

Cognitive Systems

High Performance Embedded Computing Workshop

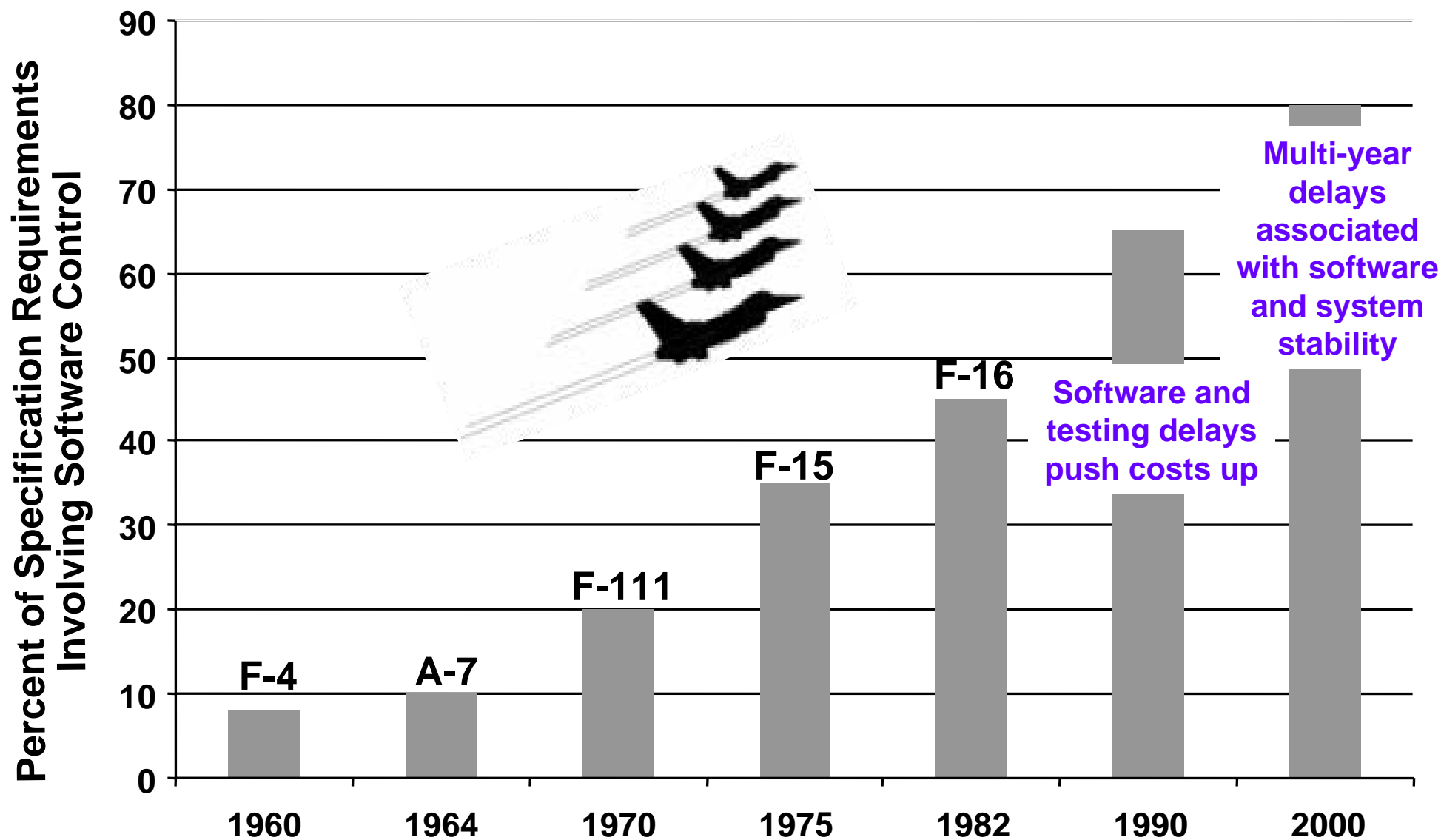
**Robert Graybill
DARPA IPTO
Sept 28, 2004**

- **Computer systems are the backbone of key national infrastructure and critical DoD systems**
 - Virtually all important transactions involve massive amounts of software and multiple computer networks
 - DoD future vision is “network-centric warfare”

- **While *computational performance* is increasing, productivity and effectiveness are not keeping up**
 - Cost of building and maintaining systems is growing out of control
 - Systems have short lifespans with decreasing ROI
 - Demands on expertise of users are constantly increasing
 - Users have to adapt to system interfaces, rather than vice versa

- **As a result, systems have grown more complex, more fragile, and more difficult to develop**

We need to change the game



Developing Cognitive Systems: *Systems that know what they're doing*

- **A cognitive system is one that**
 - can **reason**, using substantial amounts of appropriately represented knowledge
 - can **learn** from its experience so that it performs better tomorrow than it did today
 - can **explain** itself and **be told** what to do
 - can be aware of its own capabilities and **reflect** on its own behavior
 - can **respond robustly** to surprise

