

Software Development Environments for  
Scientific and Engineering Software:  
A Series of Case Studies

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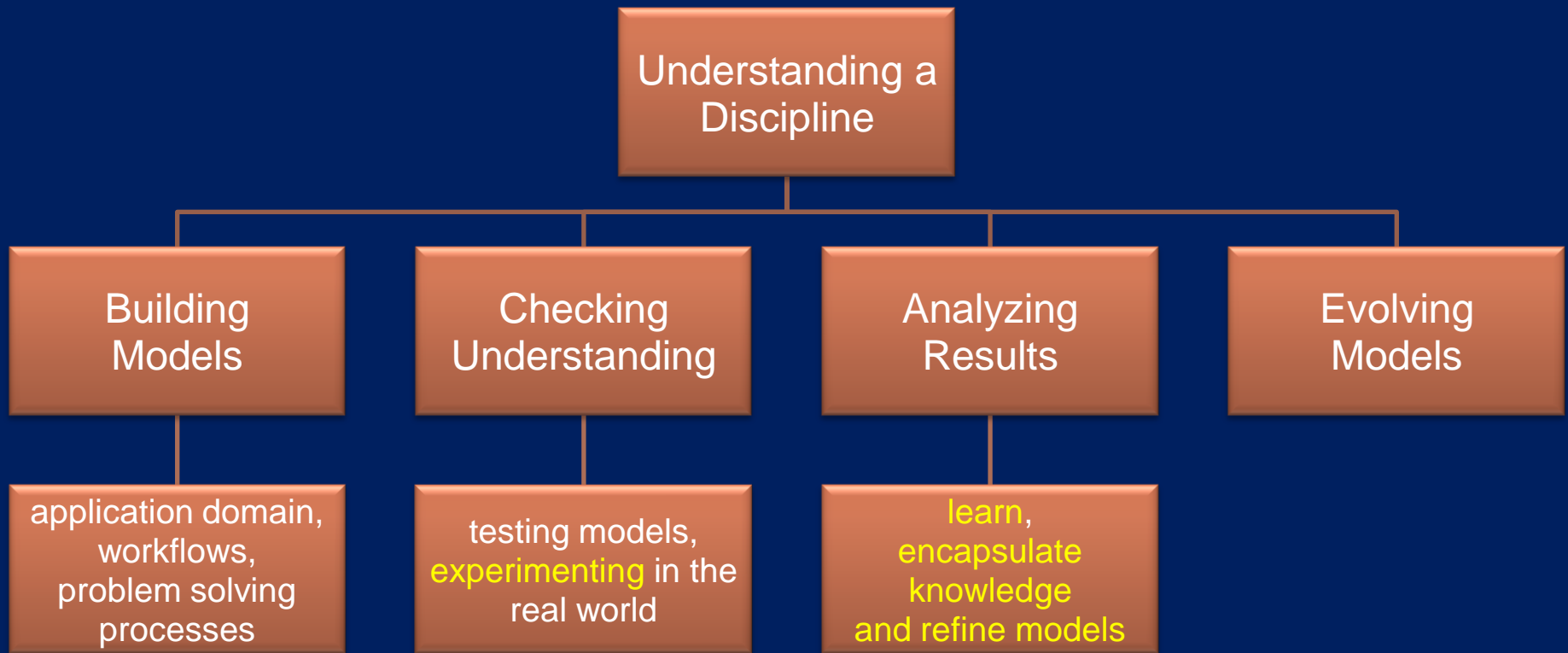
# Outline

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# Introduction

- Software engineering is an engineering discipline
- We need to understand products, processes, and the relationship between them (we assume there is one)
- We need to conduct human-based studies (case studies and experiments)
- We need to package (model) that knowledge for use and evolution
- Recognizing these needs changes how we think, what we do, what is important, and the nature of the discipline

