

“Telling More“: a 2006 NASA SARP project (Technical Briefing)

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Introduction: Use it or Lose it (and this is the last “business-level” slide)

Introduction

❖ Use it or Lose it

❖ This talk...

❖ Disclaimer etc

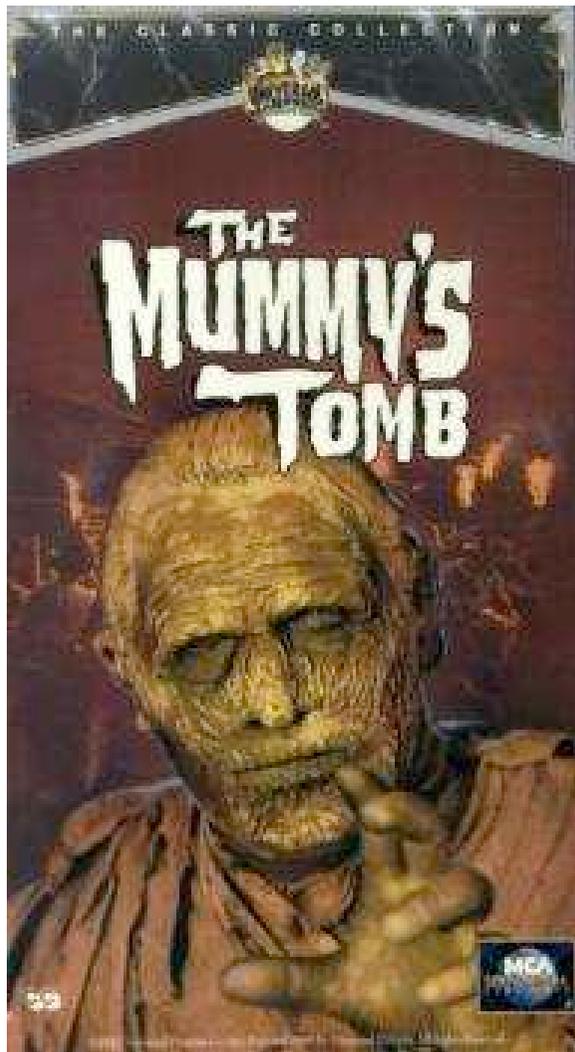
Knowledge Farming

SILAP

mb1

What's next?

Conclusion



- NASA data = active repository or data tomb?
 - ◆ write once;
 - ◆ read never;
 - ◆ buried;
 - ◆ doomed;
- If an organization spends millions of dollars on data collection and archiving...
 - ◆ It should spend tens of thousands (at least) in analyzing that data.
- NASA IV&V is in a unique position to review and comment on much of the NASA software enterprise.
 - ◆ We see more, but what have we *learned*?
 - ◆ What can we *tell*?

This talk...

Introduction

❖ Use it or Lose it

❖ This talk...

❖ Disclaimer etc

Knowledge Farming

SILAP

mb1

What's next?

Conclusion

- “Knowledge farming” on “SILAP” models
 - ◆ “Knowledge farming”: see below
 - ◆ “SILAP”: how IV&V selects which tasks to perform for a given project;
- Results:
 - ◆ SILAP v.1: an open source version of SILAP
<http://unbox.org/wisp/trunk/silap>;
 - ◆ MB1:
 - 433 SLIAPed software elements from NASA
 - divided into different project types.
 - <http://unbox.org/wisp/trunk/silap/data/mb1.csv>
 - ◆ Experiments showing:
 - Selected IV&V tasks often the *same*, despite processing *different* projects;
 - Identification of the *really different* project types that lead to different IV&V tasks

Disclaimer etc

Introduction

❖ Use it or Lose it

❖ This talk...

❖ Disclaimer etc

Knowledge Farming

SILAP

mb1

What's next?

Conclusion

● Disclaimer

- ◆ Reference herein to any specific commercial product, process, or service by trade name, trademark, manufacturer, or otherwise, does not constitute or imply its endorsement by the United States Government.
- ◆ While the planning&scoping team at Fairmont is currently reviewing this material....
 - the views expressed here-in are the author.
 - They do NOT reflect official NASA policy
 - They do NOT reflect the views of NASA civil servants.

