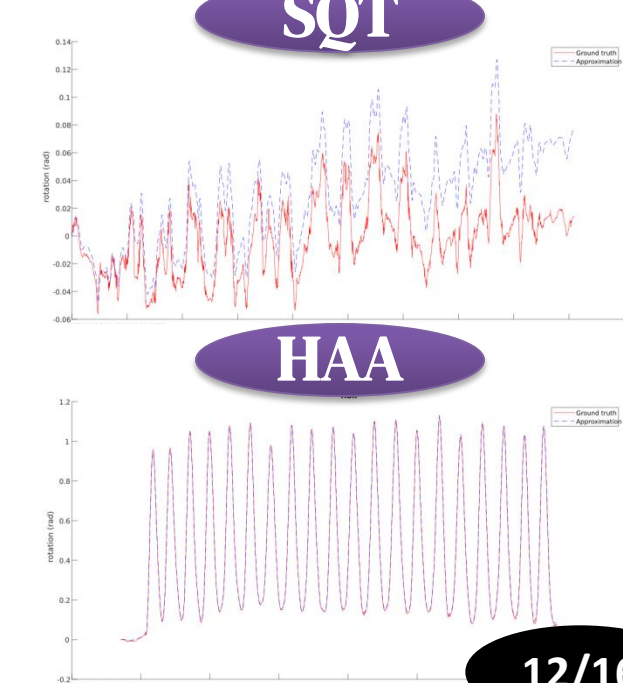
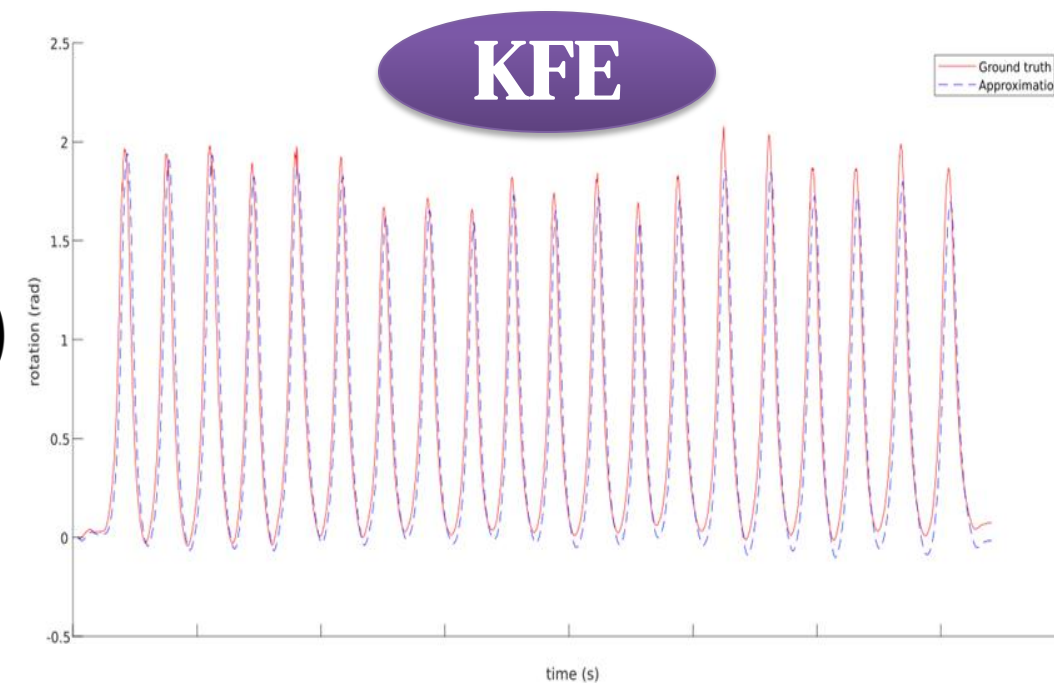
Ψ



