An Intelligent Propulsion Control Architecture to Enable More Autonomous Vehicle Operation

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This paper describes an intelligent propulsion control architecture that coordinates with the flight control to reduce the amount of pilot intervention required to operate the vehicle. Objectives of the architecture include the ability to: automatically recognize the aircraft operating state and flight phase; configure engine control to optimize performance with knowledge of engine condition and capability; enhance aircraft performance by coordinating propulsion control with flight control; and recognize off-nominal propulsion situations and to respond to them autonomously. The hierarchical intelligent propulsion system control can be decomposed into a propulsion system level and an individual engine level. The architecture is designed to be flexible to accommodate evolving requirements, adapt to technology improvements, and maintain safety.

Nomenclature

IA = Increasing Autonomy
 IFC = Intelligent Flight Control
 IPC = Intelligent Propulsion Control
 IMM = Intelligent Mission Management
 OEI = One Engine Inoperative

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I. Introduction

ECONOMIC pressure on airlines and a pilot shortage¹ are providing the impetus to reduce the size of the cockpit Crew, and eventually remove the pilot altogether, once the idea gains acceptance with the flying public.² With the current high level of automation, it might seem like a short leap to a single pilot in the cockpit, to a fully unpiloted vehicle, or to a single pilot on the ground monitoring multiple vehicles at once. Currently, in highly automated modern commercial aircraft, pilots spend an average of under seven minutes per flight actually flying the aircraft.³ This hand flying occurs almost exclusively during takeoff and landing,⁴ which is also when most accidents and incidents take place.⁵ It should not be surprising that lack of participation in the flying process can lead to boredom, which can lead to a decrease in performance.⁶ High profile accidents^{7,8} point out not only the loss of piloting skills, but also confusion about the automation itself.⁹ However, pilots can also identify extremely unusual emergency situations and demonstrate tremendous skill in reacting to them.^{10,11,12} This reveals a paradox: while most commercial aviation accidents are attributed to human error,¹³ humans can also prevent accidents through heroic action that cannot be anticipated or learned through training. Therefore, any system that replaces the pilot flying must not only perform as well as or better than the human on mundane tasks, it must exceed the human in extraordinary situations, or risk that the promise of autonomous aircraft will not be fulfilled.

Approaches to the design of such a system can be through either comprehensive automation, or automation that has the ability to reason about situations it was not explicitly designed to address. The latter approach would involve the development of sufficient vehicle self-awareness to at least replace the pilot's senses when it comes to diagnosing problems.

Autonomous operation of civil aircraft will require the propulsion control to be highly integrated with the flight control to remove the need for the pilot to serve as the integrator. This integration can also provide an added benefit of overall system optimization. Although recognition of off-nominal situations in the propulsion system is one of the capabilities at which pilots excel, pilot error is still cited as the cause of most accidents. Therefore, on-board autonomy algorithms will need to be able to recognize and respond to off-nominal propulsion situations even better than pilots do today. There are numerous examples of propulsion system malfunction plus inappropriate crew response¹⁴ leading to accidents. Addressing these situations through greater engine state awareness and better communication to the pilot is the initial step, with the autonomy taking on more responsibility over time.

The specific goal here is to develop an intelligent engine control architecture that enables the propulsion system to work harmoniously with the flight control, with reduced pilot intervention required as the architectural framework is populated. It will automatically recognize the vehicle operating mode and configure the engine control scheme to provide optimal performance for that mode based on knowledge of engine condition and capability. The propulsion control will take cues from the flight control to operate autonomously. It enables "personalized" control of an engine, which will facilitate the ability to assess the current engine condition and its performance. The architecture provides an interface with the flight control to communicate engine capability, receive engine thrust commands and set the engine into the appropriate control mode; it also enables safe accommodation of "off-nominal" conditions by intelligent decision making to replace decisions currently made by the pilot – such as whether to shut down an engine, continue operating at reduced performance, etc. Increased system integration will result in enhanced information that will ultimately improve safe decision making. There will be a natural progression from the current situation where the pilot monitors and controls the engine to a fully autonomous propulsion system. This approach will maintain safety and still be compatible with an increasingly autonomous flight control.

