Article

Real-Time Adaptive TECS Gain Tuning Using Neural Networks for Tiltrotor eVTOL

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Abstract: Tiltrotor electric Vertical Take-Off and Landing (eVTOL) aircraft encounter significant control challenges during the transition from hover to forward flight, particularly when using open-source autopilot systems like PX4. During this phase, the autopilot employs open-loop tilt angle control and linear weighting of rotor and fixed-wing control outputs, neglecting the aircraft's nonlinear dynamics and resulting in altitude loss. After the transition, when Total Energy Control System (TECS) activates in fixed-wing mode, its fixed gains fail to rapidly correct accumulated errors, exacerbating altitude subsidence. This paper proposes a novel approach to enhance TECS performance after forward transition by dynamically adjusting its gains in real-time using a simple neural network. By adapting gains based on the aircraft's state, this method minimizes altitude loss and stabilizes airspeed and pitch attitude more effectively. Simulation results demonstrate that the neural network-based adaptive TECS significantly reduces altitude subsidence and improves flight stability compared to static gain configurations. This research provides a practical solution to enhance the control performance of tiltrotor eVTOLs, addressing limitations in open-source autopilots and supporting their application in urban air mobility.

Keywords: Tiltrotor eVTOL; TECS; neural networks; adaptive gain tuning; (List three to ten pertinent keywords specific to the article; yet reasonably common within the subject discipline.)

1. Introduction

The development of electric Vertical Take-Off and Landing (eVTOL) aircraft, particularly those with tiltrotor configurations, has gained significant attention due to their potential in urban air mobility and logistics. These hybrid aircraft integrate the vertical lift capabilities of helicopters with the forward flight efficiency of fixed-wing airplanes, necessitating advanced control systems to manage the transition from hover to cruise modes [1].

Open-source autopilot systems, such as PX4, are widely adopted for their accessibility and flexibility. TECS, introduced by Lambregts in 1983 [?], manages total energy—combining kinetic and potential energy—to regulate airspeed and altitude via coordinated throttle and elevator inputs. While effective for fixed-wing aircraft, its application to tiltrotor eVTOLs requires adaptation to address the unique dynamics of the transition phase. Previous studies have validated adaptive TECS strategies for fixed-wing aircraft [?] and proposed modifications for unmanned aerial vehicles [?], yet these approaches have not been explored during the transition phase of tiltrotor eVTOLs.

In PX4, the transition phase relies on open-loop tilt angle control and linear weighting of rotor and fixed-wing control outputs, which neglects nonlinear dynamics and leads to altitude loss. TECS activates only after the transition, in fixed-wing mode, where its fixed gains fail to promptly correct errors accumulated during the transition. This study proposes a real-time adaptive TECS gain tuning framework using a simple neural network to address these limitations after forward transition. The neural network dynamically adjusts TECS gains based on the aircraft's current state (e.g., altitude error, airspeed, pitch rate), aiming to minimize altitude loss and ensure rapid stabilization of airspeed and pitch attitude. This approach provides a practical alternative to designing a complex transition control system.

The paper is structured as follows: Section 2 analyzes the transition phase of open-source autopilots and the limitations of TECS implementation. Section 3 details the neural network-based gain tuning methodology. Section 4 presents simulation results comparing the adaptive TECS with conventional methods. Section 5 concludes with implications and future research directions. Through this work, we aim to enhance the reliability and efficiency of control systems for next-generation eVTOL platforms.
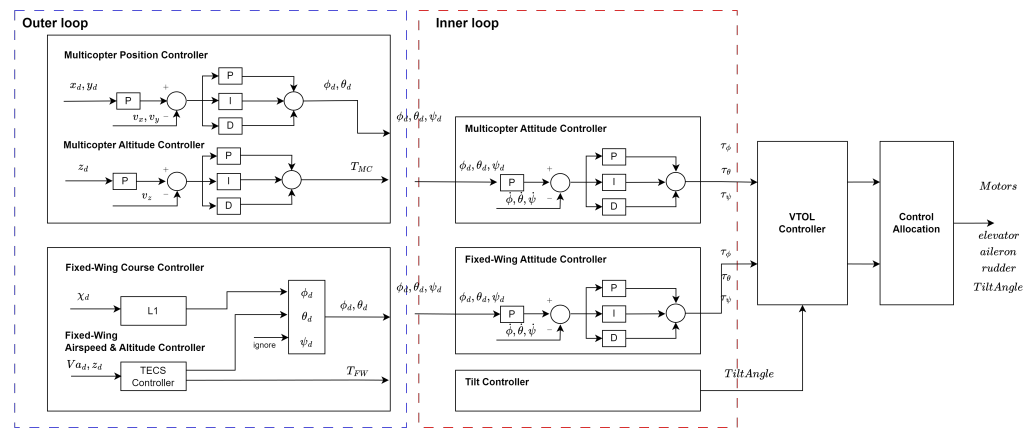
Figure 1. PX4 VTOL Control Structure