# Empirical Studies

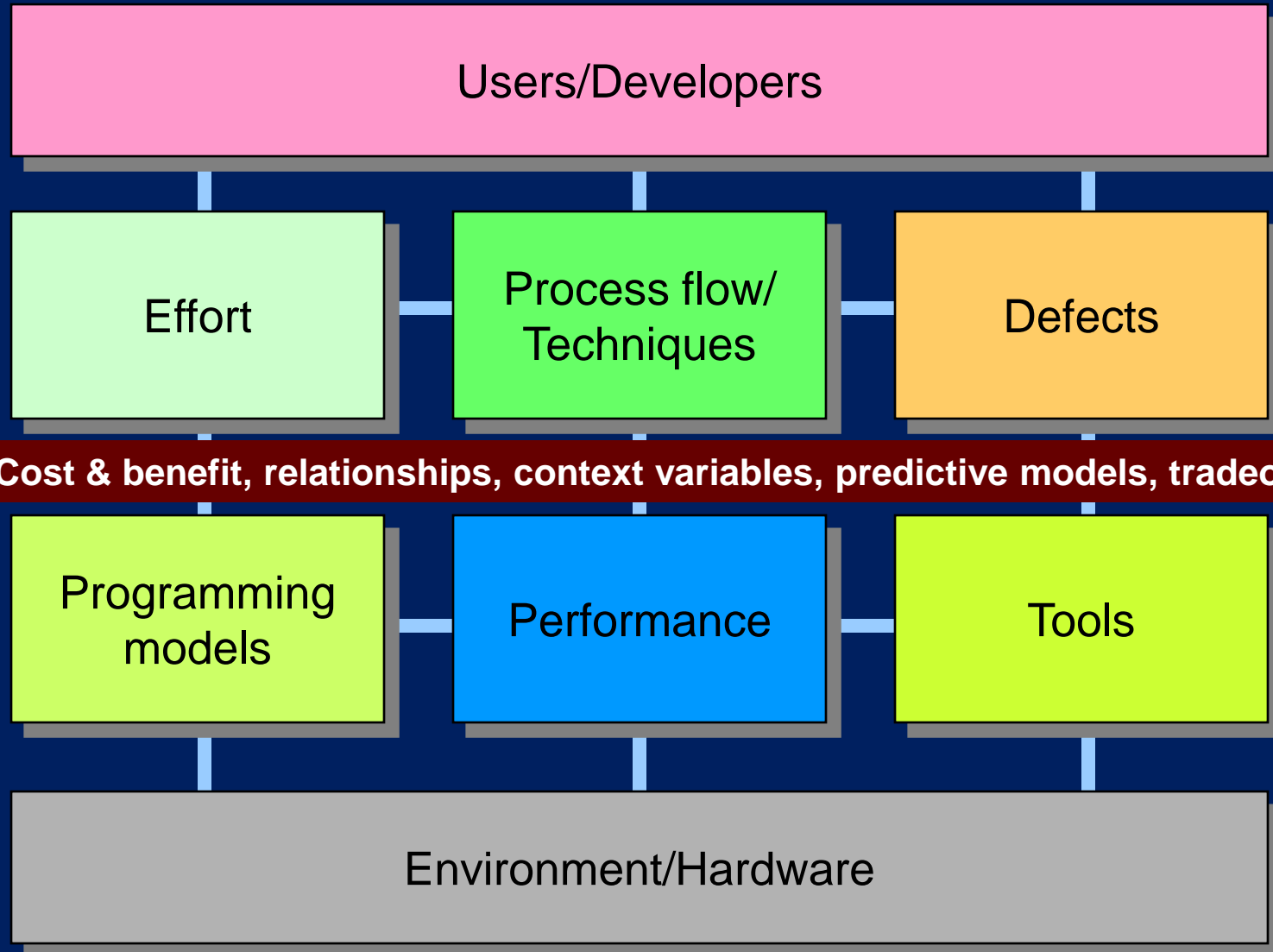


- The empirical paradigm has been used in many other fields, e.g. physics, medicine, manufacturing

# High Productivity Computing Systems (HPCS)

- **Problem:** How do you build sufficient knowledge about high end computing (HEC) so you can improve the time and cost of developing these codes?
- **Project Goal:** Improve the buyer's ability to select the high end computer for the problems to be solved based upon productivity, where productivity means  
$$\text{Time to Solution} = \text{Development Time} + \text{Execution Time}$$
- **Research Goal:** Develop theories, hypotheses, and guidelines that allow us to characterize, evaluate, predict and improve how an HPC environment (hardware, software, human) affects the development of high end computing codes.
- **Partners:** MIT Lincoln Labs, MIT, MSU, UCSD, UCSB, UCSD, UH, UMD, UNL, USC, FC-MD, ISU

# Areas of Study



# Types of HPCS Studies



## Controlled experiments

Study programming in the small under controlled conditions to: Identify key variables, check out methods for data collection, get professors interested in empiricism