The remainder of this paper consists of a description of the architecture, followed by several examples of the partially populated architecture implemented and operating in a simulation testbed to identify and mitigate safety concerns. A discussion of the examples and a lessons learned section come next, followed by conclusions.

II. Intelligent Propulsion System Control Architecture

With advancements in computing and information systems technology, it is reasonable to assume that there will be a natural progression from the current cockpit procedures to fully autonomous aircraft operation. The intelligent propulsion system control architecture must be flexible enough to accommodate evolving requirements while maintaining safety (note that some recent accidents have been attributed to loss of piloting skills due to over-reliance on automation¹⁵). Therefore, the types of functions currently performed by pilots, as well as those that would need to be performed with no pilot present, must be addressed. These types of functions include detecting, diagnosing, ¹⁶ and accommodating faults (Figure 1), and reducing pilot workload in high stress situations. They can extend to optimizing aircraft performance by coordinating the flight and propulsion control under certain situations, without an explicitly integrated flight and propulsion control design approach. An intelligent propulsion control architecture has been developed, as shown in Figure 2. The intelligent propulsion control comprises the engine systems with their individual

	Engine Separation	Severe Damage	Surge	Bird Ingestion/FOD	Seizure	Flameout	Fuel control problems	Fire	Tailpipe fires	Hot Start	lcing	Reverser Inadvertent Deploy	Fuel Leak
Bang	0	Χ	Χ	0	0						0		
Fire warning	0	0		0				Χ					
Visible flame	0	0	0	0				0	Χ	0			
Vibration		Χ	0	Χ	0						Χ	Χ	
Yaw	0	0	0	0	0	0	0					Χ	
High EGT		Χ	Χ	0	0		Χ		0	Χ	0		
N1 change	Χ	Χ	0	0	Χ	Χ	Χ						Χ
N2 change	Χ	Χ	0	0	Χ	Х	Χ						Х
Fuel flow change	Χ	0	0		0	Х	0	0					Х
Oil indication change	Χ	0	0		0	Х		0					
Visible cowl damage	Χ	Χ						0				Χ	
Smoke/odor in cabin bleed air		0		0	0								
EPR change	Χ	Χ	Х	0	Χ	Χ	Χ						Х

X = Symptom very likely

O = Symptom possible

Note: blank fields mean that that symptom is unlikely

Figure 1. Propulsion system malfunctions and observed symptoms (from Ref [16]).

controls at the lowest level, along with the propulsion system level above them. In parallel is a presumably similar, although undefined, flight control structure (this flight control structure should support such functionality as reconfiguration in response to a stuck actuator or damaged flight control surface, ¹⁷ for example). Above these two structures is an Intelligent Mission Management (IMM) level, which coordinates the flight and propulsion control as necessary. This architecture maintains the traditional roles of distinct flight and propulsion control systems as is standard practice today, but supports integrated flight and propulsion control as necessary to accommodate faults or optimize performance. Thus, the architecture is suitable for a progression from current practice to a more fully integrated design, as might be expected in the future. To enable this progression, more and more of the intelligent decision-making pilot function can be done autonomously, similar to how automation has assumed much of the role of flying today.

The hierarchical nature of this architecture is advantageous, since it can be populated from the bottom up, meaning that existing subsystem and engine level diagnostic algorithms can be implemented initially, while higher level functions that depend on them to enable autonomous operation can be added later. This strategy builds on years of development of algorithms for engine health management, diagnostics, prognostics, estimation, and control, and directly supports the concept of increasing autonomy (IA). Increasingly autonomous systems are defined as lying along the spectrum of system capabilities that begin with the abilities of current automatic systems, such as autopiloted and remotely piloted (nonautonomous) unmanned aircraft, and progress toward highly sophisticated systems. For the purposes of this work, autonomy is defined as a process that occurs without human intervention, and increasing autonomy means that the complexity or abstraction of the procedure increases. For instance, the appropriate pilot responses to a variety of engine malfunctions are defined in the aircraft flight manual. These checklist procedures define the sequence of steps that pilots are to follow in the event of a given malfunction. With IA, the engine level algorithms might help diagnose a problem, which could lead to execution of the appropriate checklist. Eventually, as the higher levels of the architecture are populated with algorithms, evaluation of the situation and reasoning about any actions that are currently up to the pilot's discretion could be achieved. The