- **...reflect on what goes wrong when an anomaly occurs and anticipate its occurrence in the future**
- **...respond to naturally-expressed user directives to change behavior or increase functionality**
- **...be configured and maintained by non-experts**
- **...reconfigure themselves in response to environmental changes and mission events**
- **...reduce the effort to develop and maintain software**
- **...thwart adversarial systems that don't know what they're doing**
- **...preserve “corporate memory” to ease transitions for rotational personnel**

Systems That Know What They're Doing

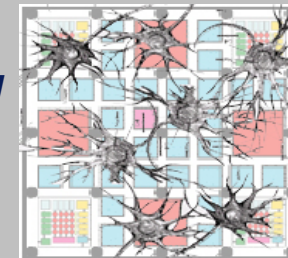
What's Next?

□ Intelligent Systems

- - Architectures for Cognitive Information Processing (ACIP)



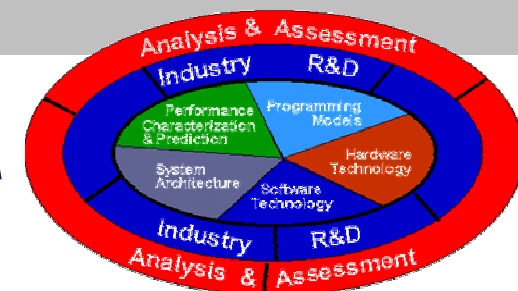
+ Cognitive Processing
Hardware Elements
SBIRs



□ High-End Application Responsive Computing

- High Productivity Computing Systems Program (HPCS)

HPCS + **HECURA**



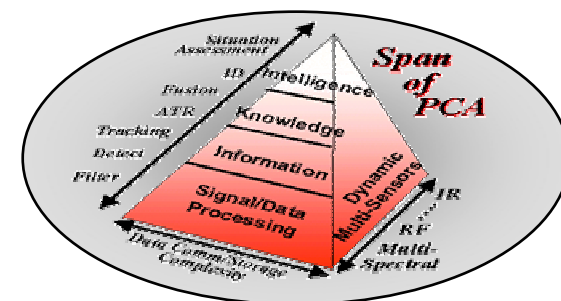
□ Mission Responsive Architectures

- Polymorphous Computing Architectures Program (PCA)



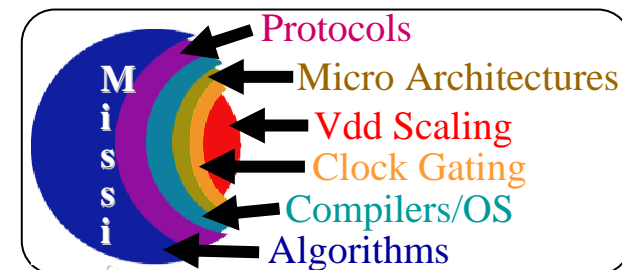
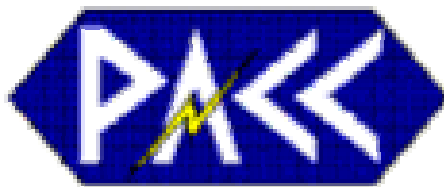
+ OneSAF
Objective System

+ XPCA??



□ Power Management

- Power Aware Computing and Communications Program (PAC/C)