2. Literature Review

2.1. Analysis of Open-Source Autopilot Transition Phase

Open-source autopilot systems, such as PX4, support a variety of unmanned aerial vehicles (UAVs), including tiltrotor eVTOLs. However, their transition logic from hover to forward flight relies on simplified strategies that inadequately address the complex dynamics of hybrid aircraft.

2.1.1. Transition Phase Operation

The transition phase in open-source autopilots involves two key mechanisms:

- Open Loop Tilt Angle Control
  The front rotor of the tilt rotor tilts to a certain angle to achieve airspeed during the transition. Once the transition starting speed (BLENDED_ASPD) is reached, the transition algorithm begins. Until transition completion airspeed, tilt angle keep particular tilt angle(VT_TILT_TRANS). When the transition completion speed (TRANSITION_ASPD) is reached, the switch to fixed-wing mode starts.
- Linear Weighting of Control Outputs
  At this point, the outputs for attitude control of the fixed-wing and rotary-wing are divided based on the current airspeed with weighting applied. Once the transition completion speed (TRANSITION_ASPD) is reached, the output of the fixed-wing control takes over completely.

```
Start in Multicopter Mode.
Tilt the rotors forward to 15°.
If airspeed < BLENDED_ASPD m/s:
    Maintain Multicopter Mode.
Begin Transition Mode P1:
    Set weight = 1.0 - (airspeed - BLENDED_ASPD) / (TRANSITION_ASPD - BLENDED_
    Distribute attitude control based on the weight.
    Set thrust to the TRANSITION_THRUST
    Tilt the rotors forward to 45°.
If airspeed < TRANSITION_ASPD m/s:
    Maintain Transition Mode P1.
Begin Transition Mode P2:
    Tilt the rotors forward to 90°.
Switch to Fixed-Wing Mode.
```

In the transition phase, the autopilot does not consider the aircraft's nonlinear dynamics, such as rotor tilt angle changes, which can lead to altitude loss and instability post-transition.

2.2. Total Energy Control System (TECS)

The Total Energy Control System (TECS) is a control strategy used in aircraft to manage their total energy, which consists of potential energy ($E_P = mgh$) and kinetic energy ($E_K = \frac{1}{2}mv^2$). TECS regulates airspeed and altitude by coordinating throttle and pitch inputs: throttle adjusts the total energy, while pitch controls the distribution of energy between altitude and speed.
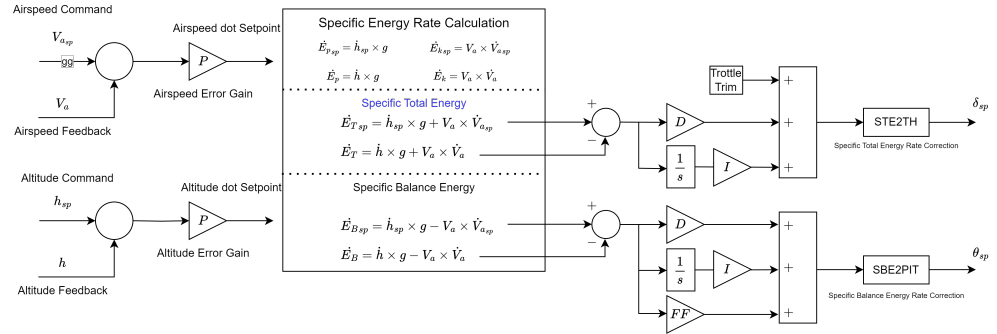
Figure 2. PX4 TECS Control Structure

In the Open-Source autopilot framework, TECS remains inactive during the transition phase of tiltrotor eVTOLs and only activates once the aircraft fully enters fixed-wing mode.