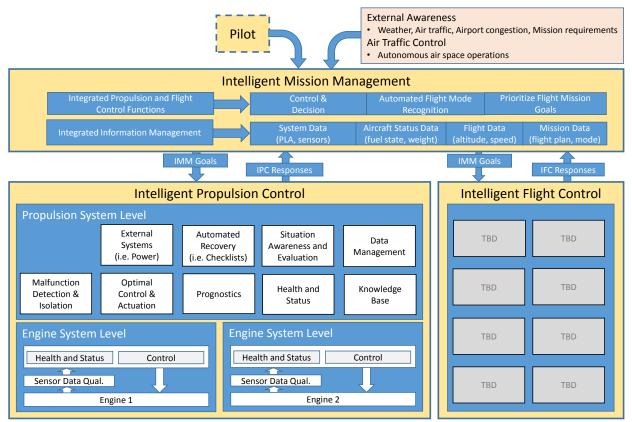


Figure 2. Intelligent Propulsion Control Architecture.

distinction between intelligence and autonomy is somewhat blurred, but here it means that the diagnostic, prognostic, and—to an extent—accommodation algorithms build on and support each other to produce actionable knowledge about the engine and its operational capabilities (intelligence), which can be used to re-plan the mission or perform other higher level functions (autonomy).

Basic definitions of the functional blocks comprising the architecture in Figure 2 are described next.

Intelligent Mission Management (IMM)

Integrated Propulsion and Flight Control Functions

- Control and Decision Issues command goals to the intelligent control elements and monitors progress.
- Automated Flight Mode Recognition Determines the phase of flight.
- Prioritize Flight Mission Goals Specifies the overall flight mission objectives.

Integrated Information Management

- System Data Sensor data and cockpit control setting information.
- Aircraft Status Data Information on the aircraft status (e.g., weight, fuel quantity).
- Flight Data Information on the status of the flight (e.g., vertical speed, airspeed, altitude, bank angle).
- *Mission Data* Information of the overall flight (e.g., planned cruising altitude, current flight segment, planned airspeed).

Goals and Responses

- *IMM Goals*—"Goals" sent from the Intelligent Mission Management level to the lower level control elements. These are considered goals rather than commands so that each individual intelligent control element can determine its own optimal approach to satisfy the IMM objective.
- *IPC/IFC Responses* Status of the goal response sent back to IMM. Provides information on whether the goal is achievable and its status, allowing the IMM to determine if the command goals should be modified to something more achievable.

Intelligent Propulsion Control (IPC)

- *Propulsion System Level* Intelligence control level that oversees the entire propulsion system. Most of the propulsion intelligence resides at this level. A process continuously performs the Observe, Orient, Decide and Act (OODA) decision loop on the command goals for the propulsion system.
 - o *External System* Controls engine auxiliary output, which includes electric power generation and cabin bleed air.
 - Automated Recovery Performs automated pilot check list actions for events to reduce pilot workload.
 - Situation Awareness and Evaluation Determines situation and decides if current trends are desirable or if further action is necessary.
 - o Data Management Manages the data flow throughout the IPC.
 - o *Malfunction Detection and Isolation* Actively senses and implements mitigating actions (i.e., executes the appropriate checklists) for propulsion system faults.
 - Optimal Control and Actuation Determines how the engines can be operated to achieve a
 desired goal such as increased efficiency or performance considering condition of the engines,
 current flight environmental conditions and the given flight plan.
 - Prognostics Forecasts future propulsion system health status (i.e., life available). May estimate time to failure during an engine fault (e.g., time to engine seizure resulting from an oil leak).
 - o *Health and Status* Determines and monitors overall propulsion system health status.
 - *Knowledge Base* A repository of information used by the IPC to made decisions. It can be updated with information that has been "learned."
- *Engine System Level* The level is mostly an interface to the engine controller with some logic to enhance functionality.
 - o *Health and Status* Determines engine health status.
 - Control An interface to the engine controller (Full Authority Digital Engine Control, or FADEC).
 - Sensor Data Qualification Evaluates engine sensor data to validate that the data are correct.

Intelligent Flight Control (IFC)

• Function elements are TBD.