E.g., compare effort required to develop code in MPI vs. OpenMP

## Observational studies

Characterize in detail a realistic programming problem in realistic conditions to: validate data collection tools and processes

E.g., build an accurate effort data model

## Case studies and field studies

Study programming in the large under typical conditions

E.g., understand multi-programmer development workflow

## Surveys, interviews & focus groups

Collect “folklore” from practitioners in government, industry and academia

e.g., generate hypotheses to test in experiments and case studies

# Current Study

- **Environment**

- Computational Science and Engineering projects

- **Goals**

- Understand and document software development practices
- Gather initial information about what practices are effective / ineffective

- **Approach**

- Series of retrospective case studies



# Case Study Methodology



# Projects Studied:

## FALCON

**GOAL:** Develop a predictive capability for a product whose performance involves complex physics to reduce the dependence of the sponsor on expensive and dangerous tests.

**DURATION:** ~10 years

**STAFFING:** 15 FTEs

**USERS:** External; highly knowledgeable product engineers



**LANGUAGE:** OO-FORTRAN

**CODE SIZE:** ~405 KSLOC

**TARGET PLATFORM:**

- Shared-memory LINUX cluster (~2000 nodes)
- Vendor-specific shared-memory cluster (~1000 nodes)

# Projects Studied:

## HAWK

**GOAL:** Develop a computationally predictive capability to analyze the manufacturing process allowing the sponsor to minimize the use of time-consuming expensive prototypes for ensuring efficient product fabrication.

**DURATION:** ~ 6 Years

**STAFFING:** 3 FTEs

**USERS:** Internal and external product engineers; small number



**LANGUAGE:** C++ (67%); C (18%); FORTRAN90/Python (15%)

**CODE SIZE:** ~134 KSLOC

**TARGET PLATFORM:**

- SGI (Origin 3900)
- Linux Networx (Evolocity Cluster)
- IBM (P-Series 690 SP)
- Intel-based Windows platforms

# Projects Studied:

## CONDOR

**GOAL:** Develop a simulation to analyze the behavior of a family of materials under extreme stress allowing the sponsor to minimize the use of time-consuming expensive and infeasible testing.

**DURATION:** ~ 20 Years

**STAFFING:** 3-5 FTEs

**USERS:** Internal and external; several thousand occasional users; hundreds of routine users



**LANGUAGE:** FORTRAN77  
(85%)

**CODE SIZE:** ~200 KSLOC

**TARGET PLATFORM:**

- PC – running 106 cells for a few hours to a few days (average)
- Parallel application – 108 cells on 100 to a few 100s of processors

# Projects Studied:

## EAGLE

**GOAL:** Determine if parallel, real-time processing of sensor data is feasible on a specific piece of HPC hardware deployed in the field

**DURATION:** ~ 3 Years

**STAFFING:** 3 FTEs



**LANGUAGE:** C++

**CODE SIZE:** < 100 KSLOC

**USERS:** Demonstration project – no users

**TARGET PLATFORM:**

- Specialized computer that can be deployed on military platforms
- Developed on – SUN Sparcs (Solaris) and PC (Linux)

# Projects Studied:

## NENE

**GOAL:** Calculate the properties of molecules using a variety of computational quantum mechanical models

**DURATION:** ~25 Years

**STAFFING:** ~10 FTEs  
(Thousands of contributors)

**USERS:** 200,000 installations and estimated 100,000 users



**LANGUAGE:** FORTRAN77  
subset of FORTRAN90

**CODE SIZE:** 750 KSLOC

**TARGET PLATFORM:**  
All commonly used platforms  
except Windows-based PCs

# Projects Studied:

## OSPREY

**GOAL:** One component of a large weather forecasting suite that combines the interactions of large-scale atmospheric models with large-scale oceanographic models.