2.2.1. TECS Control Structure

TECS manages two primary functions:

- Total Energy Control: Adjusts throttle to regulate the total specific energy rate ($\dot{E}_T = \dot{E}_P + \dot{E}_K$), where:
  - $\dot{E}_P = \dot{h}g$ (potential energy rate, linked to altitude change),
  - $\dot{E}_K = v\dot{v}$ (kinetic energy rate, linked to speed change).
- Balance Energy Control: Adjusts pitch to control the balance specific energy rate ($\dot{E}_B = \dot{E}_P - \dot{E}_K$), ensuring proper energy distribution between altitude and airspeed.

The setpoints for these energy rates are defined as:

$$\dot{E}_{T_{\text{sp}}} = \dot{h}_{\text{sp}}g + v_a\dot{v}_{a_{\text{sp}}}, \quad \dot{E}_{B_{\text{sp}}} = \dot{h}_{\text{sp}}g - v_a\dot{v}_{a_{\text{sp}}} \tag{1}$$

where $\dot{h}_{\text{sp}}$ is the desired altitude rate, v_a is the current airspeed, and $\dot{v}_{a_{\text{sp}}}$ is the desired airspeed rate.

The control outputs—thrust (δ_{sp}) and pitch (θ_{sp})—are calculated as:

$$\delta_{\text{sp}} = \left(D_T(\dot{E}_{T_{\text{sp}}} - \dot{E}_T) + I_T \int (\dot{E}_{T_{\text{sp}}} - \dot{E}_T)\, dt + T_{\text{cruise}} \right) \frac{1}{\dot{E}_{T,\text{max}} - \dot{E}_{T,\text{min}}} \tag{2}$$

$$\theta_{\text{sp}} = \left(D_B(\dot{E}_{B_{\text{sp}}} - \dot{E}_B) + I_B \int (\dot{E}_{B_{\text{sp}}} - \dot{E}_B)\, dt + ff_B\dot{E}_{B_{\text{sp}}} \right) \frac{1}{v_a g} \tag{3}$$

where:

- D_T, I_T: Derivative and integral gains for specific total energy rate,
- D_B, I_B: Derivative and integral gains for specific balance energy rate,
- ff_B: Feedforward gain,
- T_{cruise}: Cruise thrust,
- $\dot{E}_{T,\text{max}} = g \times \text{max\_climb\_rate}$, $\dot{E}_{T,\text{min}} = g \times \text{max\_descent\_rate}$: Limits on energy rates.

For tiltrotors, the transition phase introduces unique dynamics, such as rotor tilt angle changes, which affect the balance between $\dot{E}_P$ and $\dot{E}_K$. Fixed gains, as used in PX4's default TECS, fail to adapt to these rapid changes, resulting in delayed altitude recovery and instability post-transition.

3. Methodology

3.1. Neural Network-Based Gain Tuning Methodology

To overcome the limitations of static TECS gains, this study proposes a neural network-based approach for real-time gain tuning, specifically tailored to the post-transition phase of tiltrotor eVTOLs. The methodology dynamically adjusts the proportional (K_p) and integral (K_i) gains of TECS based on the aircraft's flight state, mitigating altitude loss and improving stability immediately after the transition to fixed-wing mode.

3.1.1. Why Use a Neural Network?

Immediately after the transition to fixed-wing mode, the aircraft experiences residual effects from the transition phase, such as aerodynamic shifts and thrust vector changes. Fixed gains cannot adequately respond to these variations, leading to poor altitude control. A neural network enables real-time adaptation by learning and adjusting gains based on current errors (e.g., altitude and energy rate deviations), offering a robust solution for open-source platforms like PX4.

