



## Final Project

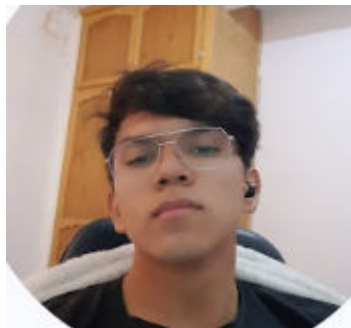
Departamento de Ingeniería Eléctrica y Electrónica

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### General Information



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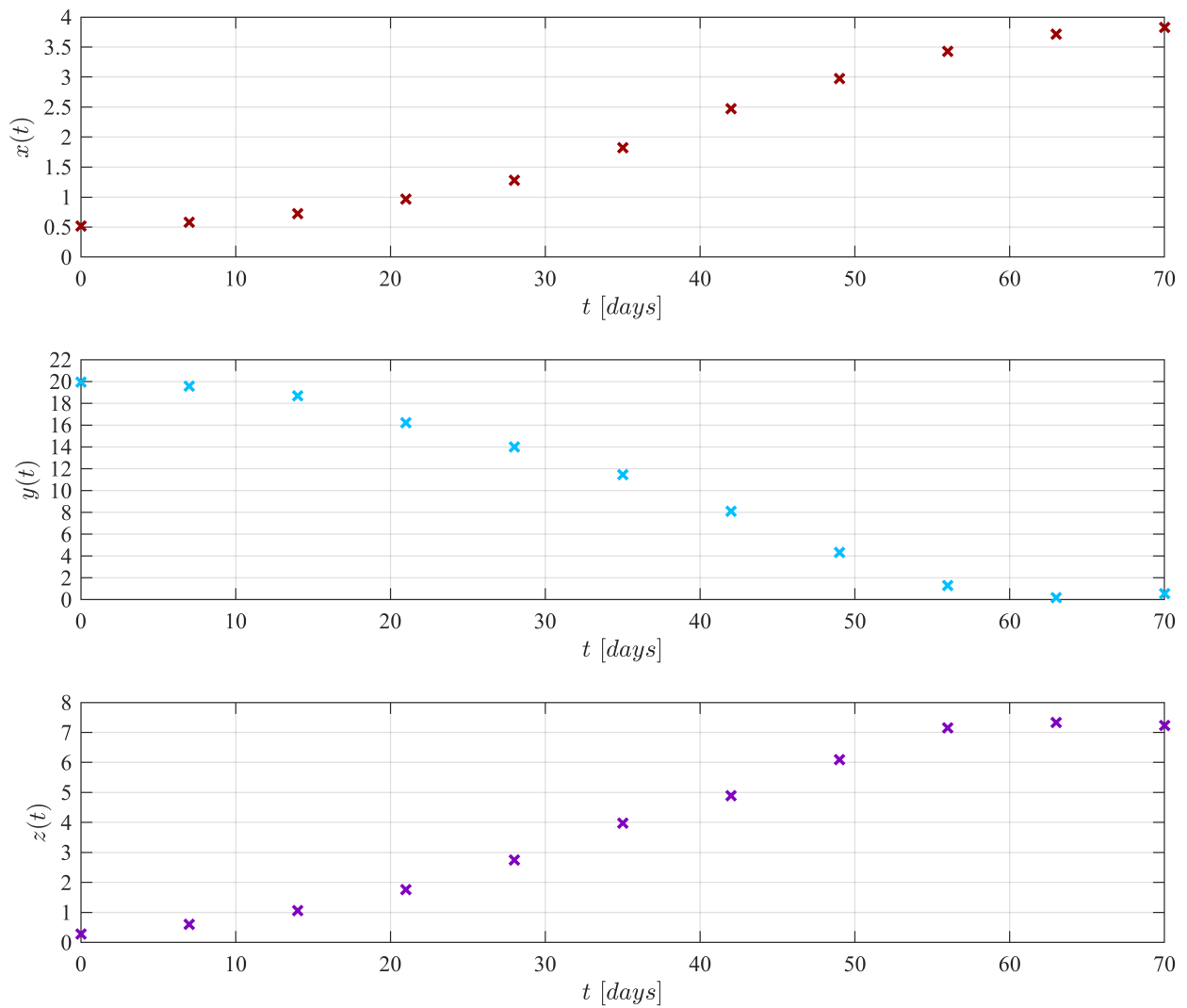
**Objective:** To integrate specific competencies related to the modeling, analysis, and control of biological systems, in order to illustrate and predict their dynamics in both the short and long term. The goal is to lay the groundwork for the development of digital twins capable of addressing and solving problems within the field of Systems Biology.

## Simulation Data

```
clc; clear; close all; warning('off', 'all');  
filename = 'data.csv';  
sys = readmatrix(filename);  
t = sys(:,1);  
x = sys(:,2);  
y = sys(:,3);  
z = sys(:,4);  
T = array2table([t, x, y, z], 'VariableNames', {'tiempo (days)', 'x', 'y', 'z'});  
disp(T);
```

tiempo (days)	x	y	z
0	0.517	19.939	0.281
7	0.58	19.578	0.609
14	0.726	18.678	1.063
21	0.965	16.215	1.758
28	1.282	14.005	2.752
35	1.825	11.467	3.975
42	2.473	8.096	4.895
49	2.974	4.327	6.096
56	3.43	1.287	7.152
63	3.717	0.194	7.328
70	3.829	0.56	7.232