Θ

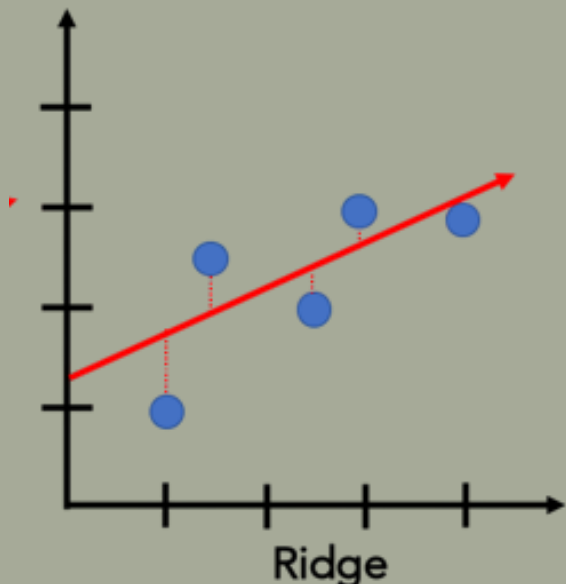


Φ



Motion results

$$\begin{aligned} a(t) \cdot M_z(\psi, \theta, \phi) - \frac{d^2z(t)}{dt^2} &= 0 \\ a(t) \cdot M_y(\psi, \theta, \phi) - \frac{d^2y(t)}{dt^2} &= 0 \\ a(t) \cdot M_x(\psi, \theta, \phi) - \frac{d^2x(t)}{dt^2} &= 0 \end{aligned}$$



$$M = \begin{bmatrix} \cos \psi \cos \theta & \sin \psi \cos \theta & -\sin \theta \\ -\sin \psi \cos \theta + \cos \psi \sin \theta \sin \phi & \cos \psi \cos \theta + \sin \psi \sin \theta \sin \phi & \cos \theta \sin \phi \\ \sin \psi \sin \theta + \cos \psi \sin \theta \cos \phi & -\sin \psi \sin \theta + \sin \psi \sin \theta \cos \phi & \cos \theta \cos \phi \end{bmatrix}$$

Table 5.19: Error metrics of KFE linear displacement.

Axis (m)	Mean error	Max error	Std error
Z	0.03	0.12	0.02
Y	0.01	0.04	0.01
X	0.05	0.10	0.02