Biological Clues
Cognitive Algorithm Clues
DoD Mission Challenge Clues



ACIP Phase 1
Early Architecture
Concepts & In-
Context Evaluation
- 2 Years -

MTO
3-D Interconnects,
Optical, Nano

Functional
Demonstrations
& Algorithm
Developments

Physical Devices,
Interconnect, and
Packaging

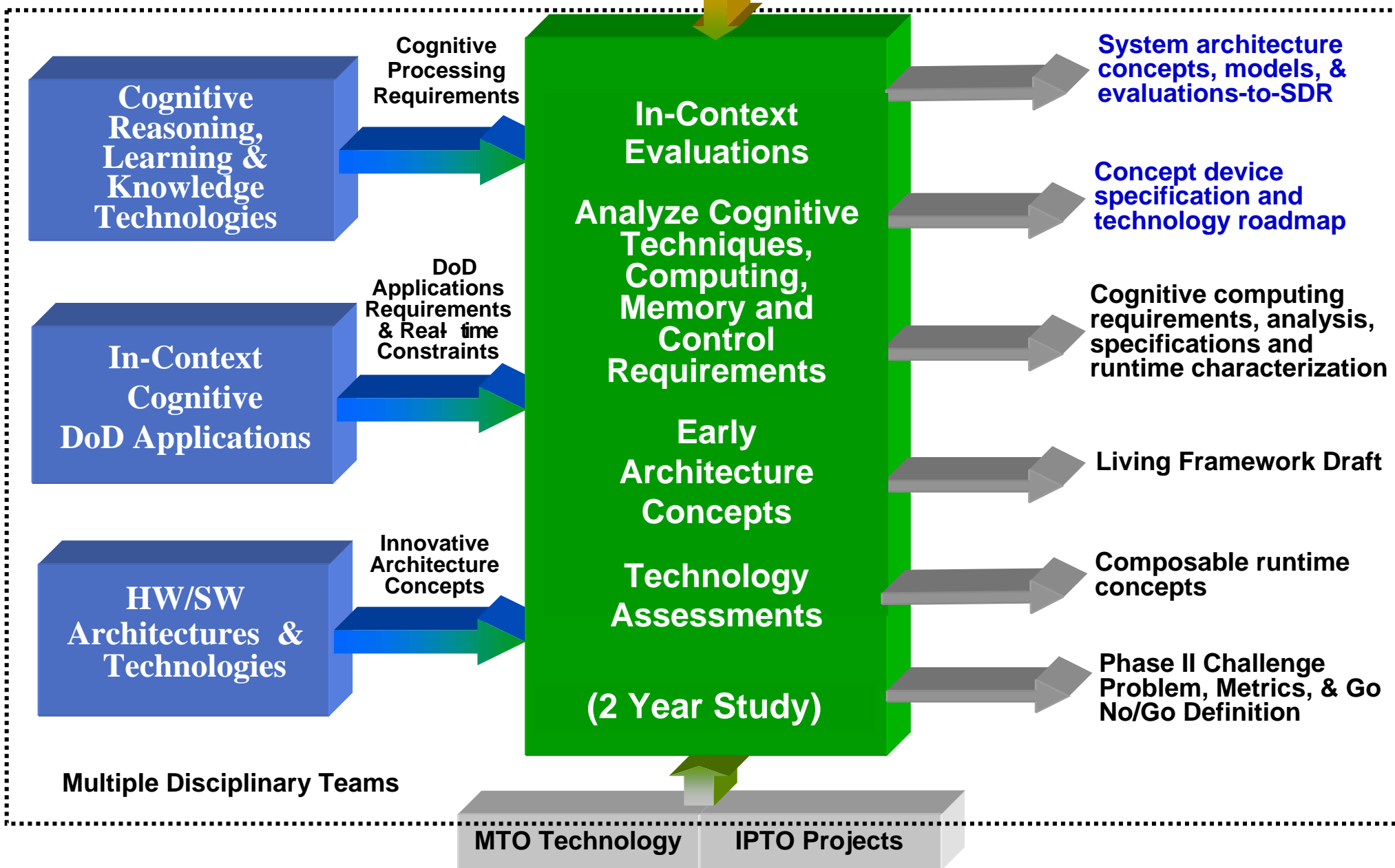
ACIP Phase 2
Full Scale
Implementation and
Demonstration
- 4 Years -

ACIP Phase 3
Cognitive
Technology DOD
System Transitions
- 2 Years -

Study Considerations

Metric Evaluation

Deliverables



- **Fantastic Response!!!**
- **Participation Mix (Including Subs)**
 - **9 Defense Contractors**
 - **11 Research Laboratories**
 - **51 Universities**
 - **30 Commercial Companies**

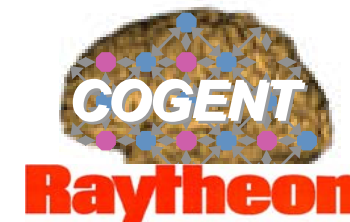
Broad Multi-Disciplinary Coverage required for System Innovation

Large Diverse and Robust Teams Resulted in the Best Concepts

Study Technical Framework Concept has Emerged

- **COGNitive ENGINE Technology (COGENT)**

- Raytheon Company - Network Centric Systems



- **Polymorphous Cognitive Agent Architecture (PCAA)**

- Lockheed Martin Advanced Technology Labs



- **CEARCH: Cognition Enabled Architecture**

- University of Southern California/ISI



- **Reservoir Labs Inc – Cognitive Processing Hardware and Software elements**
- **Intelligent Automation Inc. – Hardware Architectures for Flexible Component Based Hybrid Cognitive Systems**
- **Hoplite Systems LLC – Cognitive Processing Hardware Elements**
- **Cardinal Research LLC – Cognitive Processing Hardware Elements**
- **Saffron Technologies – Associative Memory Hardware Elements for Cognitive Systems (Funded by AFRL)**