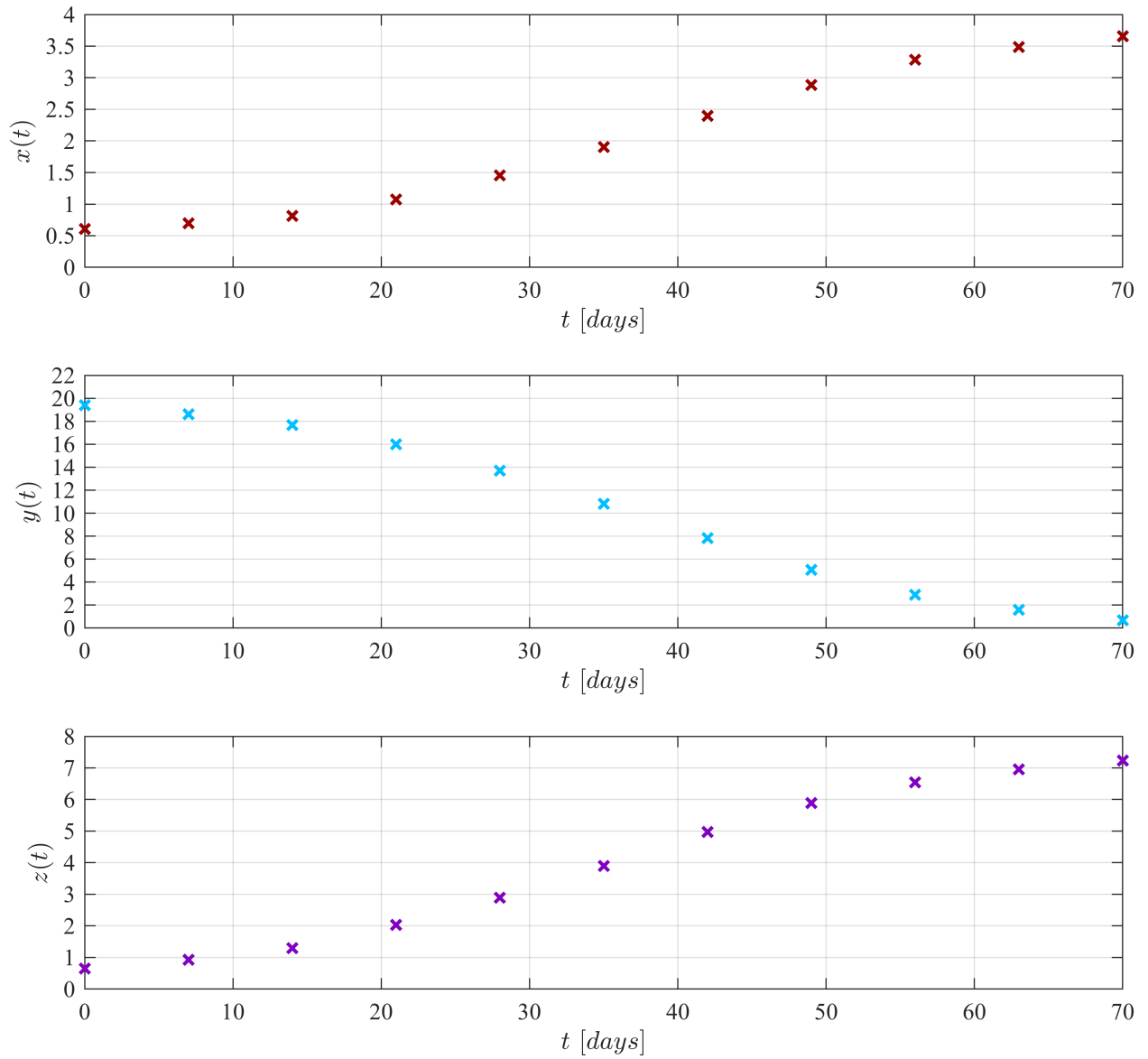
```
plotdata(t, x, y, z);
```



## Smooth Data

```
clc; clear; close all; warning('off','all')
filename = 'Data.csv';
sys = readmatrix(filename);
t= sys(:,1);
x= sys(:,2);
y= sys(:,3);
z= sys(:,4);
plotDataS(t, x, y, z);
```

Data smoothed exported as: "data\_fit.csv".



## Nonlinear Regression

The equations considered for the approach of the model are the following:

$$D(x, T) = 0.00122xy + 0.000557xyz$$

$$D(y, T) = 0.0133y - 0.0281xy$$

$$D(z, T) = 0.00201z + 0.00328yz$$

```
warning('off')
```

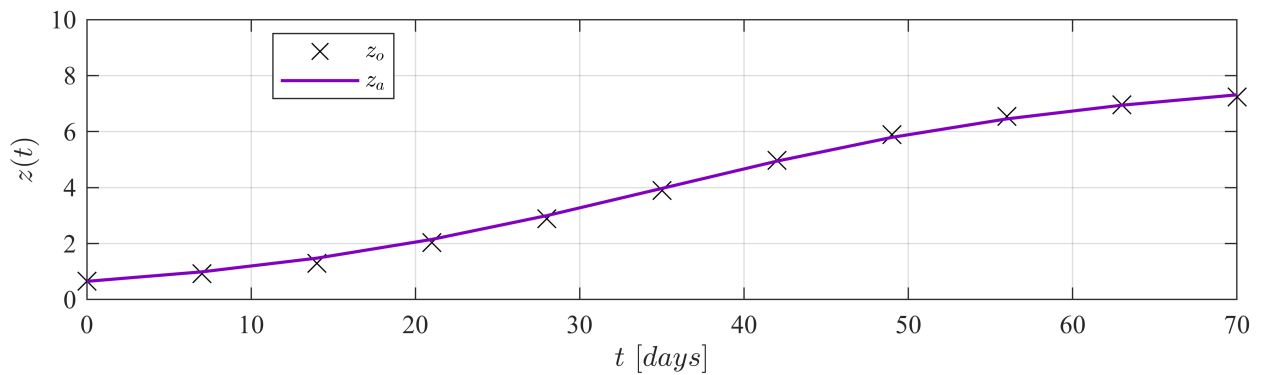
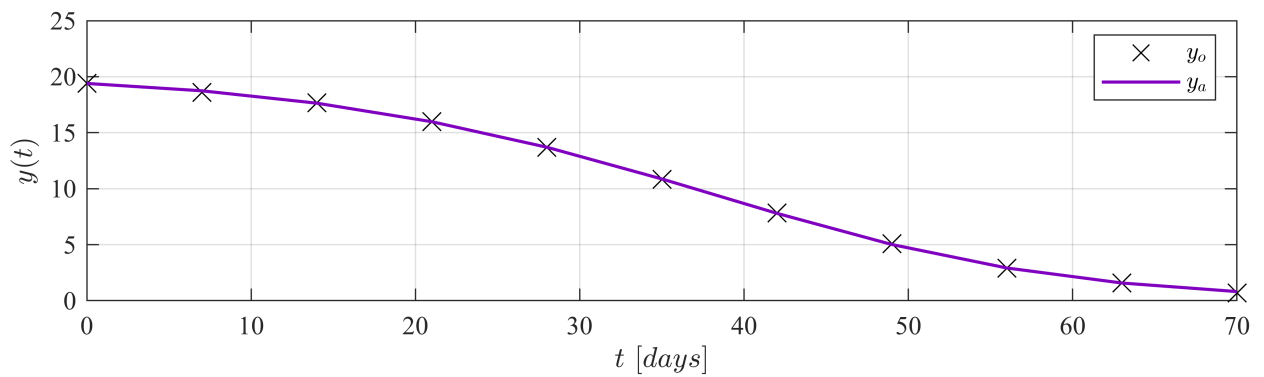
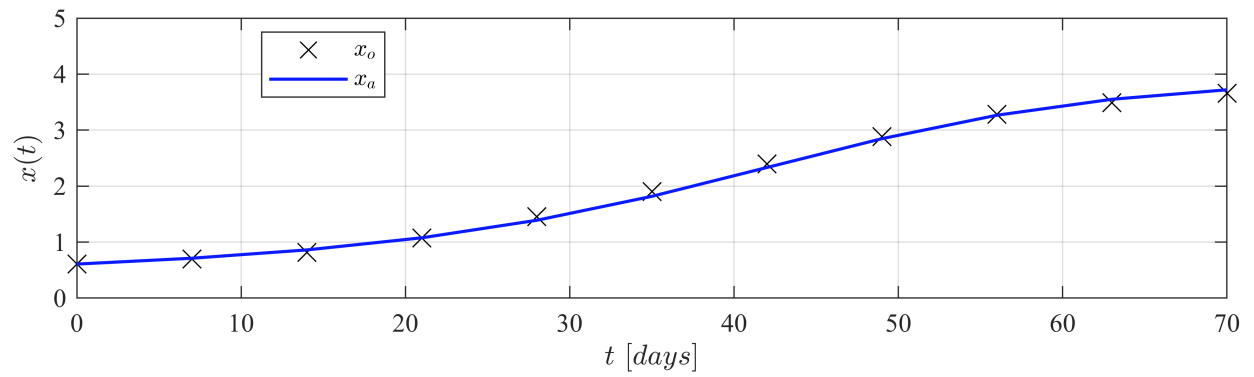
```

sys = readmatrix('data_fit.csv');
t = sys(:,1);
x = sys(:,2);
y = sys(:,3);
z = sys(:,4);
P0 = [0.0012; 0.00055; 0.013; 0.028; 0.0020; 0.0033];
[mdl, xa, ya, za] = Varied(t, x, y, z, P0);

```

Sample size (n): 11  
 Parameters to be estimated: 6  
 Degrees of freedom: 27  
 R-squared: 0.9998  
 Corrected AIC: -65.2787

Param	Estimate	SE	MoE	CI95		pValue
"p1"	0.0006015	9.2161e-05	0.0001891	0.00041241	0.0007906	5.3482e-07
"p2"	0.00072444	3.8939e-05	7.9896e-05	0.00064454	0.00080434	6.3191e-17
"p3"	0.014804	0.00042563	0.00087331	0.013931	0.015677	5.8103e-24
"p4"	0.030033	0.00055695	0.0011428	0.02889	0.031176	4.956e-29
"p5"	0.0040202	0.00067335	0.0013816	0.0026386	0.0054018	2.2873e-06
"p6"	0.0029262	5.5581e-05	0.00011404	0.0028122	0.0030402	9.4077e-29



## Jacobian Matrix

```
clear; close all; clc
syms x y z p1 p2 p3 p4 p5 p6 real
dx = p1*x*y + p2*x*y*z;
dy = p3*y - p4*x*y;
dz = p5*z + p6*y*z;
J = jacobian([dx, dy, dz], [x, y, z]);
disp('Jacobian matrix of the system:'); disp(J);
```

Jacobian matrix of the system:

$$\begin{pmatrix} p_1 y + p_2 y z & p_1 x + p_2 x z & p_2 x y \\ -p_4 y & p_3 - p_4 x & 0 \\ 0 & p_6 z & p_5 + p_6 y \end{pmatrix}$$

## Equilibrium points

```
clear; close all; clc
syms x y z p1 p2 p3 p4 p5 p6 real
dx = p1*x*y + p2*x*y*z;
dy = p3*y - p4*x*y;
dz = p5*z + p6*y*z;
eq1 = dx == 0;
eq2 = dy == 0;
eq3 = dz == 0;
edos = solve([eq1, eq2, eq3], [x, y, z], 'Real', true);
n = length(edos.x);
fprintf('The system has %d equilibrium point(s).\n\n', n);
```

The system has 2 equilibrium point(s).

```
for i = 1:min(2,n)
fprintf('Equilibrium point %d:\n', i);
fprintf('x = %s\n', char(edos.x(i)));
fprintf('y = %s\n', char(edos.y(i)));
fprintf('z = %s\n\n', char(edos.z(i)));
end
```

```
Equilibrium point 1:
x = 0
y = 0
z = 0
Equilibrium point 2:
x = p3/p4
y = -p5/p6
z = -p1/p2
```