Z

Y

X

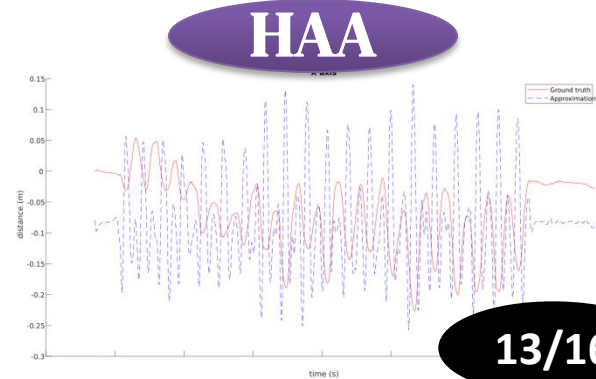
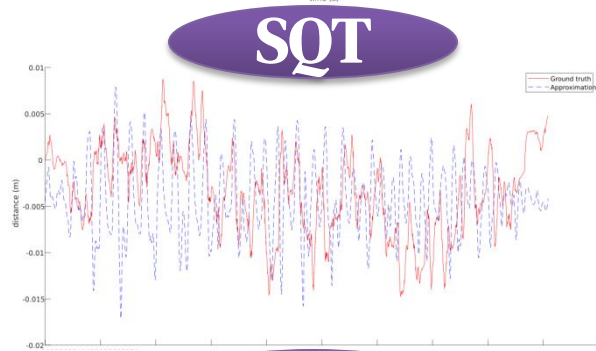
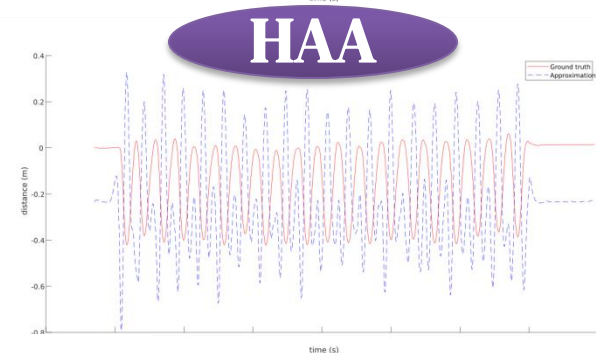