**DURATION:** ~10 years  
(predecessor > 25 years)

**STAFFING:** ~10 FTEs

**USERS:** Hundreds of installations – some have hundreds of users



**LANGUAGE:** FORTRAN

**CODE SIZE:** 150 KLOC  
(50 KLOC Comments)

**TARGET PLATFORM:** SGI,  
IBM, HP, and Linux

# Projects Studied: Summary

	FALCON	HAWK	CONDOR	EAGLE	NENE	OSPREY
<b>Application Domain</b>	Product Performance	Manufacturing	Product Performance	Signal Processing	Process Modeling	Weather Forecasting
<b>Duration</b>	~ 10 years	~ 6 years	~ 20 years	~ 3 years	~ 25 years	~10 years
<b># of Releases</b>	9 (production)	1	7	1	?	?
<b>Staffing</b>	15 FTEs	3 FTEs	3-5 FTEs	3 FTEs	~10 FTEs (100's of contributors)	~10 FTEs
<b>Customers</b>	< 50	10s	100s	None	~ 100,000	100s
<b>Code Size</b>	~ 405,000 LOC	~ 134,000 LOC	~200,000 LOC	< 100,000 LOC	750,000 LOC	150,000 LOC
<b>Primary Languages</b>	F77 (24%), C (12%)	C++ (67%), C (18%)	F77 (85%)	C++, Matlab	F77 (95%)	Fortran
<b>Other Languages</b>	F90, Python, Perl, ksh/csh/sh	Python, F90	F90, C, Slang	Java Libraries	C	C
<b>Target Hardware</b>	Parallel Supercomputer	Parallel Supercomputer	PCs to Parallel Supercomputer	Embedded Hardware	PCs to Parallel Supercomputer	Parallel Supercomputer



# Lessons Learned

# Lessons Learned: Validation and Verification

## Validation

- Does the software correctly capture the laws of nature?
- Hard to establish the correct output of simulations *a priori*
  - Exploring new science
  - Inability to perform experimental replications

## Verification

- Does the application accurately solve the equations of the solution algorithm?
- Difficult to identify problem source
  - Creation of mathematical model by domain expert
  - Translation of mathematical model into algorithm(s)
  - Implementation of algorithms in software

# Lessons Learned: Validation and Verification

*I have tried to position CONDOR to the place where it is kind of like your trusty calculator – it is an easy tool to use. Unlike your calculator, it is only 90% accurate ... you have to understand that then answer you are going to get is going to have a certain level of uncertainty in it. The neat thing about it is that it is easy to get an answer in the general sense <to a very difficult problem>.*

## ■ Implications

- Traditional methods of testing software then comparing the output to expected results are not sufficient
- These developers need additional methods to ensure quality and limits of software

# Lessons Learned: Language Stability

- Long project lifecycles require code that is:
  - Portable
  - Maintainable
- FORTRAN
  - Easier for scientists to learn than C++
  - Produces code that performs well on large-scale supercomputers
- Users of the code interact frequently with the code
- **Implications**
  - FORTRAN will dominate for the near future
  - New languages have to have benefits of FORTRAN plus some additional benefits to be accepted

# Lessons Learned:

## Use of Higher-Level Languages

*I'd rather be closer to machine language than more abstract. I know even when I give very simple instructions to the compiler, it doesn't necessarily give me machine code that corresponds to that set of instructions. If this happens with a simple do-loop in FORTRAN, what happens with a monster object-oriented thing?*

### •MATLAB

- Code is not efficient or fast enough
- Used for prototyping

### •C++

- Used by some newer teams
- Mostly used the C subset of C++

## ■ Implications

- These developers place more constraints on the language than traditional IT developers
- A language has to
  - Be easy to learn
  - Offer reasonably high performance
  - Exhibit stability
  - Give developers confidence in output of compiler

# Lessons Learned:

## Development Environments

*They all [the IDEs] try to impose a particular style of development on me and I am forced into a particular mode*

- Developers prefer flexibility of the command line over an Integrated Development Environment (IDE). They believe that:
  - IDEs impose too much rigidity
  - They are more efficient when typing commands than when navigating menus
- **Implications** – developers do not adopt IDEs because:
  - They do not trust the IDE to automatically perform a task in the same way they would do it manually
  - They expect greater flexibility than is currently provided
  - Prefer to use what they know rather than change

# Lessons Learned: External Software

- Projects view external software as a risk
  - Long duration
  - Fear that software may disappear or become unsupported
  - Prefer to develop tools in-house or use open-source
- Exception – NENE
  - Employed a librarian to thoroughly test code before integrating into code base
  - Designed the project so that it was not dependent on external software to meet its commitments
- **Implication - Tool problem**
  - Very few quality tools for this environment
  - Catch-22 situation