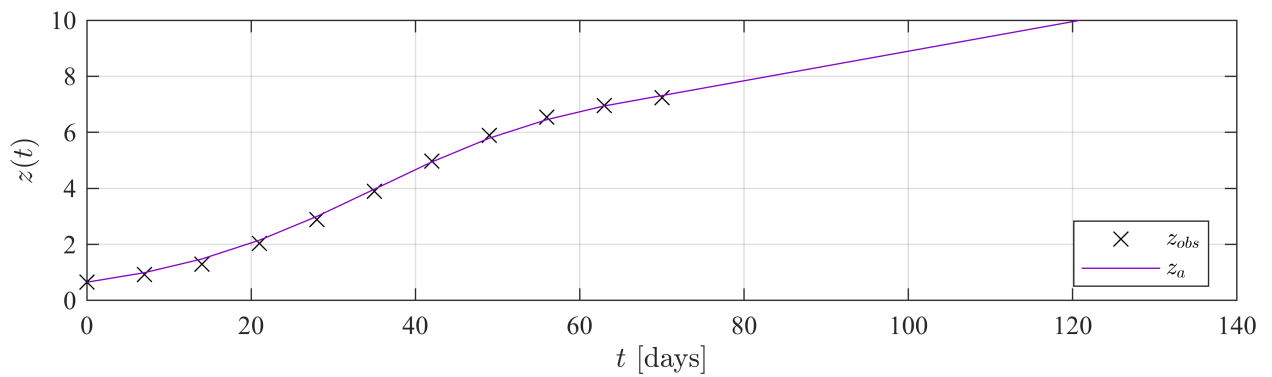
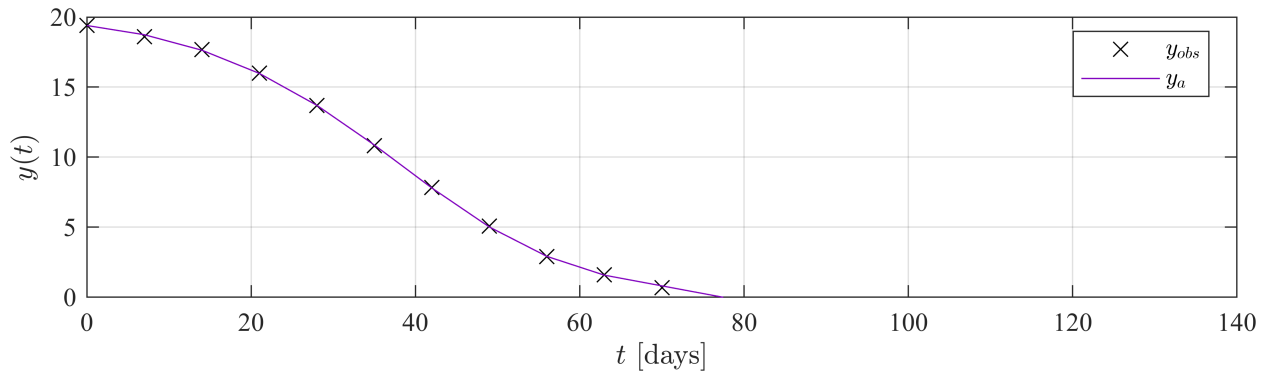
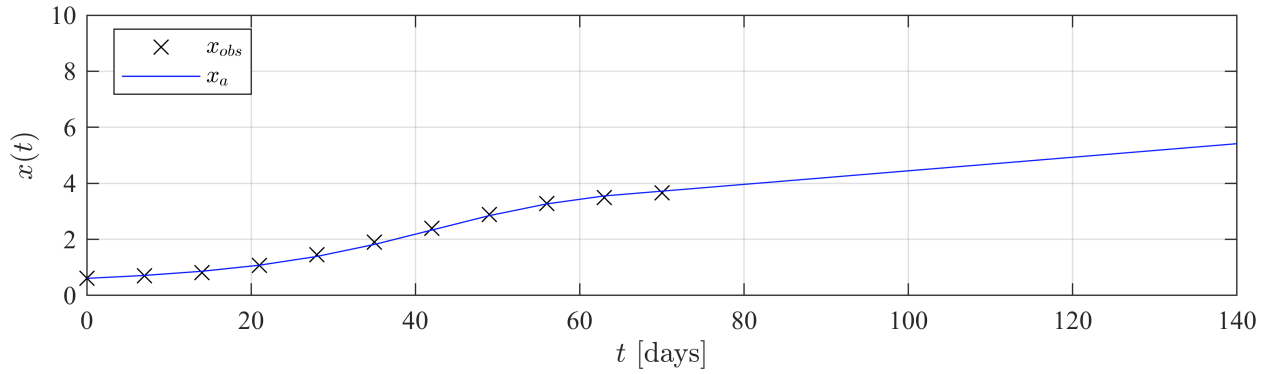
## 2tMax

```
warning('off')
sys = readmatrix('data_fit.csv');
t = sys(:,1);
x = sys(:,2);
y = sys(:,3);
z = sys(:,4);
P0 = [0.0012; 0.00055; 0.013; 0.028; 0.0020; 0.0033];
[mdl, xa, ya, za, t_ext, x_ext, y_ext, z_ext] = VariedExtended(t, x, y, z, P0);
```

```
[EXTENDED] Sample size (n): 11
Parameters to be estimated: 6
Degrees of freedom: 27
R-squared: 0.9998
Corrected AIC: -65.2787
```