● Acknowledgments

- ◆ The research described in this talk was carried out at West Virginia University under a contract with the National Aeronautics and Space Administration.
- ◆ This work would not have been possible without the advise, assistance, and access to data offered by the NASA software Independent Verification & Validation planning & scoping team (at Fairmont WV).
- ◆ Special thanks to (in alpha order): Markland Benson, Ken Costello, Ken McGill, Christina Moats, Melissa Northey.

Introduction

Knowledge Farming

- ❖ Types
- ❖ Why not KM?
- ❖ Examples
- ❖ More examples
- ❖ Yet more examples

SILAP

mb1

What's next?

Conclusion

Knowledge Farming

Data/Knowledge Farming

Introduction

Knowledge Farming

❖ Types

❖ Why not KM?

❖ Examples

❖ More examples

❖ Yet more examples

SILAP

mb1

What's next?

Conclusion

- Andrew Kusiak : J. Ops. research, 2005: “data farming”
 - ◆ DF generalizes DM (data mining)
 - ◆ DF = models + practices used to define most appropriate features for data collection/ transformation/ assessment
 - ◆ DF.effort greater than data mining effort
- DOD : “data farming”
 - ◆ Using a super-computer...
 - ◆ ... simulate the h#ck out a model
 - ◆ e.g. the Marine Corps’ *Project Albert*
 - ◆ understand the landscape of potential simulated outcomes, enhance intuition, find surprises and outliers, and identify potential options
 - ◆ Heavily augmented with interactive visualization tools
 - ◆ ??? no automatic summarization methods (e.g. data miners)
- Me, ASE 2000: “Knowledge farming”
 - ◆ **Plant** the seeds; i.e. build a simulator that can reproduce domain conditions;
 - ◆ **Grow** the seedlings; i.e. Monte Carlo the simulator to generate data;
 - ◆ **Harvest** the crop; i.e. use data miners to find patterns in the data;
 - ◆ Core problem: how to build human-readable succinct rules from gigabytes of data

The Case Against Knowledge Farming (KM)

Introduction

Knowledge Farming

❖ Types

❖ Why not KM?

❖ Examples

❖ More examples

❖ Yet more examples

SILAP

mb1

What's next?

Conclusion

- Q: Won't you just re-learn the original model?
 - ◆ A: Nope
 - ◆ Learned model = input data + summarization method;
 - ◆ Different summarization, different models;
 - ◆ KM + *relevancy tests* often prunes away variables that are noisy, low variance, under-sampled, not informative, etc etc
 - ◆ KM's harvest smaller than original.
 - ◆ Effects that are obscure in original are clear in summary
- Q: Aren't you just learning quirks in the model?
 - ◆ A₁: If you have domain data, don't rely on a (possibly incorrect) model
 - ◆ A₂: What's the difference?
 - If the model is being used to make policy decisions, then we need to know the model, warts and all.
 - ◆ A₃: if your model has subtle quirks, how will you find them otherwise? (e.g. KARDIO)

Examples of knowledge farming (KARDIO: Bratko'89; ESA: Pearce'88)