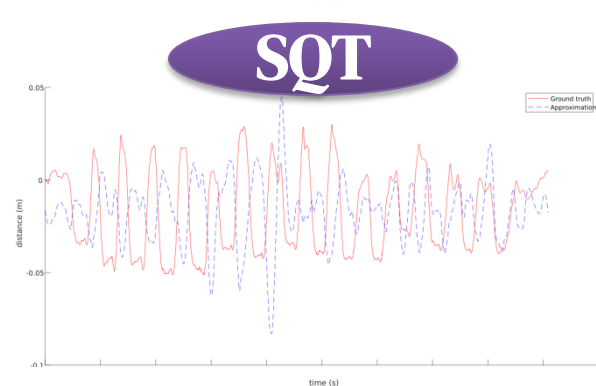
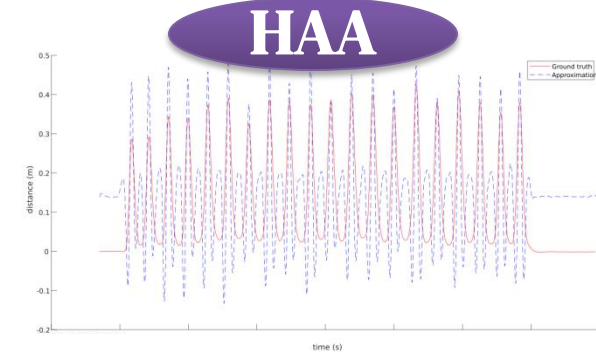
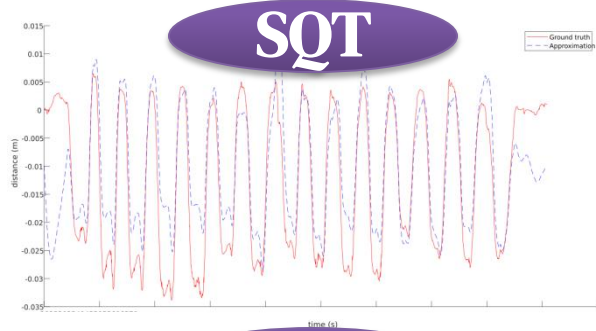