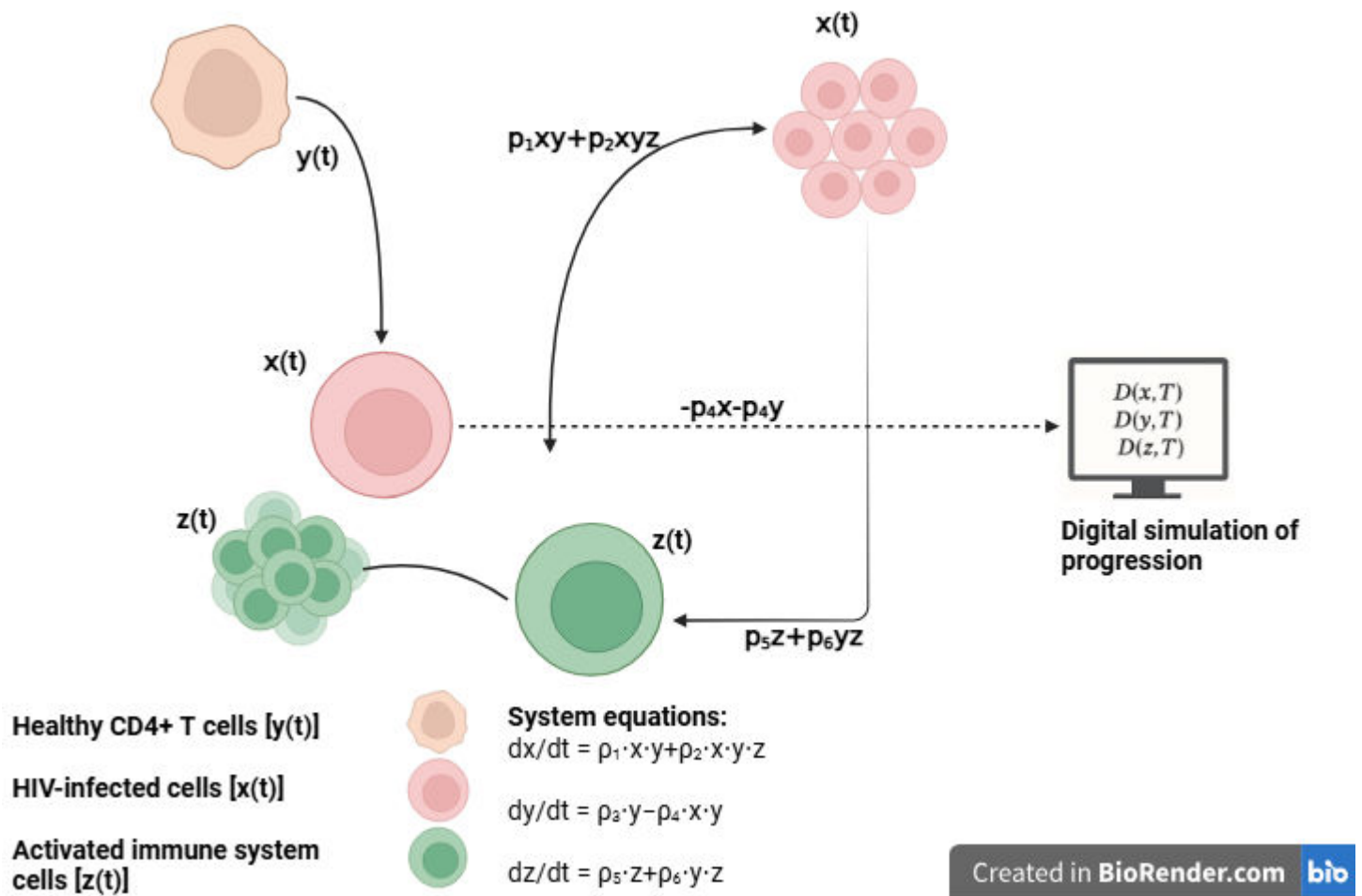
Param	Estimate	SE	MoE	CI95		pValue
"p1"	0.0006015	9.2161e-05	0.0001891	0.00041241	0.0007906	5.3482e-07
"p2"	0.00072444	3.8939e-05	7.9896e-05	0.00064454	0.00080434	6.3191e-17
"p3"	0.014804	0.00042563	0.00087331	0.013931	0.015677	5.8103e-24
"p4"	0.030033	0.00055695	0.0011428	0.02889	0.031176	4.956e-29
"p5"	0.0040202	0.00067335	0.0013816	0.0026386	0.0054018	2.2873e-06
"p6"	0.0029262	5.5581e-05	0.00011404	0.0028122	0.0030402	9.4077e-29

```
plot2t(t, x, y, z, t_ext, x_ext, y_ext, z_ext);
```



## Representation





The diagram illustrates a dynamic model of HIV infection, developed from in vitro experimental data. The system is composed of three interacting immune cell populations:

- Normal Cells  $y(t)$ : CD4+ T cells not yet infected.
- Infected Cells  $x(t)$ : CD4+ T cells that have been infected.
- Immune System Cells  $z(t)$ : Activated effector cells (e.g., CD8+ cytotoxic T lymphocytes).

Arrows between the cell types represent biological interactions expressed through nonlinear mathematical terms formulated as differential equations. These expressions describe the rate of change of each population over time, using symbolic parameters  $p_1$  through  $p_6$ , which represent different biological interaction rates.

For example:

- The infection process is modeled by the interaction term  $p_1xy$ , and it is further influenced by the immune system through  $p_2xyz$ .
- The immune response is described by the activation of immune cells via intrinsic growth ( $p_5z$ ) and stimulation by healthy cells ( $p_6yz$ ).
- The loss of healthy cells due to infection is modeled by the term  $-p_4xy$ , while their natural proliferation is given by  $p_3y$ .

On the right side, the diagram includes a digital twin simulation block, representing the system of differential equations:

$$\begin{aligned}\frac{dx}{dt} &= \rho_1xy + \rho_2xyz \\ \frac{dy}{dt} &= \rho_3y - \rho_4xy \\ \frac{dz}{dt} &= \rho_5z + \rho_6yz\end{aligned}$$

This simulation-based model enables the prediction of system dynamics over time and supports the exploration of long-term behavior and stability under controlled computational conditions, without requiring specific clinical assumptions.

## Conclusions

In this project, a nonlinear dynamic model was built, simulated, and analyzed with the aim of representing the behavior of a physiological system made up of three interrelated variables over time. The modeling process was driven by the data collected, relying exclusively on experimental time series collected over a specific period. The differential equations developed describe the temporal evolution of each variable, incorporating six parameters that represent the interactions between them. These parameters were estimated through nonlinear regression, using a smoothed version of the original dataset to minimize measurement noise and discontinuities.

