

When Systems Break, VANNA Prime Takes Control

# 1. Executive Summary

Virtually Autonomous Neural Network Applications (VANNA Prime), developed by DeepFlight, is a next-generation fallback flight control system designed to safeguard public safety and civilian UAV missions following partial system failures. While current commercial UAV controllers provide excellent stability in standard operations, they lack the adaptability to respond when unpredictable or dangerous conditions arise. This limitation can jeopardize critical operations, from search-and-rescue missions to agricultural surveys, where mission failure can lead to significant loss of time, resources, and even human life.

VANNA Prime addresses this challenge by providing an intelligent, real-time fallback controller. It continuously monitors system state while remaining passive during nominal conditions and takes over autonomously if damage or signal loss is detected. VANNA's neural network—based architecture is trained on simulated failure scenarios, allowing it to rapidly recognize abnormal flight states and reconfigure thrust or control commands to recover or safely terminate a mission.

By integrating features such as thrust reallocation, system overrides, and dynamic fault response, VANNA Prime offers public agencies, first responders, and industrial operators a robust layer of resilience far beyond what traditional PID-based controllers or rigid failsafes can deliver. Early testing in simulation environments has demonstrated VANNA Prime's ability to recover from propeller loss, sensor failures, and severe control asymmetries, which would typically ground or destroy a conventional UAV.

Over the coming year, DeepFlight will focus on integrating VANNA Prime with common commercial UAV autopilots, testing real-time anomaly detection in field conditions, and scaling its architecture for applications ranging from emergency disaster response to precision agriculture. With the growing role of drones in public safety, disaster relief, and essential commercial industries, VANNA Prime offers a transformative path to safer, more resilient, and more autonomous UAV missions.

DeepFlight is seeking commercial and public-sector partnerships to help transition VANNA Prime from advanced simulation to full-scale field validation. Support will enable sensor integration, hardware prototyping, and scenario testing to ensure VANNA Prime is ready to protect missions that matter most.

# 2. Background

**Project Short Title: VANNA** 

Project Title: Virtually Autonomous Neural Network Applications

**Project Overview:** This project proposes the development of a cost-effective and modular flight algorithm capable of allowing unmanned aerial vehicles (UAVs) to operate despite sudden malfunctions (e.g., propeller loss, sensor failure, etc.) The algorithm aims to leverage artificial intelligence to adapt thrust and control responses to respond to damage detected in real time, with significant implications to the aerospace industry.

**Technical Objectives/Deliverables:** Demonstrate UAV's ability to maintain flight and autonomously return to base following mid-mission damage (e.g., rotor loss, GPS jamming, or IMU spoofing), with >90% mission completion rate over 20 field trials

# 3. Project Rationale

Maintaining operational capability after partial UAV failure is essential for mission success in public safety, disaster response, and commercial operations. Whether supporting first responders searching for missing people, delivering emergency medical payloads, or surveying vital crop health, UAVs cannot afford to fail due to unexpected damage or degraded conditions.

From local emergency management agencies to commercial agricultural operators, all levels of modern UAV deployment increasingly rely on dependable, adaptable systems that can withstand damage, sensor failures, or environmental disruptions. These environments often include GPS-denied zones, heavy electromagnetic interference, or physical threats that can compromise critical flight components.

Today's conventional control solutions — including PID-based flight controllers, fixed failsafe routines, and redundant hardware — are limited by static logic and cannot adapt to dynamically changing damage profiles. These architectures lack the ability to recognize abnormal flight behavior, classify failure states, and reconfigure control logic in real time. As a result, recoverable failures too often become total mission losses.

The Federal Aviation Administration (FAA) emphasized that as commercial drones grow in scale and sophistication, their safety and resilience must keep pace. According to a 2023 FAA report, "expanding beyond-visual-line-of-sight (BVLOS) operations and public safety missions will require advanced flight risk management capabilities, including automatic contingency responses to component failure" [1].

**Table 4-1** underscores the constraints of existing UAV controllers. These limitations highlight why a fallback control system like VANNA Prime is no longer optional but vital to meet the evolving demands of commercial and public safety operations.

## 4. Introduction

As the adoption of unmanned aerial vehicles (UAVs) accelerates across public safety, disaster response, and commercial operations, the demands placed on these systems have grown rapidly. Conventional UAV flight controllers, typically built on PID-based routines and rigid failsafe protocols, deliver reliable and efficient flights under nominal conditions. However, these systems remain unprepared for unpredictable disruptions, which are common in complex mission environments.

Current technologies, including standard autopilot controllers, return-to-home routines, and redundant hardware failovers, excel at controlling aircraft when everything goes as planned. However, under challenging conditions such as signal jamming, sudden component failures, or sensor spoofing, UAVs can enter degraded states that conventional control architectures are unable to detect or recover from. These systems lack the ability to classify abnormal flight behavior or adapt control strategies in real time, leading to mission failure, potential equipment loss, or in the worst cases, jeopardizing lives during time-critical operations.

