

# Spatiotemporal Inferential Intelligence – SII™

## 1. The Problem

Too often machines are required to fail before you can do anything about it.

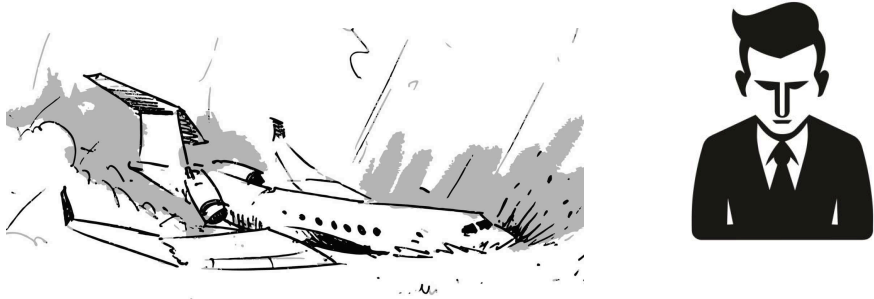


Figure 1. Illustration of a crashed commercial airliner and frustrated engineer

## 2. Innovation

Spatiotemporal Inferential Intelligence (SII™) is a new safety and autonomy paradigm in inferential sensing that enables aircraft and spacecraft to leverage a few sensors to:

- Comprehensively anticipate and mitigate failures and degradations by inferentially monitoring the condition of an exceeding and unprecedented proliferation of systems and parameters
- Autonomously design and perform ad hoc in-space collision avoidance and rendezvous maneuvers by continuously inferring **all** the parameters of the applicable equations of motion for transition based on current and desired orbital conditions

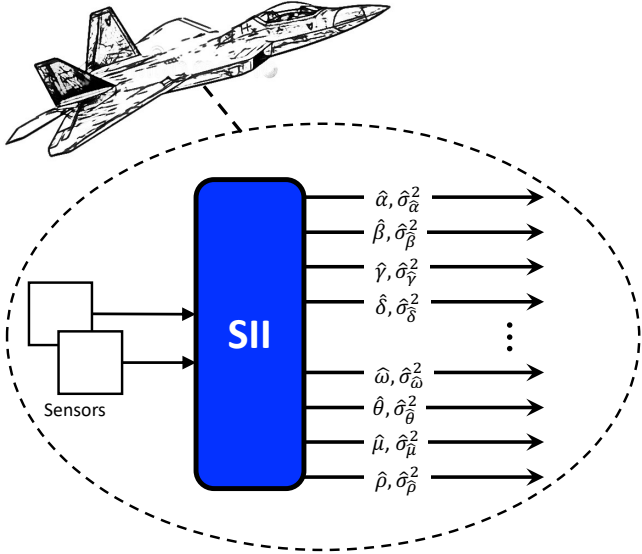
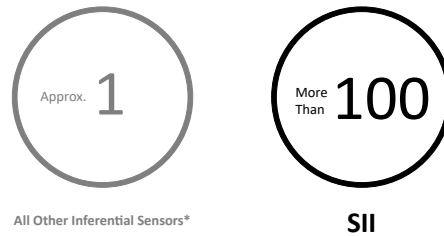


Figure 1. SII extracts an arbitrarily high number of time-varying physical parameter behaviors from only a few sensors, allowing unprecedented decision, control, and autonomy

**2.1. Advantage Relative to the State of the Art (*Secret Sauce*)**

Number of Parameters Inferred per Sensed Measurement



\*E.g.,(State of the Art) Emerson, Honeywell, ABB, Siemens, General Electric (GE), Schlumberger, Rockwell Automation, Microsoft, IBM, SAP, National Instruments (NI)

**Figure 2.** Comparisons of SII’s expansive parameter estimation capability to the state-of-the-art.

**2.2. *Secret Sauce* in *Tech-Speak***

For nonlinear dynamical system  $\mathbf{y}(t) = f(\mathbf{A}(t)\mathbf{x}(t))$ , where vectors  $\mathbf{y}(t)$  and  $\mathbf{x}(t)$  are the system outputs and states, respectively, and matrix,  $\mathbf{A}(t)$ , comprises the time-varying system parameters, then **SII** not only estimates  $\mathbf{x}(t)$ , but also estimates all elements of  $\mathbf{A}(t)$  for any arbitrary size  $\mathbf{A}$ .

$$SII(\mathbf{y}(t)) \rightarrow \hat{\mathbf{A}}(t)\hat{\mathbf{x}}(t) \tag{1}$$

SII effectively treats a stochastic system with observations,  $\{\mathbf{y}(t)\}$ , as a superposition state, which, when measured, collapses to  $\mathbf{A}(t)\mathbf{x}(t)$  (as expressed in Equation 1), with separable observable features  $\mathbf{A}(t)$  and  $\mathbf{x}(t)$ .

This result is impossible for the prior state of the art, which may only estimate a few elements of  $\mathbf{A}(t)$ , roughly equal to the number of elements in  $\mathbf{y}(t)$ .

**2.3. Additional Key Features**

Feature	Benefits
Quantifies estimation uncertainties	Supports statistical analyses and compliance
Auto-tunes control systems in real-time to account for system wear, maintenance, manufacturing tolerances, degradations, and failures	Eliminates frequent inferential sensor software updates, over the life of the system, providing robust safety and reliability

**3. Seeking**

Commercial partner and investment support for:

- Developing demonstrators concerning increasingly sophisticated problems
  - Field testing access concerning control and monitoring problems of commercial interest

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