One of the most notable outcomes of the model fitting was the coefficient of determination ( $R^2 = 0.9998$ ), which shows that the system of equations closely matches the observed data. The corrected Akaike Information Criterion ( $AICc = -65.28$ ) also supports the adequacy of the model's structure, achieving a balance between the quality of fit and model complexity. All six estimated parameters were statistically significant, with p-values close to zero and narrow confidence intervals, indicating that each one contributes meaningfully to the system's dynamics. Moreover, the model does not include redundant or unnecessary terms, which is essential for both interpretation and potential generalization.

The Jacobian matrix analysis provided useful insights into the system's local sensitivity and the interdependence among variables. Two equilibrium points were identified, including a non-trivial fixed point with non-zero values, suggesting that the system tends toward stability under certain conditions. These equilibrium states help characterize long-term behavior and could serve as a basis for future stability studies or control strategies.

Also was implemented a different approach for the work (2t model), in which the system's behavior was projected over a time span twice as long as the original dataset. The results of this extension showed that the model remains stable and consistent outside the initial data range. No abrupt divergences or unrealistic trends were observed, supporting the idea that the model is not only descriptive but also predictive within a reasonable timeframe. Lastly, visual comparisons between the observed data, the smoothed data, and the simulated outputs confirmed that the model effectively follows the expected trajectories, capturing key features such as exponential growth, decay, saturation, or delayed responses between variables.

## Function

## Raw data plot

```
function plotdata(t, x, y, z)
    c1 = [0.60, 0.00, 0.00];
    c2 = [0.00, 0.75, 1.00];
    c3 = [0.50, 0.00, 0.75]; +
    fig = figure('Color', 'w', 'Units', 'Centimeters', 'Position', [2, 2, 22, 18]);

    subplot(3,1,1);
    plot(t, x, 'x', 'LineWidth', 1.5, 'Color', c1, 'DisplayName', '$x_1(t)$');
    set(gca, 'FontName', 'Times New Roman', 'FontSize', 10);
    xlabel('$t$ $[days]$', 'Interpreter', 'latex');
    ylabel('$x(t)$', 'Interpreter', 'latex');
    xlim([0 70]); xticks(0:10:70);
    ylim([0 4]); yticks(0:0.5:4);
    grid on;

    subplot(3,1,2);
    plot(t, y, 'x', 'LineWidth', 1.5, 'Color', c2, 'DisplayName', '$y_1(t)$');
    set(gca, 'FontName', 'Times New Roman', 'FontSize', 10);
    xlabel('$t$ $[days]$', 'Interpreter', 'latex');
    ylabel('$y(t)$', 'Interpreter', 'latex');
    xlim([0 70]); xticks(0:10:70);
    ylim([0 22]); yticks(0:2:22);
    grid on;

    subplot(3,1,3);
    plot(t, z, 'x', 'LineWidth', 1.5, 'Color', c3, 'DisplayName', '$z_1(t)$');
    set(gca, 'FontName', 'Times New Roman', 'FontSize', 10);
    xlabel('$t$ $[days]$', 'Interpreter', 'latex');
    ylabel('$z(t)$', 'Interpreter', 'latex');
    xlim([0 70]); xticks(0:10:70);
    ylim([0 8]); yticks(0:1:8);
    grid on;
    print(fig, 'RawData', '-dpdf', '-bestfit');
end
```

## Smooth data plot

```
function plotDataS(t, x, y, z)
    c1 = [0.60, 0.00, 0.00];
    c2 = [0.00, 0.75, 1.00];
    c3 = [0.50, 0.00, 0.75];
    x_s = smoothdata(x, 'movmean', 5);
    y_s = smoothdata(y, 'movmean', 5);
    z_s = smoothdata(z, 'movmean', 5);
    fig = figure('Color', 'w', 'Units', 'Centimeters', 'Position', [2, 5, 20, 18]);
```

```

tmax = 70;

subplot(3,1,1);
plot(t, x_s, 'x', 'LineWidth', 1.5, 'Color', c1, 'DisplayName', '$x_1(t)$');
set(gca, 'FontName', 'Times New Roman', 'FontSize', 10);
xlabel('$t$ $[days]$', 'Interpreter', 'latex');
ylabel('$x(t)$', 'Interpreter', 'latex');
xlim([0 tmax]); xticks(0:10:tmax);
ylim([0 4]); yticks(0:0.5:4);
grid on;

subplot(3,1,2);
plot(t, y_s, 'x', 'LineWidth', 1.5, 'Color', c2, 'DisplayName', '$y_1(t)$');
set(gca, 'FontName', 'Times New Roman', 'FontSize', 10);
xlabel('$t$ $[days]$', 'Interpreter', 'latex');
ylabel('$y(t)$', 'Interpreter', 'latex');
xlim([0 tmax]); xticks(0:10:tmax);
ylim([0 22]); yticks(0:2:22);
grid on;

subplot(3,1,3);
plot(t, z_s, 'x', 'LineWidth', 1.5, 'Color', c3, 'DisplayName', '$z_1(t)$');
set(gca, 'FontName', 'Times New Roman', 'FontSize', 10);
xlabel('$t$ $[days]$', 'Interpreter', 'latex');
ylabel('$z(t)$', 'Interpreter', 'latex');
xlim([0 tmax]); xticks(0:10:tmax);
ylim([0 8]); yticks(0:1:8);
grid on;
datos_fit = table(t(:), x_s(:), y_s(:), z_s(:), ...
    'VariableNames', {'T', 'x', 'y', 'z'});
writetable(datos_fit, 'data_fit.csv');
disp('Data smoothed exported as: "data_fit.csv".');
print(fig, 'DataSmooth', '-dpdf', '-bestfit');
end