There are multiple advantages of this architecture: pilot workload can be reduced; flight safety can be improved; aircraft performance can be enhanced; flight operations can be accomplished in a more efficient and cost effective manner; and actionable knowledge can be produced and exploited.

III. Demonstration

Intelligent propulsion control provides oversight, diagnostics, advice, and automated responses, so it can work with the pilot or on its own to improve safety and performance. Knowledge of the engine state can be critical in emergencies because it can provide guidance for an appropriate response. In the Kegworth air disaster of 1989,²⁰ a fan blade on a British Midland-operated Boeing 737-4Y0 detached, resulting in heavy vibration and smoke in the cabin. The pilot misdiagnosed the problem engine based on an incorrect understanding of the ventilation system, and shut down the wrong engine, unknowingly continuing to fly on the bad one. Eventually, the damaged engine ceased operating and burst into flames, resulting in a deadly crash. In another example, TransAsia Airways Flight 235, a twinengine turboprop, experienced an engine flameout on takeoff. The flight crew did not perform the appropriate failure identification procedure, resulting in confusion regarding the identification and nature of the actual propulsion system malfunction. The pilot reduced power on the working engine and then shut it down.²¹ The aircraft climbed, and before the engines could be restarted, it stalled. The altitude was insufficient for recovery, and the plane crashed.

With knowledge of the engine state, the preliminary implementation of the intelligent propulsion control architecture is able to flag engines as good/working or bad/nonworking/failed, so that a pilot-initiated in-flight shutdown of a good engine is inhibited or at least the pilot is alerted. The following examples focus on situations that are easy to visualize; the underlying algorithms and their resulting diagnoses and actions represent only a small subset of the potential features of the fully populated architecture.

A. Testbed

The testbed used for evaluation was the Transport Class Model (TCM)²² airframe simulation integrated with two copies of the Commercial Modular Aero-Propulsion System Simulation 40k (C-MAPSS40k)^{23,24} engine model in the NASA Glenn flight simulator.²⁵ The TCM represents a generic, mid-sized, single-aisle transport aircraft with two wing-mounted engines. C-MAPSS40k models a 40,000-lbf thrust class, high-bypass ratio turbofan engine with a representative commercial-type controller. There have been multiple control and diagnostic algorithms demonstrated using the combined simulation,²⁶⁻²⁸ or C-MAPSS40k²⁹⁻³¹ by itself, so the initial implementation of the architecture framework was in place.

B. Examples

The examples presented below involve engine malfunctions and shutdowns, and their impact on the vehicle. Commercial aircraft are designed to fly safely with an engine out. Normally with both engines functioning, the throttles are moved together to produce symmetric thrust. When there is one engine inoperative (OEI), the pilot is presented with flying challenges due to the off-centerline thrust from the remaining engine, which affects control of the aircraft in yaw, pitch and roll. Flight control surfaces are designed to be large enough to counteract this imbalance, but will incur an associated performance penalty. Because flight control effectiveness is related to airspeed, thrust imbalance is more difficult to counteract when the aircraft flies slowly. At faster airspeeds, flight control authority is improved and the aircraft is much easier to manage.

When there is an OEI with a twin engine aircraft, there is the expected loss of 50% of the available thrust, but there is a more significant loss of performance from the aircraft flying in an uncoordinated fashion. There is added drag resulting from the flight control inputs needed to offset the thrust imbalance, and from the drag penalty originating from the inoperative engine. These performance impairments must be managed as much as possible by the pilot to reduce their impact.

When an engine fails, it may be difficult to determine which one it is initially. Because of windmilling of the inoperative engine, both engines may appear to be functioning, since the instruments of the failed engine have readings that may make it seem as though it is still functioning.

The above background provides some context for the following three examples.

1. Propulsion System Malfunction Detection and Identification

The first case demonstrates propulsion system diagnostics. Here, the right engine is severely deteriorated and cannot produce as much thrust as the left engine. This is clear from the top two plots in Figure 3. While the throttles are moved together to the same power settings, the estimated thrust³² of the two engines is very different, especially at high power. The third plot shows that this leads to a temporary declaration of asymmetric thrust, which is subsequently unflagged as power is reduced and the thrust mismatch decreases at about 46s. However, a power demand increase at about 85 s again results in an asymmetric thrust declaration. The right engine begins to shut down at about 92 s, and the output of the shutdown detection logic that flags it is shown in the fourth plot.