Table 4-1 – Current Capabilities of Commercial UAV Control Systems

Current Technology	Benefits	Limitations
Conventional PID Controllers	<ul> <li>Simple, time-tested</li> <li>Efficient for nominal flight conditions</li> </ul>	<ul> <li>Cannot adapt to damage scenarios</li> <li>Fail under sensor blackout or actuator failure</li> </ul>
Predefined Fail-Safes	<ul> <li>Enables basic safety behaviors (e.g., Return-to- home, Forced Landing)</li> </ul>	<ul> <li>Do not account for aerodynamic changes</li> <li>Often terminate missions prematurely</li> </ul>
Redundant Sensor System	Provides backup data streams for fault tolerance	<ul> <li>Adds system complexity and weight</li> <li>Does not solve asymmetric control or actuator issues</li> </ul>

These foundational systems lack the resilience needed for UAVs operating in public safety scenarios, agricultural missions, or critical infrastructure inspections. Without a contingency controller capable of recognizing failures and reconfiguring flight behavior on the fly, even minor anomalies can escalate into total mission loss.

VANNA (Virtually Autonomous Neural Network Applications) was developed to address this shortfall by augmenting standard UAV controllers with a flexible, intelligent fallback capability. By monitoring the UAV's flight status continuously and activating only when a failure is detected,

VANNA Prime ensures that time-critical missions, such as searching for missing persons, delivering emergency supplies, or surveying vital farmland, can continue or land safely even in the face of unexpected damage.

This white paper details the architecture, planned development milestones, and validation efforts for VANNA Prime, providing a clear overview of its capabilities and its potential to transform the resilience of next-generation commercial and public-safety UAV systems.

## 5. Solution Overview

To address these critical gaps, VANNA Prime integrates a modular, neural network—based fallback control system that works alongside standard UAV autopilots. By remaining passive during nominal conditions and activating only when flight anomalies occur, VANNA Prime empowers commercial UAV platforms to recover from partial damage or system failures without immediate human intervention. Its architecture is designed for seamless integration into existing flight stacks and is capable of reconfiguring flight controls dynamically, preserving mission success or enabling a safe landing even under adverse conditions.

**Figure 5-1** provides a visual breakdown of the VANNA system architecture, illustrating how to control transitions from the baseline autopilot to VANNA under failure conditions.

(Standard Autopilot) **Baseline** Flight System **Passive Monitoring** Layer (VANNA) [IF] [IF] **NOMINAL** CONTINGENCY **CONDITIONS** TRIGGERED VANNA Neural **Control** (RNN + CFD-trained Model) Thrust/Actuation Subsystem

Figure 5-1 – VANNA System Diagram

#### **VANNA Integration Strategy & Risks/Mitigations:**

To execute this contingency response, VANNA is structured into **five subsystems**. Each component plays a critical role in detecting instability, evaluating conditions, and issuing corrective control outputs.

- **State Monitoring Module:** Detects anomalies by tracking IMU, GPS, and power data for rotor drop, signal loss, etc.
- **CFD-Guided Neural Control:** Predicts control commands using a neural network trained on simulated flight data.
- Thrust Reallocation Logic: Redistributes thrust to compensate for asymmetric conditions (e.g., damaged propeller).
- **Control Transition Layer:** Switches control from the default autopilot to VANNA when failure thresholds are passed.
- **Simulation-to-Reality Transfer Pipeline:** Refines VANNA behavior by blending synthetic data with real-world flight feedback.

Table 5-2 – VANNA Risks & Mitigations

Risk	Impact	Mitigation
AI doesn't respond well to unexpected damage	May fail to stabilize the UAV if the damage type wasn't seen during training	Keep improving the AI using real flight data to teach it how to handle more damage types
Triggers fallback mode when there's no real issue	Could override a healthy autopilot and cause instability	Use better thresholds and combine data from multiple sensors to avoid false alarms
Slow reaction time after failure	May not correct the issue fast enough, risking a crash	Optimize code and hardware to reduce delay when switching to fallback mode

## 6. Future Direction

Although the current implementation of VANNA Prime demonstrates the viability of a learning-based fallback control system for commercial UAVs, its modular architecture, designed from day one for extensibility, enables targeted upgrades to nearly every subsystem. With proper funding and suitable flight-test environments, VANNA Prime can be adapted rapidly for deployment in increasingly complex public safety, agricultural, and commercial scenarios. With an estimated budget of \$15,000, we can implement the following upgrades to improve VANNA Prime's real-world adaptability, robustness, and operational readiness:

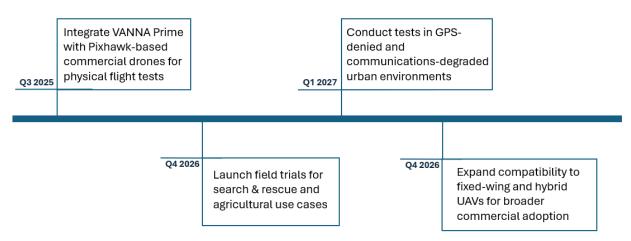
#### **Near-Term Upgrades (Next 3–6 Months):**

Integrate VANNA Prime with drone simulators to evaluate scenarios including GPS jamming, propeller damage, and sensor spoofing.

Extend the training data library with agricultural, urban, and emergency mission profiles to enhance domain generalization.

Develop a to reflight state classifier cognize abnormal flight conditions and reliably trigger fallback transitions.

#### **Strategic Timeline (12 Months):**



These milestones will position VANNA Prime for commercial field readiness and support the broader objective of bringing intelligent contingency flight control to high-stakes civilian missions.

As VANNA Prime transitions beyond prototyping, our long-term efforts will focus on scaling its deployment across multiple commercial UAV sectors, from public safety to precision agriculture, in close collaboration with industry and academic partners.

# 7. References

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7. References

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