```

## Nonlinear regression algorithm

```

function [mdl, xa, ya, za] = Varied(t, x, y, z, P0)
    x0 = x(1); y0 = y(1); z0 = z(1);

    t3 = [t; t; t];
    fo = [x; y; z];

    function fi = model(p, tvec)
        dt = 0.01;
        tmat = reshape(tvec, [], 3);
        t_local = tmat(:,1);
        time = (0:dt:max(t_local))';
    end

```

```

n = numel(time);

x_sim = zeros(n,1); x_sim(1) = x0;
y_sim = zeros(n,1); y_sim(1) = y0;
z_sim = zeros(n,1); z_sim(1) = z0;

for i = 1:n-1
    [dx, dy, dz] = f(x_sim(i), y_sim(i), z_sim(i), p);
    xp = x_sim(i) + dx*dt;
    yp = y_sim(i) + dy*dt;
    zp = z_sim(i) + dz*dt;
    [dxp, dyp, dzp] = f(xp, yp, zp, p);

    x_sim(i+1) = x_sim(i) + (dx + dxp)*dt/2;
    y_sim(i+1) = y_sim(i) + (dy + dyp)*dt/2;
    z_sim(i+1) = z_sim(i) + (dz + dzp)*dt/2;
end

xi = interp1(time, x_sim, t_local, 'linear', NaN);
yi = interp1(time, y_sim, t_local, 'linear', NaN);
zi = interp1(time, z_sim, t_local, 'linear', NaN);

fi = [xi; yi; zi];
end

function [dx, dy, dz] = f(x, y, z, p)
    dx = p(1)*x*y + p(2)*x*y*z;
    dy = p(3)*y - p(4)*x*y;
    dz = p(5)*z + p(6)*y*z;
end

mdl = fitnlm(t3, fo, @model, P0);
fa = mdl.Fitted;
fn = reshape(fa, [], 3);
xa = fn(:,1);
ya = fn(:,2);
za = fn(:,3);
coef = mdl.Coefficients;
Estimate = coef.Estimate;
SE = coef.SE;
pValue = coef.pValue;
CI95 = coefCI(mdl, 0.05);
dof = mdl.DFE;
tval = tinv(0.975, dof);
MoE = SE * tval;

Params = ["p1"; "p2"; "p3"; "p4"; "p5"; "p6"];
Results = table(Params, Estimate, SE, MoE, CI95, pValue, ...
    'VariableNames', {'Param', 'Estimate', 'SE', 'MoE', 'CI95', 'pValue'});

fprintf('\nSample size (n): %d', numel(x));

```

```

fprintf('\nParameters to be estimated: %d', numel(P0));
fprintf('\nDegrees of freedom: %d', dof);
fprintf('\nR-squared: %.4f', mdl.Rsquared.Ordinary);
fprintf('\nCorrected AIC: %.4f\n\n', mdl.ModelCriterion.AICc);
disp(Results);

X_plot = [x, xa];
Y_plot = [y, ya];
Z_plot = [z, za];
plotXYZResults(t, X_plot, Y_plot, Z_plot);

```

end

## Nonlinear regression plot

```

function plotXYZResults(t, x, y, z)
    negro = [0.00, 0.00, 0.00];
    azul = [0.04, 0.10, 1.00];
    morado = [0.50, 0.00, 0.75];
    tmax = 70;
    fig = figure('Color', 'w', 'Units', 'Centimeters', 'Position', [2, 5, 20, 18]);

    subplot(3,1,1);
    plot(t, x(:,1), 'x', 'Color', negro, 'MarkerSize', 10, 'DisplayName', '$x_{0}$');
    hold on;
    plot(t, x(:,2), '-', 'Color', azul, 'LineWidth', 1.2, 'DisplayName', '$x_{a}$');
    set(gca, 'FontName', 'Times New Roman', 'FontSize', 10);
    xlabel('$t$ [days]', 'Interpreter', 'latex');
    ylabel('$x(t)$', 'Interpreter', 'latex');
    legend('Interpreter', 'latex', 'Location', 'best');
    xlim([0 tmax]); ylim([0 5]);
    xticks(0:10:tmax); yticks(0:1:5);
    grid on;

    subplot(3,1,2);
    plot(t, y(:,1), 'x', 'Color', negro, 'MarkerSize', 10, 'DisplayName', '$y_{0}$');
    hold on;
    plot(t, y(:,2), '-', 'Color', morado, 'LineWidth', 1.2, 'DisplayName', '$y_{a}$');
    set(gca, 'FontName', 'Times New Roman', 'FontSize', 10);
    xlabel('$t$ [days]', 'Interpreter', 'latex');
    ylabel('$y(t)$', 'Interpreter', 'latex');
    legend('Interpreter', 'latex', 'Location', 'best');
    xlim([0 tmax]); ylim([0 25]);
    xticks(0:10:tmax); yticks(0:5:25);
    grid on;

```

```

subplot(3,1,3);
plot(t, z(:,1), 'x', 'Color', negro, 'MarkerSize', 10, 'DisplayName', '$z_{o}$');
hold on;
plot(t, z(:,2), '-', 'Color', morado, 'LineWidth', 1.2, 'DisplayName', '$z_{a}$');
set(gca, 'FontName', 'Times New Roman', 'FontSize', 10);
xlabel('$t$ $[days]$', 'Interpreter', 'latex');
ylabel('$z(t)$', 'Interpreter', 'latex');
legend('Interpreter', 'latex', 'Location', 'best');
xlim([0 tmax]); ylim([0 10]);
xticks(0:10:tmax); yticks(0:2:10);
grid on;
print(fig, 'NonlinearRegression', '-dpdf', '-bestfit');
end