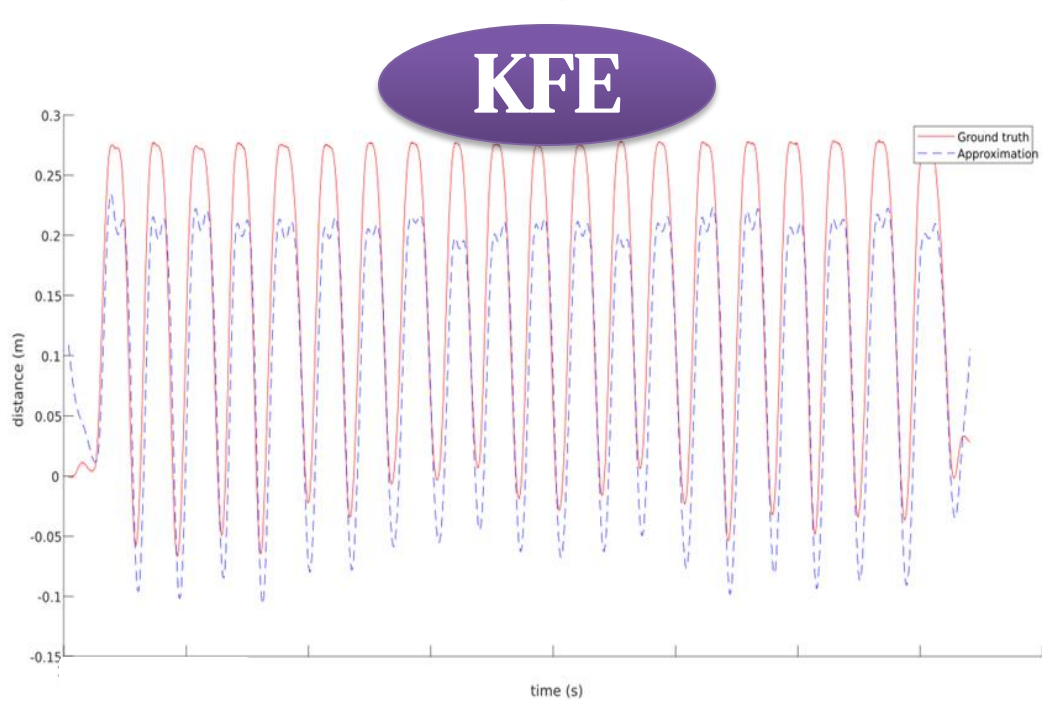
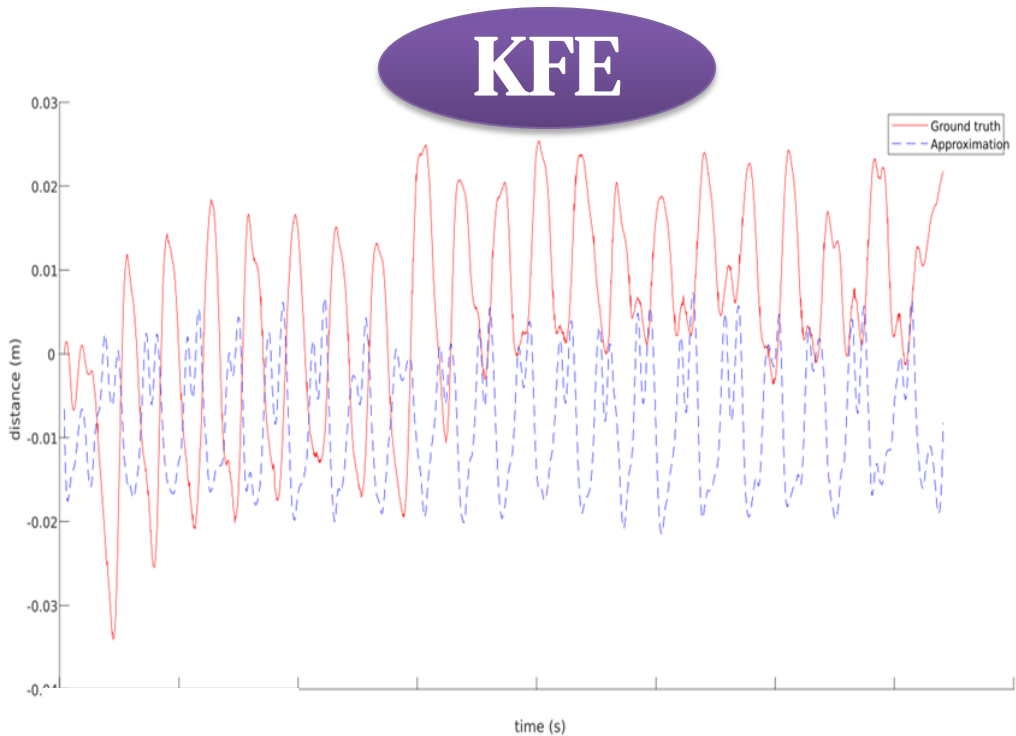
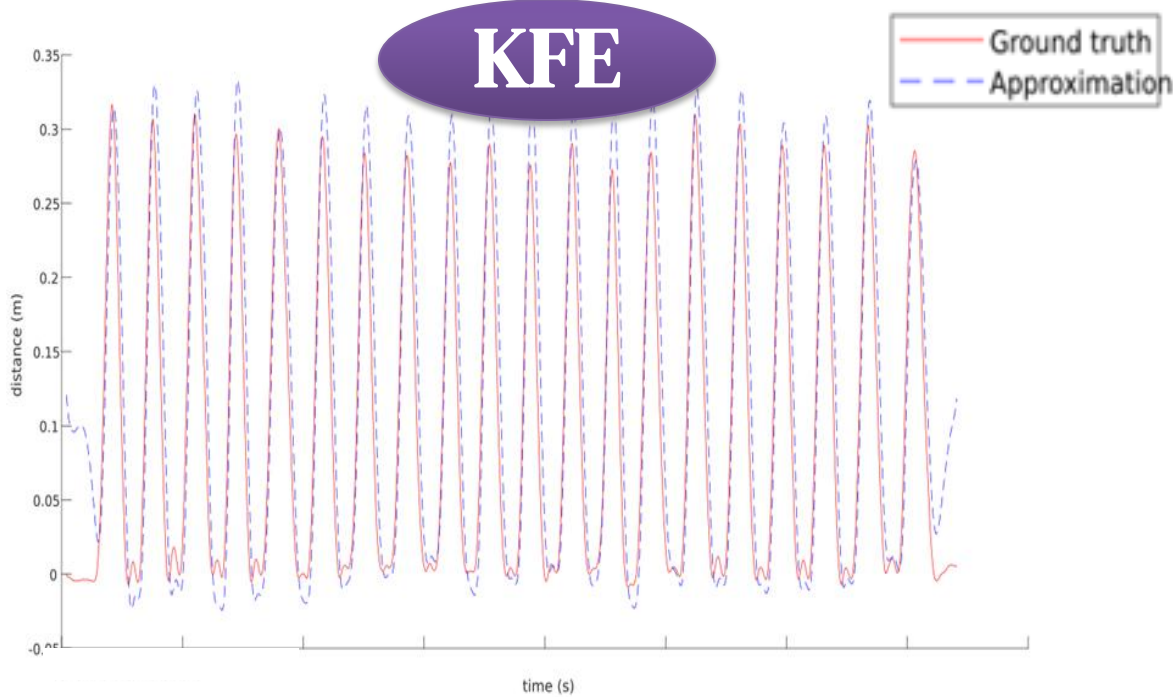