Introduction

Knowledge Farming

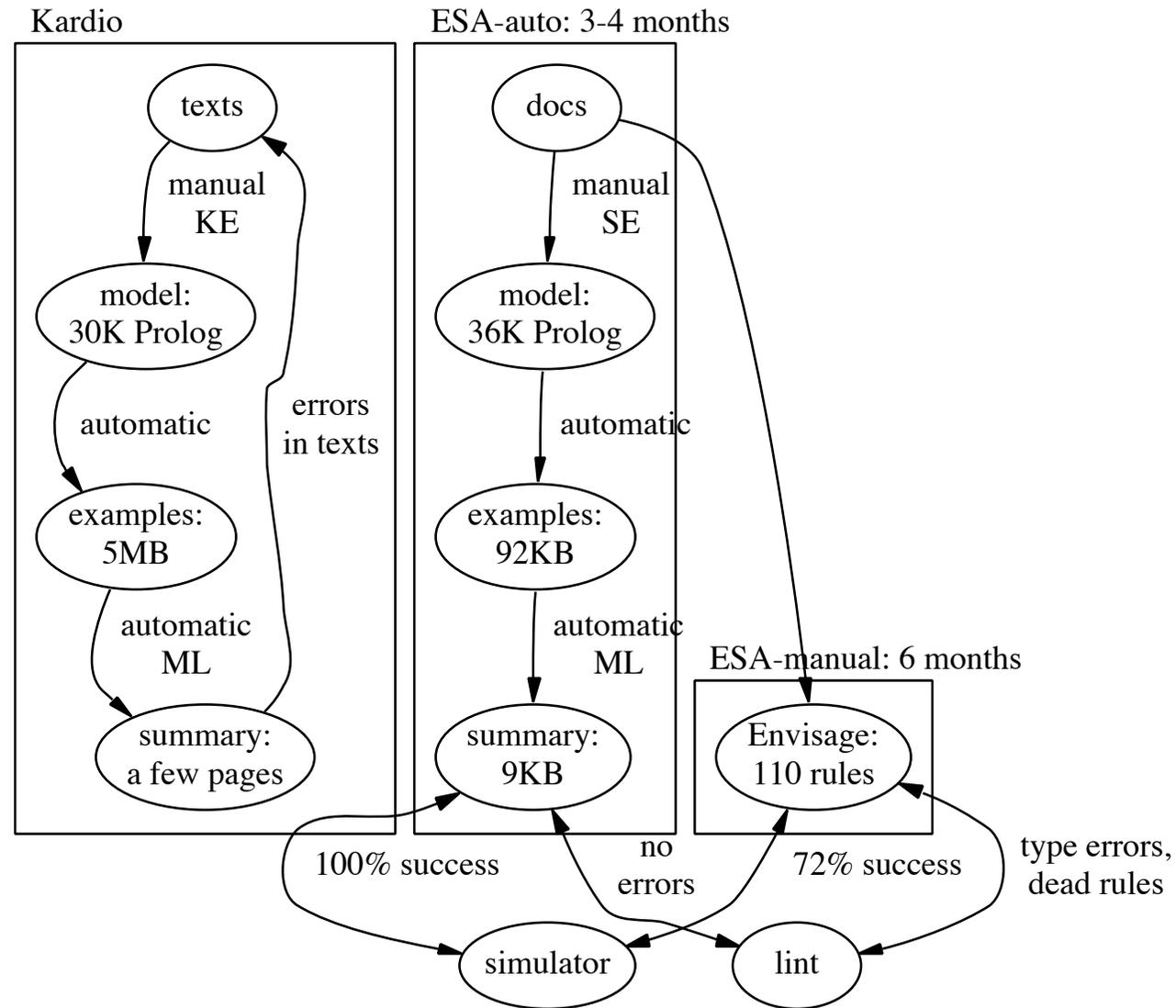
- ❖ Types
- ❖ Why not KM?
- ❖ **Examples**
- ❖ More examples
- ❖ Yet more examples

SILAP

mb1

What's next?

Conclusion



(More) Examples of knowledge farming (Menzies&Raffo: ASE'02; Menzies&Feather: RE'02)

Introduction

Knowledge Farming

- ❖ Types
- ❖ Why not KM?
- ❖ Examples
- ❖ **More examples**
- ❖ Yet more examples

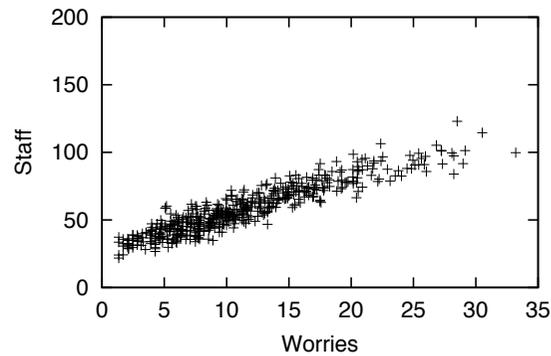
SILAP

mb1

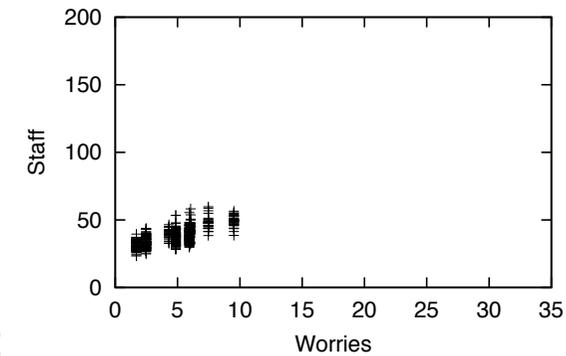
What's next?