3.1.2. Neural Network Design

The proposed neural network is a simple two-layer feedforward network:

- **Inputs**:
    - Proportional error: $e_p(k) = \dot{E}_{\text{sp}}(k) - \dot{E}(k)$,
    - Integral error: $e_i(k) = \sum \int e_p(k)\, dt$.
- **Outputs**:

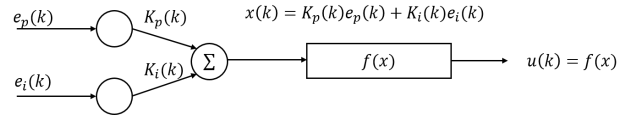– Adjusted gains $K_p(k)$ and $K_i(k)$, computed using a sigmoid activation function:

$$f(x) = \frac{2(1 - e^{-x \cdot Y_g})}{Y_g(1 + e^{-x \cdot Y_g})} \tag{4}$$

where $x(k) = K_p(k)e_p(k) + K_i(k)e_i(k)$, and Y_g is a tuning parameter shaping the sigmoid curve in figure 3.

The control input $u(k) = f(x)$ drives the TECS outputs (thrust and pitch), with gains updated dynamically.
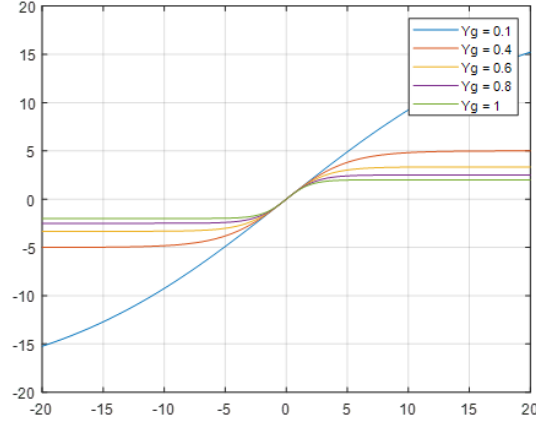
Figure 3. Sigmoid function shapes for different Y_g values

Figure 4. Neural network block diagram for adaptive TECS gain tuning

3.2. Tuning Process

The gains are tuned using the steepest descent method to minimize the cost function:

$$J(k) = \frac{1}{2}(\dot{E}_{sp}(k) - \dot{E}(k))^2 \tag{5}$$

The update rules are:

$$K_p(k+1) = K_p(k) - \eta_p \frac{\partial J(k)}{\partial K_p(k)}, \quad K_i(k+1) = K_i(k) - \eta_i \frac{\partial J(k)}{\partial K_i(k)} \tag{6}$$

with partial derivatives:

$$\frac{\partial J(k)}{\partial K_p(k)} = -e_p(k) \cdot f'(x(k)) \cdot e_p(k), \quad \frac{\partial J(k)}{\partial K_i(k)} = -e_p(k) \cdot f'(x(k)) \cdot e_i(k) \tag{7}$$

where $f'(x) = \frac{4e^{-x \cdot Y_g}}{(1 + e^{-x \cdot Y_g})^2}$, and η_p, η_i are learning rates.

This method replaces static gains with a dynamic, state-dependent tuning mechanism, ensuring computational efficiency for real-time implementation on autopilot systems.

3.3. Implementation and Validation

The proposed method will be implemented and tested using PX4 simulations, focusing on the post-transition phase in fixed-wing mode. Performance will be evaluated by comparing altitude stability and recovery time against the default fixed-gain TECS.
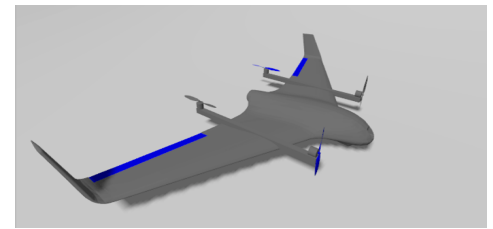
4. Model

4.1. Aircraft Model

The aircraft model used in the simulation is a tiltrotor eVTOL with the following parameters:

Symbol	Value
Physical Properties	
m	5.22 kg
J_x	1.229 kg·m²
J_y	0.1702 kg·m²
J_z	0.8808 kg·m²
J_{xz}	0.9343 kg·m²
S_{wing}	0.75 m²
b	2.10 m
$\bar{c}$	0.3571 m

Table 1. Physical Properties

Figure 5. Tiltrotor Gazebo model

Figure 6. Aircraft Model Parameters and Tiltrotor Gazebo Model

In simulation, we focus on longitudinal dynamics, including altitude, airspeed, and pitch attitude control.