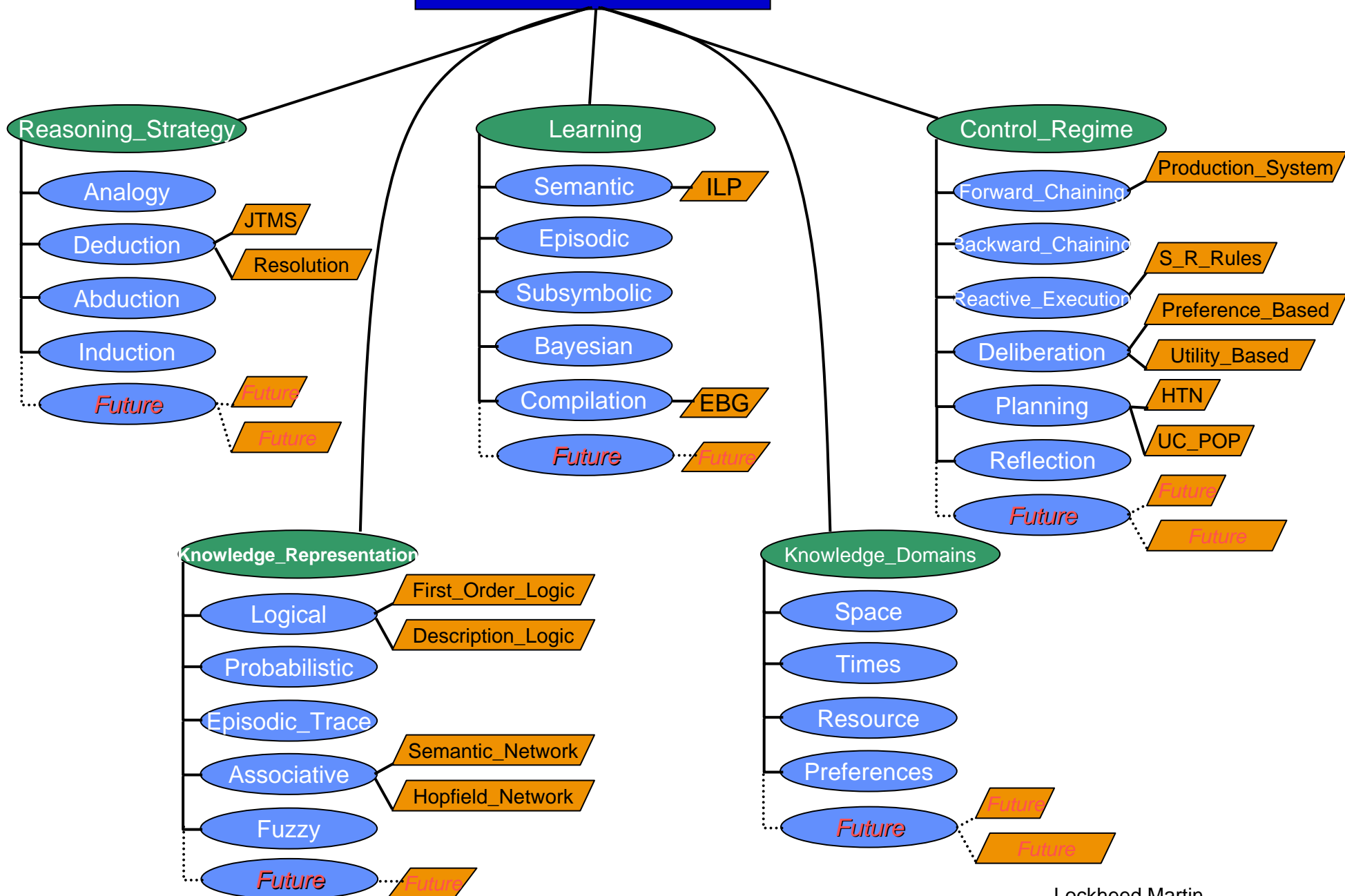
Cognitive Technology Classification

| Reasoning Algorithms | Symbolic (S) Probabilistic (P) Hybrid (H) | Ray | LM | ISI |
|--|---|-----|----|-----|
| 1st Order Reasoning | S | | X | |
| Abductive Reasoning | S,P,H | X | X | |
| Analogical Reasoning | S,H | X | | |
| Bayesian Networks | P | | | |
| Case-based Reasoning | S,H | | | |
| Causal Reasoning | P | X | | |
| Common Sense Reasoning | S | X | | |
| Counterfactual Reasoning | S | | | |
| Deductive Inference | S | X | | |
| Defeasible Reasoning | H | | X | |
| Forward & Backward Chaining | S | | | |
| Fuzzy Reasoning | H | | X | |
| Game Theory - Optimization | H | | | |
| Goal-oriented Planning | S | | | X |
| Heuristic Meta-reasoning | H | | | X |
| Inductive Reasoning | S | | X | |
| Logical Pattern Matching | S | | | |
| Logical Unification | S | | | |
| Markov Processes | P | | | |
| Mathematical Programming | H | | | |
| Maximum Likelihood | P | X | | |
| Meta-meta Reasoning | S | | | X |
| Modal Intuitionistic, Higher Order Reasoning | S | | X | |
| Model-based Reasoning | H | X | | |
| Non-monotonic Reasoning | S | | | |
| Optimal decisions - Min-Max, Auctions | P | X | | |
| Pattern Matching | H | | | |
| Probabilistic Constraint Satisfaction | H | X | X | X |
| Resource-limited Theorem Proving | S | | X | X |
| SAT - Constraint Satisfaction | S | X | | |
| Special Purpose Reasoning Algorithms | S | | | X |
| Temporal Reasoning | S,P,H | X | | |
| Utility Theory | P | X | | |
| Well-formedness Reasoning | S | | | X |