Conclusion

Optimizing a combined staffing/“worries” model: staffing: the COCOMO-II effort model; “worries”: the Madachy software risk model.

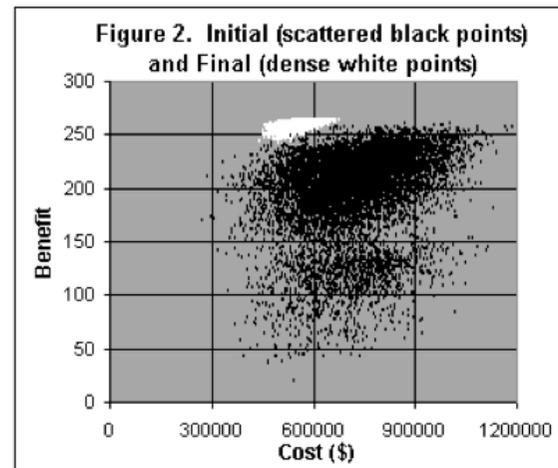


before:



after:

Trade space studies at JPL: benefits: requirements coverage; costs: cost of risk mitigation strategies (each dot is yes-no to 99 decisions).



(Yet more) Examples

Introduction

Knowledge Farming

- ❖ Types
- ❖ Why not KM?
- ❖ Examples
- ❖ More examples
- ❖ **Yet more examples**

SILAP

mb1

What's next?

Conclusion

This talk: knowledge farming and SILAP

Introduction

Knowledge Farming

SILAP

❖ Context

❖ SILAP

❖ SILAP structure

❖ WBS

mb1

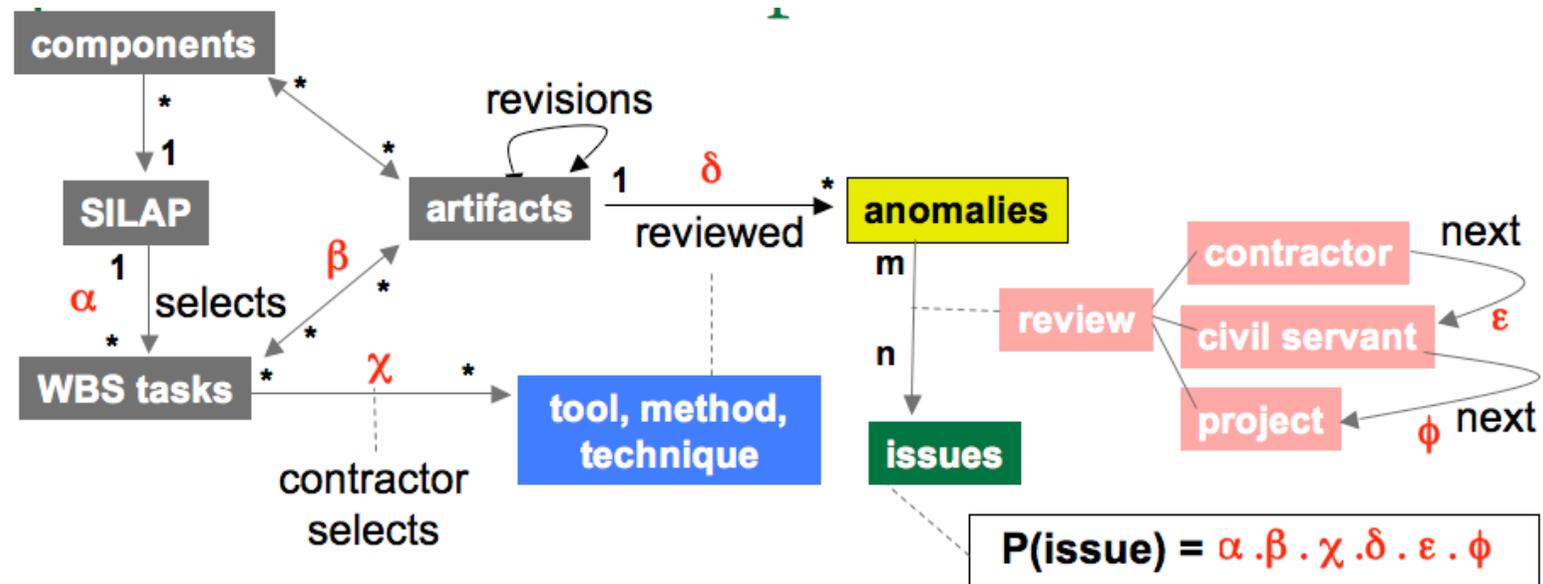
What's next?