5. Simulation Results

The proposed neural network-based adaptive TECS gain tuning method was implemented in PX4 simulations to evaluate its performance during the post-transition phase of a tiltrotor eVTOL. The results were compared against the default fixed-gain TECS configuration to assess improvements in altitude stability and recovery time.

5.1. Simulation Setup

The simulation environment used MATLAB Simulink. The aircraft model was integrated with the neural network-based adaptive TECS gain tuning method, replacing the default fixed gains. The simulation focused on the post-transition phase, with the aircraft transitioning to fixed-wing mode at a predetermined airspeed.

5.2. Altitude Response

Figure 7 shows the altitude response across the entire simulation period (0-100 sec), while Figure 8 zooms into the recovery phase immediately following the transition (20-30 sec). The neural network-based TECS demonstrates significantly faster altitude recovery compared to the PX4 fixed-gain TECS.
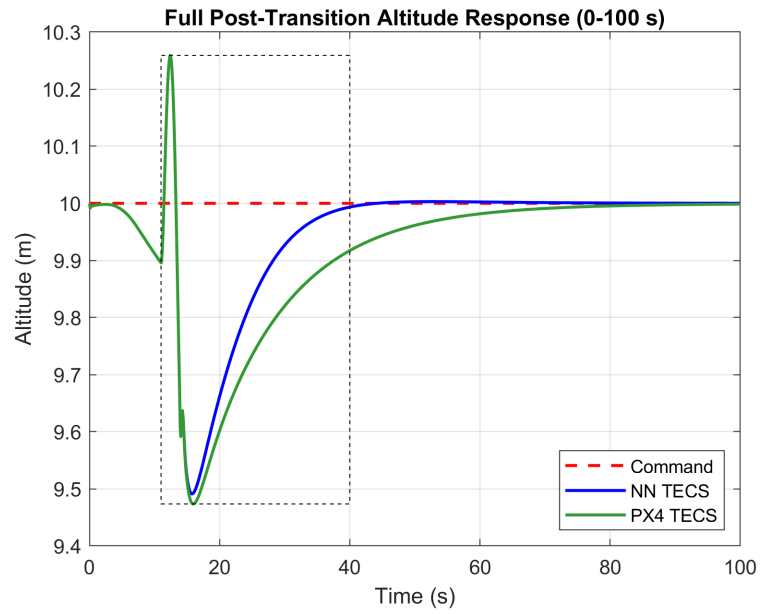
Figure 7. Altitude Response (0-100 sec): Neural Network TECS (Blue) vs. PX4 TECS (Green), Altitude Command (Red Dashed).
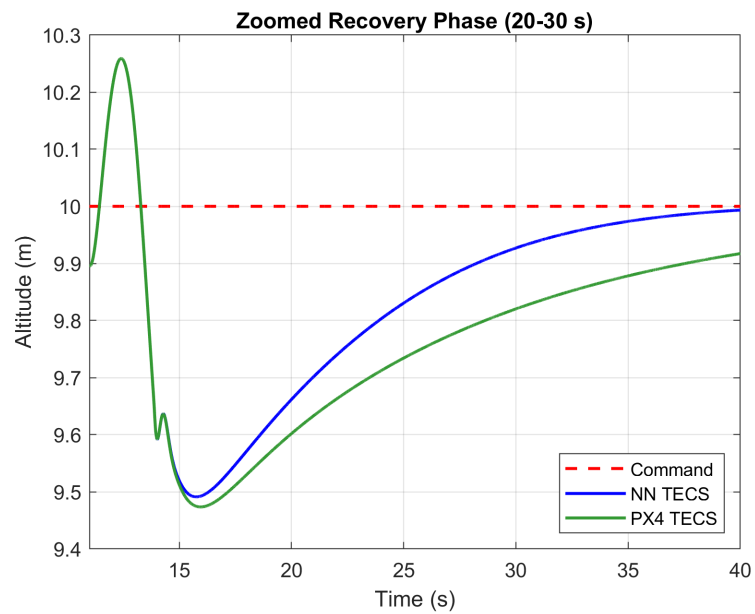
Figure 8. Zoomed Recovery Phase (20-30 sec): Neural Network TECS (Blue) vs. PX4 TECS (Green), Altitude Command (Red Dashed).
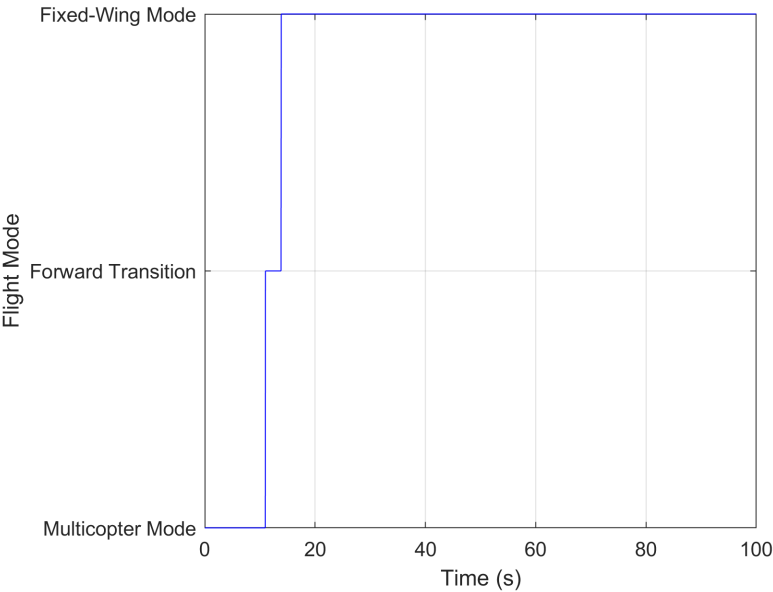
5.3. Flight Mode