| Learning Algorithms | Symbolic (S) Probabilistic (P) Hybrid (H) | Ray | LM | ISI |
|---|---|-----|----|-----|
| Abductive Learning | H | | X | |
| Abstraction | H | X | | |
| Analogical Learning | S | | | X |
| Artificial Neural Networks | P | | X | |
| Associative | H | X | | |
| Bayesian Learning | P | | X | |
| Chunking | H | X | | |
| Classification Learning | H | | | |
| Clustering | P,H | X | | |
| Constructing Analogies | S | X | | |
| Co-training | H | | | |
| Data Mining | H | | | |
| Decision Trees | H | | X | |
| Dimensionality Reduction | H | | | |
| Evolutionary Search | H | | | X |
| Genetic Algorithms | P | | X | |
| Inductive Learning | S | | X | |
| Instance-based Learning | P | | X | |
| Learning from Advice | H | X | | |
| Network Construction | P | | | |
| Parameter Learning | P | X | | |
| Plan recognition | H | | | |
| Reinforcement Learning | P,H | X | X | X |
| Relational Learning | S | X | | |
| Rule Generation Composition & Specialization | S | | | |
| Statistical Clustering | P | | X | |
| Statistical Learning (nearest neighbor, approx) | P | | | X |
| Supervised Learning | P | | | X |
| Support Vector Machine | P | X | | |

| Knowledge Representation Algorithms | Symbolic (S) Probabilistic (P) Hybrid (H) | Ray | LM | ISI |
|-------------------------------------|---|-----|----|-----|
| 1st Order Logic (with extensions) | S | X | | X |
| Bayesian Classifier | P | | | |
| Bayesian Networks | P,H | X | | |
| Case-based | S | | | X |
| Causal Networks | H | X | | |
| Conceptual Graphs | H | X | | |
| Decision Trees | H | X | | X |
| Episodic | H | | | X |
| Frames | H | X | | |
| Fuzzy Logic | H | | X | |
| Horn Clause Program | S | | | |
| Influence Diagrams | H | X | | |
| Knowledge Acquisition | H | | | |
| Logical (Prop., FOL, Frame-based) | S | | | |
| Logical Rules | S | | | X |
| Markov Models | P | X | | X |
| Multi-layer Neural Net | P | | | |
| Ontologies | H | X | | X |
| Production System | S | | | |
| Propositional Logic | H | X | | |
| Reactive Plan | S | | | |
| Relational Models | H | X | | |
| Rule-based Systems | H | | | |
| Self-knowledge | H | | | |
| Semantic Nets | S | | | |
| Situation Calculus | S | X | | |
| Taxonomic Hierarchy | P | | | |
| Temporal Networks | H | X | | |
| Type Ontologies and Constraints | S | | | X |