Conclusion

SILAP

The context of SILAP

- Components have artifacts
- SILAP selects WBS tasks
- Tasks are associated with artifacts;
- Artifacts generate anomalies;
- Filters reject bogus anomalies;
- Projects accept issues



Introduction

Knowledge Farming

SILAP

❖ Context

❖ SILAP

❖ SILAP structure

❖ WBS

mb1

What's next?

Conclusion

SILAP: why yet another risk model?

- Introduction
- Knowledge Farming
- SILAP
 - Context
 - SILAP**
 - SILAP structure
 - WBS
- mb1
- What's next?
- Conclusion

- Software Integrity Level Assessment Process
- 16 criteria (scored 1,2,3,4,5)
 - ◆ Criteria used to calculate *errorpotential* (risk) and *consequence* (severity).
 - ◆ Which, in turn, select IV&V tasks from the IV&V work breakdown structure adapted from IEEE Std 1012 (V&V standard)
- In practice, takes two weeks (or more) full-time work to do a SILAP assessment.
 - ◆ Review the documents
 - ◆ Offer a score for the criteria
 - ◆ Write a detailed rationale for the score
- Understood locally (very important):
 - ◆ Previous risk models were developed elsewhere and, to a degree were “black box”; i.e. inexplicable, not defensible
 - ◆ SILAP, on the other hand, was built locally at Fairmont IV&V after *extensive* meetings between NASA civil servants.
- Civil servants report that SILAP has simplified and clarified their discussions with projects regarding what IV&V tasks are/are not to be performed.

SILAP makes IV&V business knowledge explicit, publicly accessible.

- This record lets outsiders (like me) review their work practices.
- How many other organizations can say they have done the same?

SILAP structure

Introduction

Knowledge Farming

SILAP

❖ Context

❖ SILAP

❖ **SILAP structure**

❖ WBS

mb1

What's next?

Conclusion

variable = meaning

AM3 = Artifact Maturity

AS2 = Asset Safety

CL3 = CMM Level

CO1 = Consequence

CX3 = Complexity

DI3 = Degree of Innovation

DO3 = Development Organization

DT3 = Use of Defect Tracking System

DV2 = Development

EP1 = Error Potential

EX3 = Experience

variable =meaning

FR3 =Use of Formal Reviews

HS2 =Human Safety

PF2 =Performance

PR2 =Process

RA3 =Re-use Approach

RM3 =Use of Risk Management System

SC2 =Software Characteristic

SS3 =Size of System

UC3 =Use of CM

US3 =Use of Standards

```
function CO1( tmp) { // Consequence
    tmp=0.35*value("AS2") + 0.65 *value("PF2"); return round((value("HS2") > tmp) ? value("HS2") : tmp; ) }
```

```
function EP1() { // Error Potential
    return round(0.579*DV2() + 0.249*PR2() + 0.172*SC2() ) }
```

```
function SC2() { // Software Characteristic
    return 0.547*value("CX3") + 0.351*value("DI3") + 0.102*value("SS3") }
```

```
function DV2() { // Development
    return 0.828*value("EX3") + 0.172*value("DO3") }
```

```
function PR2() { // Performance
    return 0.226*value("RA3") + 0.242*value("AM3") + formality() }
```

```
function formality() {
    return 0.0955*value("US3") + 0.0962*value("UC3") + 0.0764*value("CL3")
    +0.1119*value("FR3") + 0.0873*value("DT3") + 0.0647*value("RM3") }
```

SILAP selects tasks from the IV&V work breakdown structure

Introduction

Knowledge Farming

SILAP

❖ Context

❖ SILAP

❖ SILAP structure

❖ WBS

mb1

What's next?