In Figure 9, we illustrate the transition from multicopter mode to fixed-wing mode, where TECS is activated in fixed-wing mode.
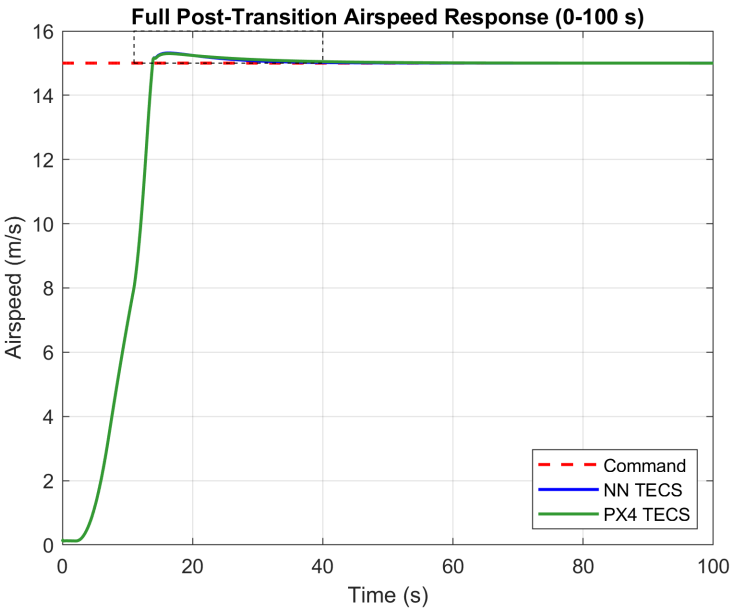
**Figure 9.** Flight Mode Transition: Multicopter to Fixed-Wing Mode.