Cognitive_Services

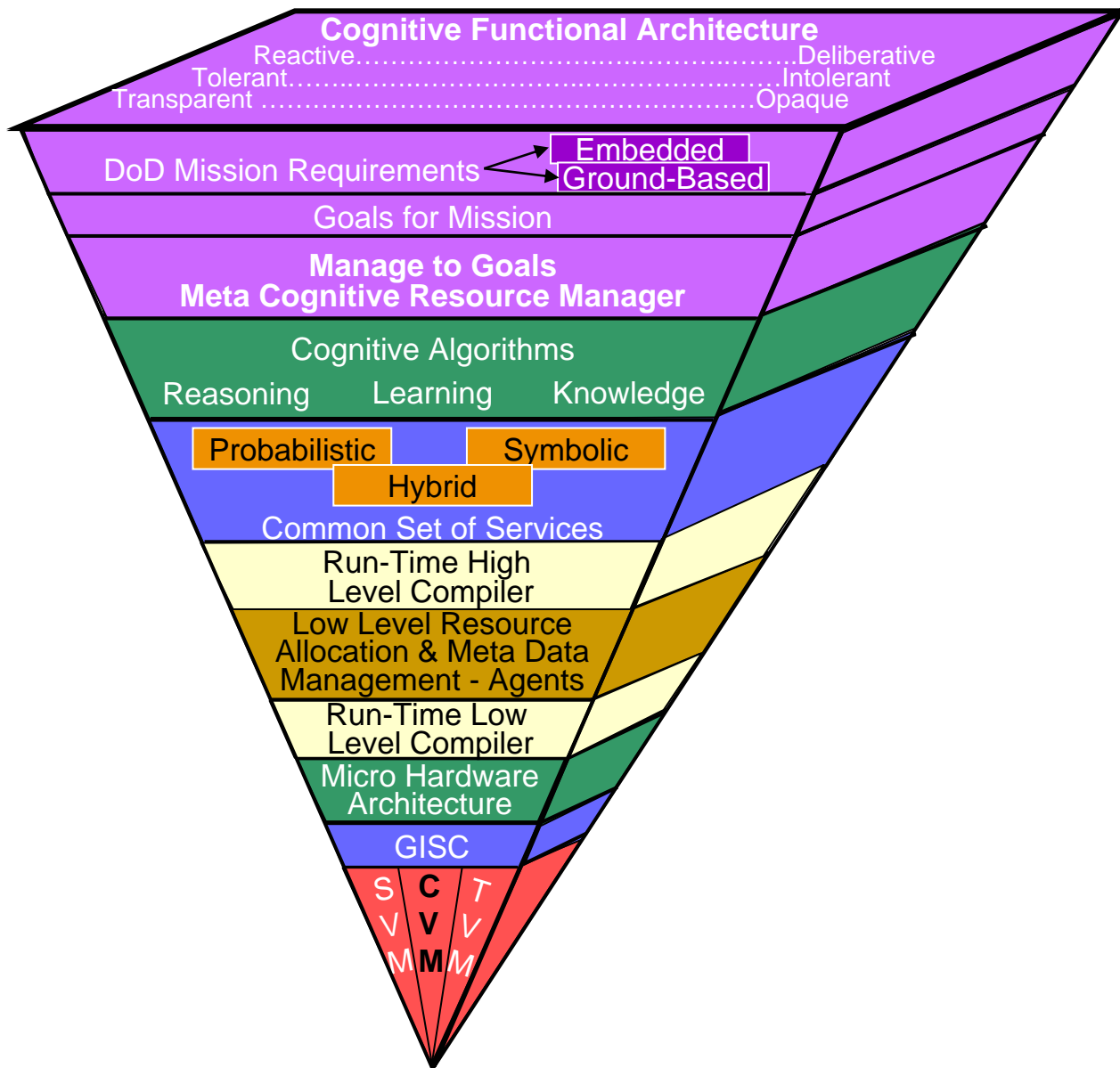


Classical Computing

- ❑ Markovian –current state only
- ❑ Processor-oriented; favors regular addressing
- ❑ Procedural, results oriented – apply this function next
- ❑ Key operations: arithmetic & simple scalar decision making
- ❑ Single deterministic result
- ❑ Parallelism difficult to extract
- ❑ Functional composition determined at compile time
- ❑ Largely static resource management

Cognitive Processing

- ❑ History of prior results guides next: “learning”
- ❑ Memory-oriented; unpredictable access patterns, with metadata guiding access
- ❑ Goal oriented – with multiple, possibly incompatible objectives,
- ❑ Process oriented – history + new perceptions => new knowledge
- ❑ Context oriented – computation based on metadata from prior results
- ❑ Key operations: wide spectrum including complex pattern matching
- ❑ Often multiple “acceptable” results
- ❑ Speculation, futures a first class activity
- ❑ Functional composition determined at run-time
- ❑ Dynamic resource management (Reasoning vs Learning Balance)



PCA (SVM+TVM) + CVM = ACIP???

Potential New Research Ideas!