2. Engine Shutdown Inhibition

This second example exercises different aspects of the architecture, using the same set of simulated flight data as the first example, to demonstrate how a commanded engine shutdown can be inhibited under unsafe conditions. The logic bases safety for such an action on a minimum airspeed or height above ground (AGL), either 135 knots or 1000 feet AGL in this case. Here the pilot attempts to shut down the engine after observing the abnormal behavior in the previous example. As shown in Figure 4, the command is given at about 83 s, but because the aircraft is too low and too slow according to the oversight logic, the shutdown is inhibited until about 93 s, when the system deems that the flight conditions are safe. This adds a layer of protection, but can be overridden by the pilot, if necessary.

3. One Engine Inoperative Vehicle Upset Recovery

The final example demonstrates minimum control airspeed protection against loss of control for an engine out situation (Figure 5). In this example, the right engine goes out (at about 10 s) and the airspeed begins to reduce. As the speed drops below the minimum control airspeed, the rudder loses effectiveness, making it unable to counteract the thrust asymmetry, and the plane begins to roll into an upset condition. However, in this case, the good engine's thrust is cut back temporarily to remove the asymmetry and arrest the roll departure. This happens a couple of times (at about 20 and 30 s), at which point the airspeed has increased enough due to a reduction in pitch to return rudder effectiveness allowing the aircraft to recover.

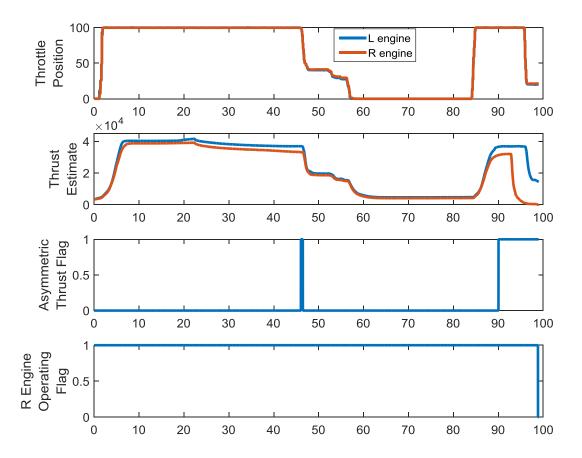


Figure 3. Engine Diagnostics Demonstration.

IV. Discussion

These examples in the previous section build on several technologies. The engine out situation itself is diagnosed without pilot intervention using detection logic related to specific engine variables (note that this algorithm operates independently of the pilot action to shut down the engine). The asymmetric thrust condition, meanwhile, is detected using a thrust estimation scheme. These two diagnoses confirm each other and, taken together, provide knowledge about the engine state, so if a loss-of-control situation becomes imminent, a recovery procedure is initiated.

Previous work has demonstrated that the engine state can be used to determine to what extent engines can augment the flight control to help in recovery of a damaged aircraft.³³ It has been shown that engines used differentially and together can be effective flight control actuators.³⁴ Advanced propulsion control schemes that take advantage of unused safety margins can produce marked improvement in aircraft stability over that provided by traditional propulsion control in these situations. Furthermore, remapping commands from a damaged flight control system to the propulsion system can provide significant handling improvements as compared to manual throttle manipulation.²⁸ The intelligent propulsion control architecture has the potential to incorporate such engine capability under the right circumstances.

Pilot checklists are used as the starting point for autonomous operation, although higher level decision-making is often necessary for taking an action. If the situation generally calls for an in-flight engine shutdown, the appropriate checklist would be executed. However, before doing so, there are considerations regarding the aircraft's altitude that must be taken into account, as a pilot would, and the shutdown would probably not be initiated with low airspeed close to the ground as in the example above. Furthermore, for emergency situations, the decisions to initiate these actions could require higher level reasoning capability rather than predefined limits.