5.4. Airspeed Response

Figure 10 presents the airspeed response during the simulation (0-100 sec), and Figure 11 zooms into the recovery phase (20-30 sec). Both methods exhibit similar stabilization performance.

Figure 10. Airspeed Response (0-100 sec): Neural Network TECS (Blue) vs. PX4 TECS (Green), Airspeed Command (Red Dashed).
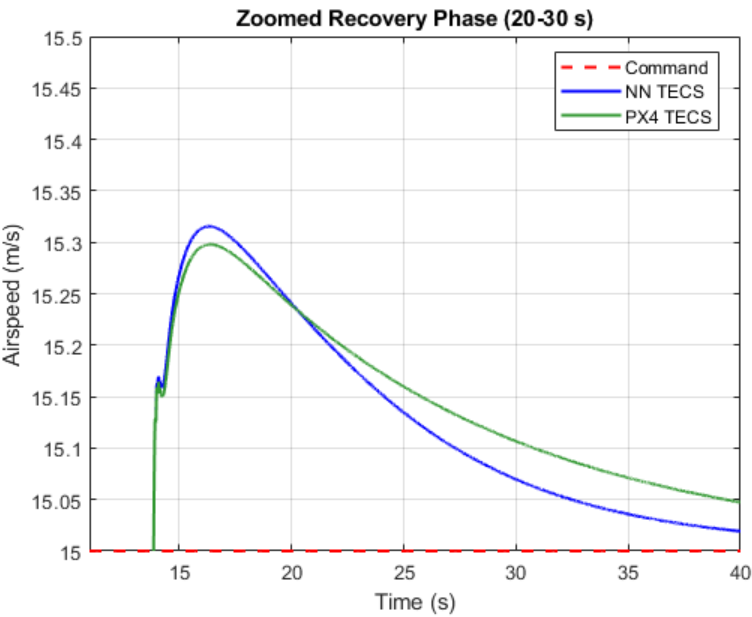
Figure 11. Zoomed Recovery Phase (20-30 sec): Neural Network TECS (Blue) vs. PX4 TECS (Green), Airspeed Command (Red Dashed).

6. Discussion

7. Conclusions

This section is not mandatory, but can be added to the manuscript if the discussion is unusually long or complex.

8. Patents

Informed Consent Statement: Any research article describing a study involving humans should contain this statement. Please add "Informed consent was obtained from all subjects involved in the study." OR "Patient consent was waived due to REASON (please provide a detailed justification)." OR "Not applicable" for studies not involving humans. You might also choose to exclude this statement if the study did not involve humans.

Written informed consent for publication must be obtained from participating patients who can be identified (including by the patients themselves). Please state "Written informed consent has been obtained from the patient(s) to publish this paper" if applicable.

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Abbreviations

The following abbreviations are used in this manuscript:

MDPI   Multidisciplinary Digital Publishing Institute
DOAJ   Directory of open access journals
TLA   Three letter acronym
LD   Linear dichroism

Appendix A

Appendix A.1

The appendix is an optional section that can contain details and data supplemental to the main text—for example, explanations of experimental details that would disrupt the flow of the main text but nonetheless remain crucial to understanding and reproducing the research shown; figures of replicates for experiments of which representative data are shown in the main text can be added here if brief, or as Supplementary Data. Mathematical proofs of results not central to the paper can be added as an appendix.

Table A1. This is a table caption.

Title 1	Title 2	Title 3
Entry 1	Data	Data
Entry 2	Data	Data

Appendix B

All appendix sections must be cited in the main text. In the appendices, Figures, Tables, etc. should be labeled, starting with "A"—e.g., Figure A1, Figure A2, etc.

References

1. Author 1, T. The title of the cited article. *Journal Abbreviation* **2008**, *10*, 142–149.
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3. Author 1, A.; Author 2, B. *Book Title*, 3rd ed.; Publisher: Publisher Location, Country, 2008; pp. 154–196.
4. Author 1, A.B.; Author 2, C. Title of Unpublished Work. *Abbreviated Journal Name* year, *phrase indicating stage of publication (submitted; accepted; in press)*.
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