Image processing

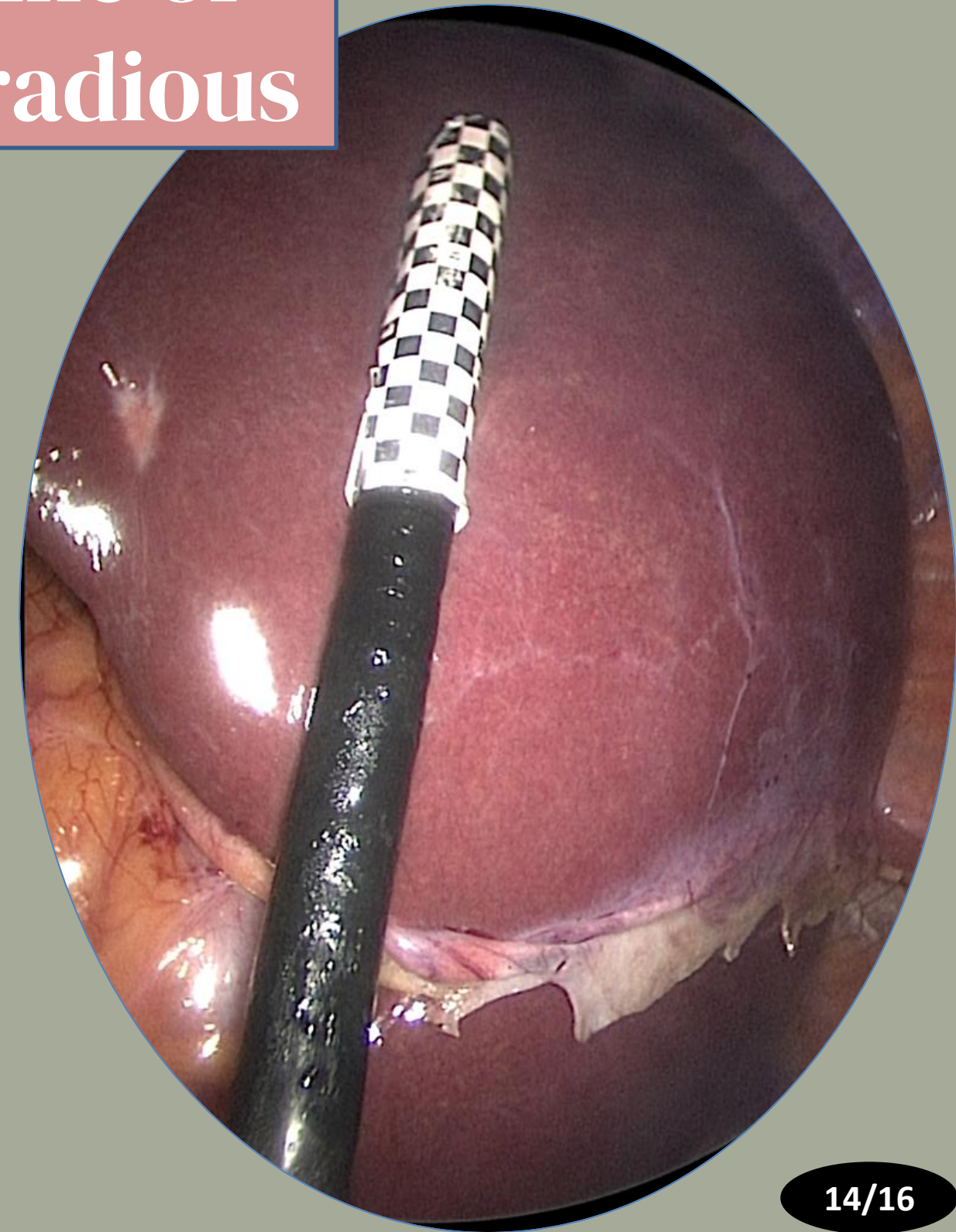
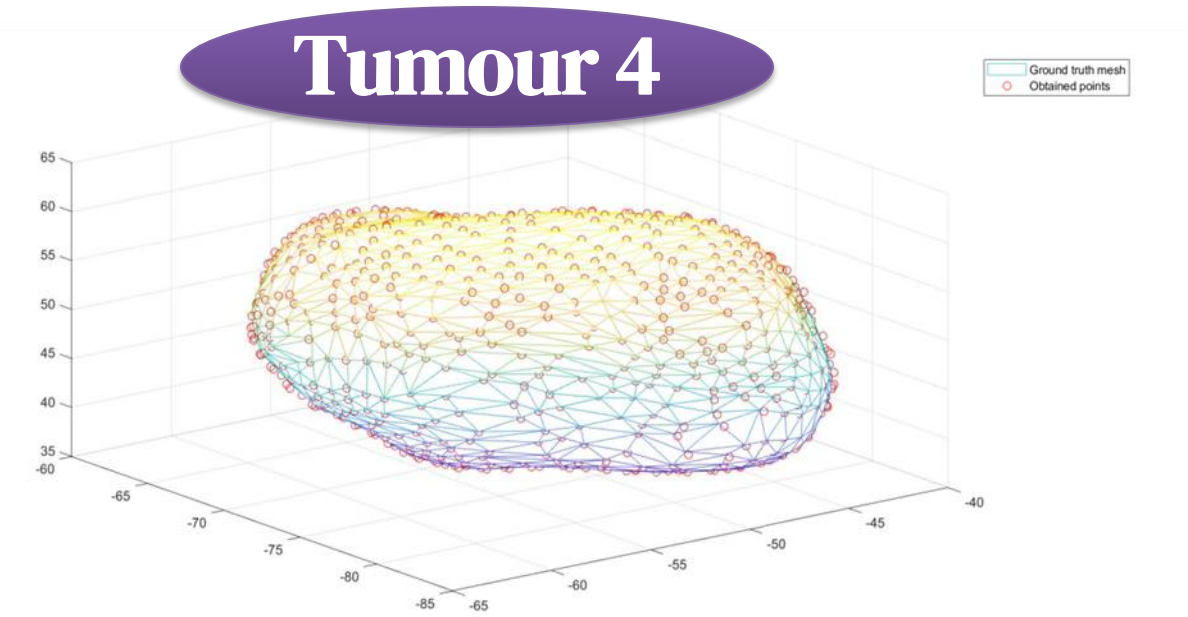
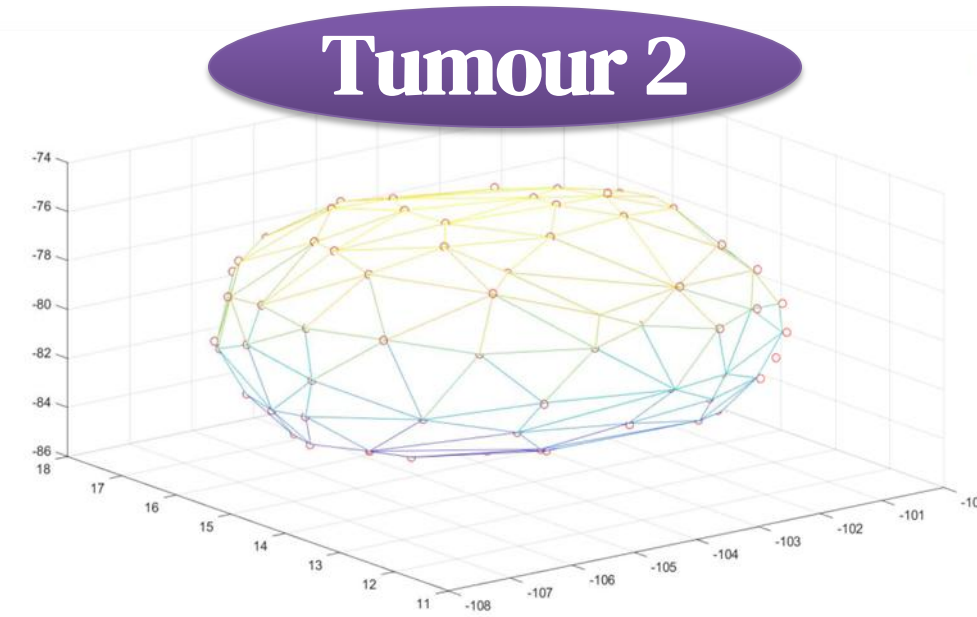
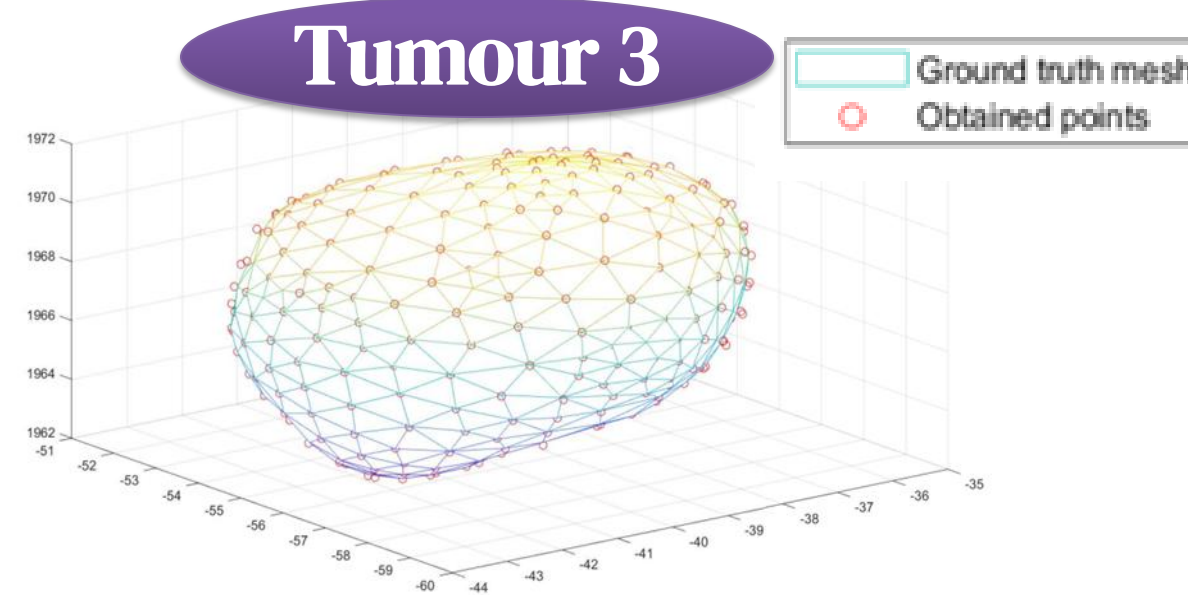
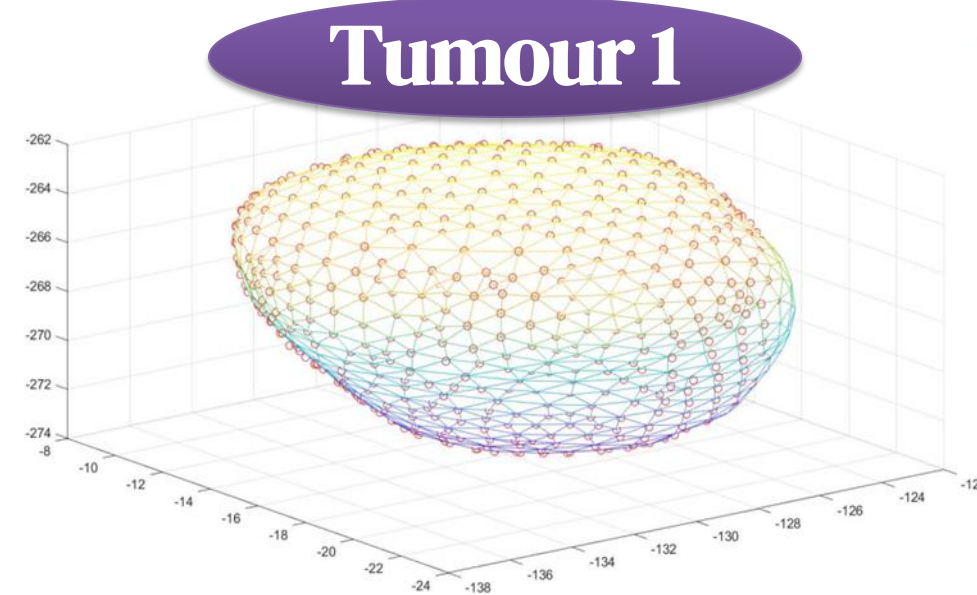
Preoperative
model



Spherical
coordinates



Spline of
the radius



Conclusions and SDGs

- 1 The method is scalable to multivariate function, systems and non-linear equations.
- 2 It is determined a good criteria for different choices including the order, the collocation sites, the viscosity, etc.
- 3 The estimation of the orientation is satisfactory and the human motion in the KFE exercise.
- 4 The preoperative model of tumours are successfully represented as differentiable splines.

SDGs: Goal 3: Good health and well-being

Non-communicable
diseases

Universal health
coverage

Health risk
management

Future lines

A1

We must evaluate the motion estimation including the time requirements.

A2

It is needed to improve the determination of the ridge parameter in the displacement estimation.



B

To provide a close path framework is needed for the circular bases domain in the preoperative tumour models.

Open path



Closed path