Conclusion

wbs	factor	CO11	CO12	CO13	CO14	CO15	EP11	EP12	EP13	EP14	EP15
1.1	Management and Planning of IV&V	X	X	X	X	X	X	X	X	X	X
1.2	Issue and Risk Tracking		X	X	X	X		X	X	X	X
1.3	Final Report Generation		X	X	X	X		X	X	X	X
1.4	IV&V Tool Support		X	X	X	X		X	X	X	X
1.5	Management & Technical Review Support	X	X	X	X	X	X	X	X	X	X
1.6	Criticality Analysis	X	X	X	X	X	X	X	X	X	X

2.1	Reuse Analysis*			X	X	X					
2.2	Software Architecture Assessment			X	X	X					
2.3	System Requirements Review			X	X	X			X	X	
2.4	Concept Document Evaluation				Z	Z					Z
2.5	SW/User Requirements Allocation Analysis				Z	Z					Z
2.6	Traceability Analysis				Z	Z					Z

3.1	Traceability Analysis - Requirements		X	X	X	X				X	X
3.2	Software Requirements Evaluation			X	X	X				X	X
3.3	Interface Analysis - Requirements				X	X		X	X	X	
3.4	System Test Plan Analysis			X	X	X					
3.5	Acceptance Test Plan Analysis					X					
3.6	Timing and Sizing Analysis									Z	Z

{CO1, EP1} computed via SILAP
 CO1 = "consequences of failure"
 EP1 = "error potential"

X = "or";
 * = for $RA3 > 1$;
 Z = for human-rated flights

Work breakdown structure (cont.)

Introduction

Knowledge Farming

SILAP

❖ Context

❖ SILAP

❖ SILAP structure

❖ WBS

mb1

What's next?

Conclusion

wbs	factor	CO11	CO12	CO13	CO14	CO15	EP11	EP12	EP13	EP14	EP15
4.1	Traceability Analysis - Design							X	X	X	X
4.2	Software Design Evaluation				X	X				X	X
4.3	Interface Analysis - Design					X				X	X
4.4	Software FQT Plan Analysis		X	X	X	X					
4.5	Software Integration Test Plan Analysis									X	X
4.6	Database Analysis							X		X	X
4.7	Component Test Plan Analysis										X

5.1	Traceability Analysis - Code					X		X	X	X	X
5.2	Source Code and Documentation Evaluation				X	X			X	X	X
5.3	Interface Analysis - Code				X	X		X	X	X	
5.4	System Test Case Analysis				X	X					
5.5	Software FQT Case Analysis				X	X					
5.6	SW Integration Test Case Analysis										X
5.7	Acceptance Test Case Analysis					X					
5.8	SW Integration Test Procedure Analysis										X
5.9	SW Integration Test Results Analysis								X	X	
5.1	Component Test Case Analysis										X
5.11	System Test Procedure Analysis					Z					
5.12	Software FQT Procedure Analysis					Z					

6.1	Traceability Analysis - Test		X	X	X	X					X
6.2	Regression Test Analysis									Z	Z
6.3	Simulation Analysis					Z					
6.4	System Test Results Analysis				X	X					
6.5	Software FQT Results Analysis				X	X					

7.1	Operating Procedure Evaluation					Z					
7.2	Anomaly Evaluation					Z					
7.3	Migration Assessment					Z					
7.4	Retirement Assessment					Z					

Introduction

Knowledge Farming

SILAP

mb1

- ❖ Components
- ❖ separation
- ❖ “All” and “Orbits”
- ❖ “Orbits” (cont.)
- ❖ “Prime”
- ❖ “Profile”
- ❖ “Profile” (cont.)
- ❖ Selections
- ❖ Separation
- ❖ Selected WBS
- ❖ Selected WBS (more)
- ❖ Weighted Frequencies

What's next?