# Lessons Learned: Development Goals

- Multiple goals are important
  - **Performance** – software is used on supercomputer
  - **Portability** and **Maintainability** – platforms change multiple times during a project
- Success of a project depends on the ability to port software to new machines
- **Implications**
  - The motivation for these projects may be different than for traditional IT projects
  - Methods must be chosen and tailored to align with the overall project goals



# Lessons Learned:

## Agile vs. Traditional Methodologies

- “Agile” refers to the philosophical approach rather than to any particular Agile method
- Projects are often doing new science, so the requirements cannot be known upfront
- Teams have been operating with an agile philosophy before they even knew what it was – favoring individuals and good practices over rigid processes and tools
- **Implications**
  - Existing SE methodologies need to be tailored for this community
  - Rigid, process-heavy approaches are not used; both for technical and cultural reasons

# Lessons Learned: Team Composition

*In these types of high performance, scalable computing [applications], in addition to the physics and mathematics, computer science plays a very major role. Especially when looking at optimization, memory management and making [the code] perform better ... You need a multi-disciplinary team. It [C++] is not a trivial language to deal with ... You need an equal mixture of subject theory, the actual physics, and technology expertise.*

- Complex problems and domains
  - Too difficult for most software engineers to understand quickly
  - Easier to teach domain scientists/engineers how to program
- Software engineers help with performance and flexibility
- **Implication**
  - Multi-disciplinary teams are important

# Lessons Learned: Key to Success

- Keeping customers (and sponsors) satisfied
- Lesson not unique to this community, but some constraints are important
  - Funding may come from one agency, while customers are members of another agency
  - Success depends on keeping both groups happy
  - HAWK project was suspended due to lack of customer support, even though it was a technical success for the funding agency
- **Implication**
  - Balancing the needs of these various stakeholders can be challenging

# Summary

- Six case studies of computational science and engineering software
  - Projects sponsored by the US Federal Government and the National Science Foundation
  - Different domains and different goals
- Nine lessons learned about the programming environment drawn across all studies
- Contributions
  - For the software engineering community
    - Highlighted some reasons why the development process is different for this type of software
    - Provided insight into why traditional SE approaches are not used
  - For the computational science and engineering community
    - Provided ideas to guide the improvement of the software engineering process

# Collaborators

- Doug Post - HPCMP
- Richard Kendall – SEI
- Dale Henderson – Los Alamos (retired)
- Andrew Mark – HPCMP
- David Fisher – HPCMP
- Clifford Rhoades, Jr. – Maui HPC Center
- Susan Squires – Tactics (formally with SUN)
- Christine Halverson - IBM

# For More Information

- *IEEE Software* special issue – Developing Scientific Software (July/August 2008)
- Carver, et al. “Software Development Environments for Scientific and Engineering Software: A Series of Case Studies.” *ICSE 2007*
- ICSE workshops
  - Software Engineering for HPC Applications (2004-2005, 2007)
  - Software Engineering for Computational Science (2008)
    - Forthcoming news article in CiSE (March/April)
  - **2009 workshop proposed**

# Selected References

- Carver, J., Kendall, R., Squires, S. and Post, D. "Software Development Environments for Scientific and Engineering Software: A Series of Case Studies." *Proceedings of the 29th International Conference on Software Engineering*. Minneapolis, USA. May 23-25, 2007. p. 550-559.
- Kendall, R., Carver, J., Fisher, D., Henderson, D., Mark, A., Post, D., Rhoades, C. and Squires, S. "Development of a Weather Forecasting Code: A Case Study." *IEEE Software*, July/August 2008. p. 59-65.
- Kendall, R., Post, D., Carver, J., and Squires, S. "Case Study of the Eagle Code Project." Los Alamos Technical Report, LA-UR-06-1092. 2006.
- Kendall, R., Carver, J., Mark, A., Post, D., Squires, S., and Shaffer, D. "Case Study of the Hawk Code Project." Los Alamos Technical Report, LA-UR-05-9011. 2005.
- Kendall, R.P., Mark, A., Post, D., Squires, S., and Halverson, C. Case Study of the Condor Code Project. Technical Report, LA-UR-05-9291. Los Alamos National Laboratories: 2005.
- Post, D.E., Kendall, R.P., and Whitney, E. "Case study of the Falcon Project". *Proceedings of Second International Workshop on Software Engineering for High Performance Computing Systems Applications (Held at ICSE 2005)*. St. Louis, USA. 2005. p. 22-26

# Thank You!

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