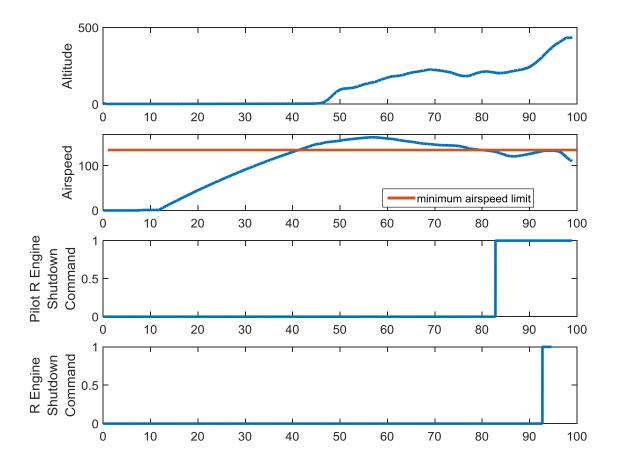


Figure 4. Example engine shutdown inhibition.

Previous research has demonstrated many of the fundamental algorithms required for the architecture to be effective. By populating the architecture with these existing algorithms and integrating them appropriately, several applications of intelligent propulsion control can be readily demonstrated that enable reduced pilot workload, enhanced vehicle performance, and improved engine diagnostic capability. These applications include:

- Minimum control airspeed protection against loss of control for an engine out situation
- Estimation of unmeasured engine variables
- Model-based engine control
- Asymmetric thrust detection and confirmation
- Detection of a disconnect between the engine power level and throttle position (i.e., stuck throttle)
- Automation of pilot checklist for engine in-flight shutdown
- Inhibition of incorrect engine shutdown

Additionally, previous implementations of integrated flight and propulsion control have demonstrated improved performance, including:

- High crosswind landings
- Augmentation of flight control with propulsion to compensate for stuck, limited, and ineffective flight control surfaces

V. Lessons Learned

The architecture is simply a framework. It displays appropriate hierarchy and interaction to offer guidance to the designer, but populating it is still up to the designer. The literature contains many examples of diagnostic and control algorithms that work well individually, so a bottom up approach is logical. However, the details of the integration were not always obvious. Use of multiple diagnostic approaches provided confidence in a conclusion, but much of the

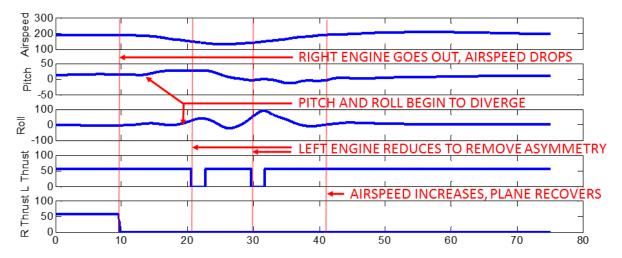


Figure 5. Minimum control airspeed protection example.

higher level functionality was still ad hoc. Pilots and checklists provided direction, but there are many variations on scenarios, and pilot experience is anecdotal, depending significantly on the details of the situation. Furthermore, rare events, or those having no precedent, still require a response that does not make matters worse, which is currently more of a philosophy than a working principle. Thus, the development of this instance of the intelligent propulsion control system relied heavily on combining past work that had been developed in the test bed system, although some higher level responses to situations were developed and refined based on testing (e.g., the low speed OEI example).

It was found that the intelligent propulsion control provides significant advantages during off-nominal situations by improving situational awareness of the pilot and by inhibiting inappropriate crew responses (e.g., reducing the "startle" factor, or helping to assess conflicting information). Conflict resolution logic is important because the sub-elements may generate opposing diagnoses/commands, and an awareness of the flight phase is needed to configure the operation of the sub-elements and to properly set logic priorities.

VI. Conclusions

An intelligent propulsion control system architecture that enables increased aircraft autonomy has been defined, partially implemented, and demonstrated on some basic scenarios. Some existing algorithms are included, as well as logic to tie the results together and act on them. This structure must still be built up to include higher level functionality that can properly take advantage of the information available. Safety needs to be a major concern in the design of these systems because of its relationship to public confidence. The placeholders in the architecture focus heavily on safety, although it is up to the designer to populate them. With increased integration and information availability, there is also the opportunity to optimize performance. The architecture is designed to accommodate types of algorithms that can coordinate automated, preplanned maneuvers better than a human pilot. As society moves closer to notional acceptance of unpiloted commercial aircraft, primarily as a cost saving measure, the issues addressed by the architecture described herein will become increasingly important.

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