```

## 2T model

```

function [mdl, xa, ya, za, t_ext, x_ext, y_ext, z_ext] = VariedExtended(t, x, y, z, P0)
    x0 = x(1); y0 = y(1); z0 = z(1);
    t3 = [t; t; t];
    fo = [x; y; z];
    function fi = model(p, tvec)
        dt = 0.01;
        tmat = reshape(tvec, [], 3);
        t_local = tmat(:,1);

        TMAX = 2 * max(t_local);
        time = (0:dt:TMAX)';
        n = numel(time);

        x_sim = zeros(n,1); x_sim(1) = x0;
        y_sim = zeros(n,1); y_sim(1) = y0;
        z_sim = zeros(n,1); z_sim(1) = z0;

        for i = 1:n-1
            [dx, dy, dz] = f(x_sim(i), y_sim(i), z_sim(i), p);
            xp = x_sim(i) + dx*dt;
            yp = y_sim(i) + dy*dt;
            zp = z_sim(i) + dz*dt;
            [dxp, dyp, dzp] = f(xp, yp, zp, p);

            x_sim(i+1) = x_sim(i) + (dx + dxp)*dt/2;
            y_sim(i+1) = y_sim(i) + (dy + dyp)*dt/2;
            z_sim(i+1) = z_sim(i) + (dz + dzp)*dt/2;
        end
    end
end

```

end

```
t_ext = time;  
x_ext = x_sim;  
y_ext = y_sim;  
z_ext = z_sim;
```

```
xi = interp1(time, x_sim, t_local, 'linear', NaN);  
yi = interp1(time, y_sim, t_local, 'linear', NaN);  
zi = interp1(time, z_sim, t_local, 'linear', NaN);
```

```
fi = [xi; yi; zi];
```

end

```
function [dx, dy, dz] = f(x, y, z, p)
```

```
dx = p(1)*x*y + p(2)*x*y*z;  
dy = p(3)*y - p(4)*x*y;  
dz = p(5)*z + p(6)*y*z;
```

end

```
mdl = fitnlm(t3, fo, @model, P0);
```

```
fa = mdl.Fitted;
```

```
fn = reshape(fa, [], 3);
```

```
xa = fn(:,1);
```

```
ya = fn(:,2);
```

```
za = fn(:,3);
```

```
coef = mdl.Coefficients;
```

```
Estimate = coef.Estimate;
```

```
SE = coef.SE;
```

```
pValue = coef.pValue;
```

```
CI95 = coefCI(mdl, 0.05);
```

```
dof = mdl.DFE;
```

```
tval = tinv(0.975, dof);
```

```
MoE = SE * tval;
```

```
Params = ["p1"; "p2"; "p3"; "p4"; "p5"; "p6"];
```

```
Results = table(Params, Estimate, SE, MoE, CI95, pValue, ...  
    'VariableNames', {'Param','Estimate','SE','MoE','CI95','pValue'});
```

```
fprintf('\n[EXTENDED] Sample size (n): %d', numel(x));
```

```
fprintf('\nParameters to be estimated: %d', numel(P0));
```

```
fprintf('\nDegrees of freedom: %d', dof);
```

```
fprintf('\nR-squared: %.4f', mdl.Rsquared.Ordinary);
```

```
fprintf('\nCorrected AIC: %.4f\n\n', mdl.ModelCriterion.AICc);
```

```
disp(Results);
```

```
TMAX = 2 * max(t);
```

```
t_ext = (0:0.01:TMAX)';
```

```
x_ext = interp1(t, xa, t_ext, 'linear', 'extrap');
```

```
y_ext = interp1(t, ya, t_ext, 'linear', 'extrap');
```

```
z_ext = interp1(t, za, t_ext, 'linear', 'extrap');
```

end



## 2T plot

```
function plot2t(t, x, y, z, t_ext, x_ext, y_ext, z_ext)
    negro = [0.00, 0.00, 0.00];
    azul = [0.04, 0.10, 1.00];
    morado = [0.50, 0.00, 0.75];
    fig = figure('Color', 'w', 'Units', 'centimeters', 'Position', [2, 2, 20, 18]);

    subplot(3,1,1);
    hold on; box on;
    plot(t, x, 'x', 'MarkerSize', 8, 'Color', negro, 'DisplayName', '$x_{obs}$');
    plot(t_ext, x_ext, '-', 'MarkerSize', 1, 'Color', azul, 'DisplayName', '$x_{a}$');
    xlabel('$t$ [days]', 'Interpreter', 'latex');
    ylabel('$x(t)$', 'Interpreter', 'latex');
    xlim([0 max(t_ext)]); ylim([0 10]);
    legend('Interpreter', 'latex', 'Location', 'best');
    set(gca, 'FontName', 'Times New Roman', 'FontSize', 10);
    grid on;

    subplot(3,1,2);
    hold on; box on;
    plot(t, y, 'x', 'MarkerSize', 8, 'Color', negro, 'DisplayName', '$y_{obs}$');
    plot(t_ext, y_ext, '-', 'MarkerSize', 1, 'Color', morado, 'DisplayName', '$y_{a}$');
    xlabel('$t$ [days]', 'Interpreter', 'latex');
    ylabel('$y(t)$', 'Interpreter', 'latex');
    xlim([0 max(t_ext)]); ylim([0 20]);
    legend('Interpreter', 'latex', 'Location', 'best');
    set(gca, 'FontName', 'Times New Roman', 'FontSize', 10);
    grid on;

    subplot(3,1,3);
    hold on; box on;
    plot(t, z, 'x', 'MarkerSize', 8, 'Color', negro, 'DisplayName', '$z_{obs}$');
    plot(t_ext, z_ext, '-', 'MarkerSize', 1, 'Color', morado, 'DisplayName', '$z_{a}$');
    xlabel('$t$ [days]', 'Interpreter', 'latex');
    ylabel('$z(t)$', 'Interpreter', 'latex');
    xlim([0 max(t_ext)]); ylim([0 10]);
    legend('Interpreter', 'latex', 'Location', 'best');
    set(gca, 'FontName', 'Times New Roman', 'FontSize', 10);
    grid on;
    print(fig, 'TrayectoriaExtendida', '-dpdf', '-bestfit');
end
```