Leveraging Embedded Computing Workshop Ideas

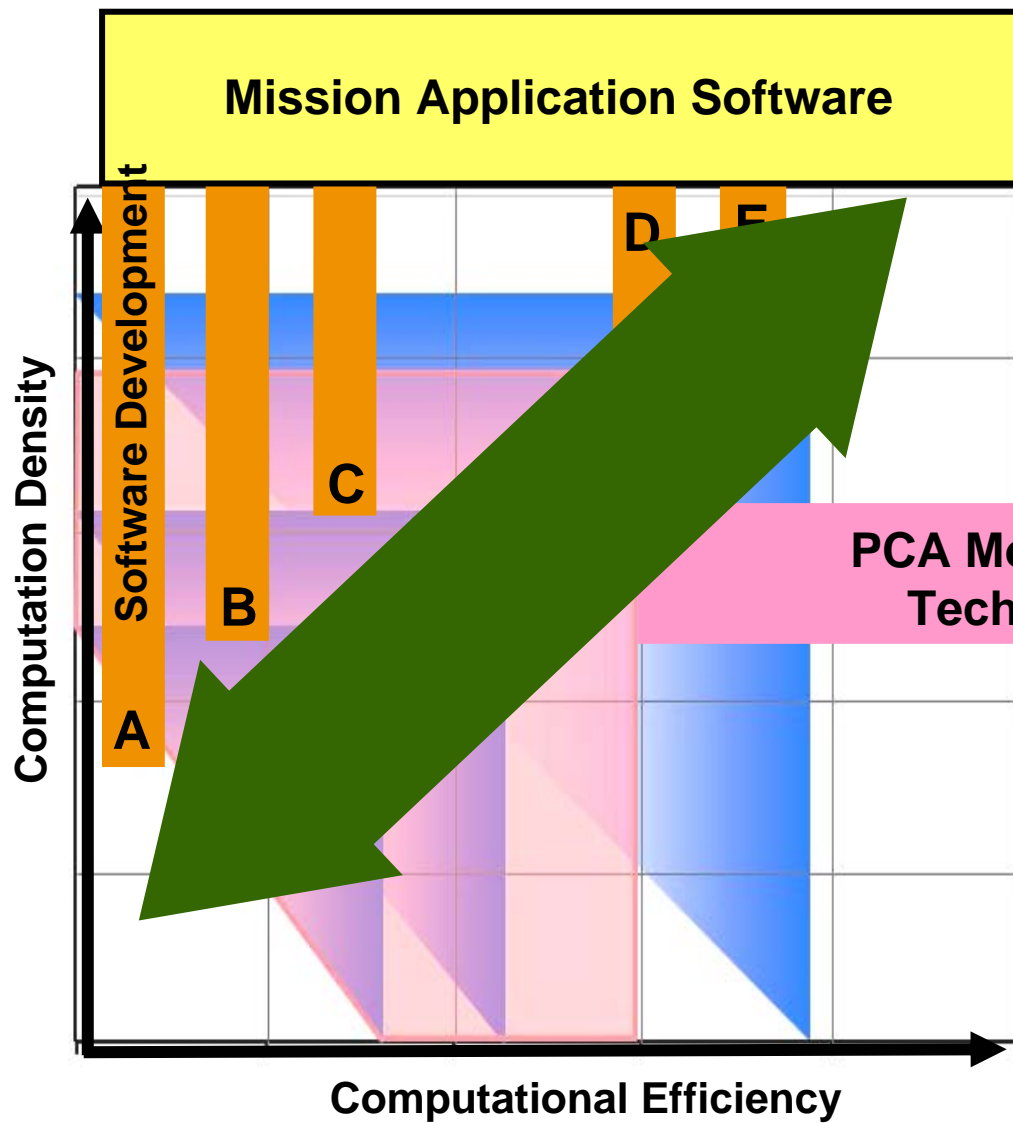
Chaired by

MIT LL and ISI

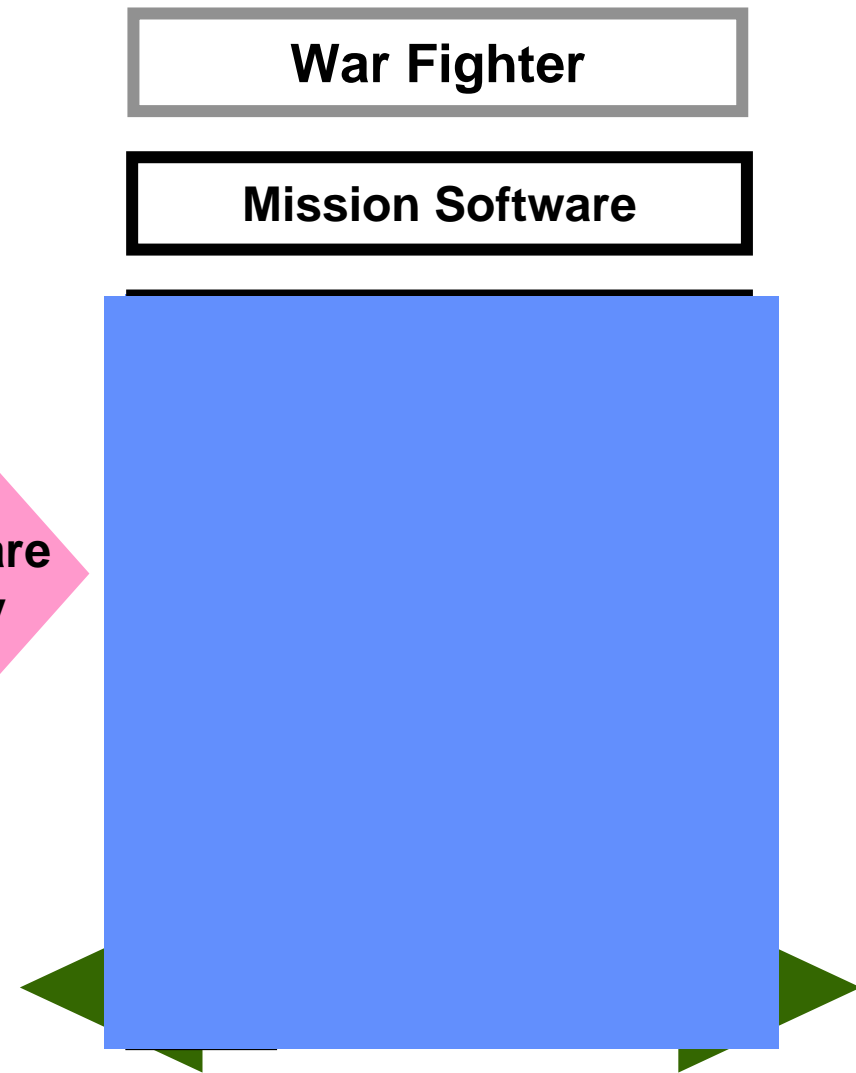
Future Role of Embedded Computing Devices:

GP, DSP, GPU, NIC, FPGA, ASIC

The Problem



The Solution



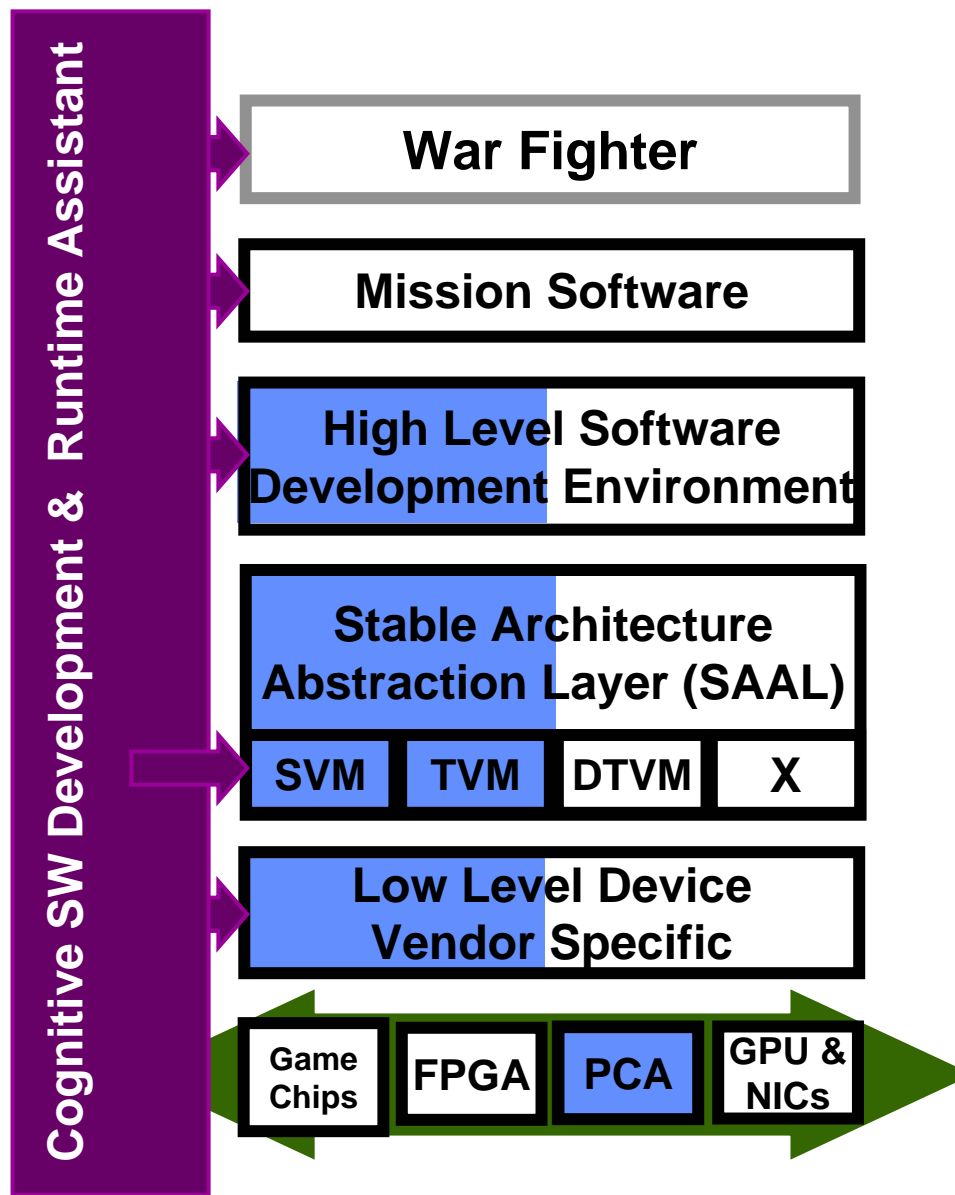
Manual Low Visibility Stove Pipe
SW Development Environments

- Developed under PCA program
- Physical (COTS) PCA Systems Concept

Embedded Computing Complexity Challenge

Embedded Software Developer

**The Solution:
Cognitive Software Developer's Assistant**



 Developed under PCA program

The Future is Yours

**Become an DARPA Program
Manager!!**