Conclusion

MB1: 433 SILAPed software elements

MB1: Components

Introduction

Knowledge Farming

SILAP

mb1

❖ Components

❖ separation

❖ “All” and “Orbits”

❖ “Orbits” (cont.)

❖ “Prime”

❖ “Profile”

❖ “Profile” (cont.)

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❖ Selected WBS (more)

❖ Weighted Frequencies

What's next?

Conclusion

433 SILAP-ed software elements from NASA (system, subsystem, CSCI, or CSC level); sanitized data, i.e. mission/center names replaced with x_1, x_2, \dots . Divided into:

- All
- “Orbits”
 - ◆ gs1: ground system
 - ◆ es2: earth orbit
 - ◆ xf3: “transfer” (a mission that does not have an Earth-centric orbit),
 - ◆ go4: ground ops on different planet
- “profile”: type of science
 - ◆ {es,hs,op,ss,su}
- “prime”: who built the sub-system (which NASA center)
 - ◆ {p1,p2,p3,p4}
- Auto: learned by a clustering algorithm (EM) that ignores all the above distinctions
 - ◆ {cluster0, cluster1,cluster2,cluster3,cluster4}

Q: for the purposes of IV&V, which of the above matter at all?

A: apply the *principle of separation*

Principle of separation

Introduction

Knowledge Farming

SILAP

mb1

❖ Components

❖ separation

❖ “All” and “Orbits”

❖ “Orbits” (cont.)

❖ “Prime”

❖ “Profile”

❖ “Profile” (cont.)

❖ Selections

❖ Separation

❖ Selected WBS

❖ Selected WBS (more)

❖ Weighted Frequencies

What's next?

Conclusion

- Different things get effected different ways.
 - ◆ $Out = f(In)$
 - ◆ *Weak separation*: All *In*s don't have the same *Out*s
 - ◆ *Strong separation*: Different *In*s have different *Out*s
- A model that does not *separate* is *blunt*; i.e. is not sensitive to changes in the inputs.
- I.e. the SILAP divisions that matter are those that:
 - ◆ Are most different to the *All* set;
 - “Profile” selects for nearly the same tasks as “All”
 - ◆ Most select different WBS tasks. different tasks.
 - All the manual divisions (“orbits”, “prime”, “profile”) select for very similar tasks.
 - Only the automatic divisions separate from each other and from “All”
- Test for separation
 - ◆ For each division, summarize the distributions;
 - ◆ Run each distribution through SILAP; record selected WBS tasks;
 - ◆ Compare the selected WBS tasks.

SILAP input parameters: distributions

“All” and “orbits”

Introduction

Knowledge Farming

SILAP

mb1

❖ Components

❖ separation

❖ “All” and “Orbits”

❖ “Orbits” (cont.)

❖ “Prime”

❖ “Profile”

❖ “Profile” (cont.)

❖ Selections

❖ Separation

❖ Selected WBS

❖ Selected WBS (more)

❖ Weighted Frequencies

What’s next?

Conclusion

All: 433 records

HS2	3 =0.00 5=0.02 2=0.02 4=0.05 1=0.35 0=0.56
AS2	5 =0.07 2=0.07 4=0.10 3=0.10 1=0.67
PF2	2 =0.09 5=0.13 4=0.16 1=0.19 3=0.43
EX3	5 =0.05 4=0.14 2=0.17 3=0.21 1=0.44
DO3	0.5=0.01 3=0.02 1=0.08 5=0.11 2=0.33 4=0.45
US3	5 =0.04 3=0.21 2=0.24 1=0.50
UC3	3 =0.07 2=0.28 1=0.65
CL3	2 =0.00 5=0.04 4=0.20 3=0.76
FR3	3 =0.01 2=0.34 1=0.65
DT3	4 =0.01 3=0.31 2=0.34 1=0.35
RM3	1 =0.18 3=0.36 2=0.45
RA3	2 =0.07 3=0.11 4=0.12 5=0.12 1=0.58
AM3	3 =0.15 2=0.21 1=0.64
CX3	5 =0.04 2=0.11 4=0.12 1=0.31 3=0.41
DI3	4 =0.02 3=0.02 2=0.12 1=0.85
SS3	1 =0.06 4=0.07 2=0.30 3=0.57

How to read these tables

- rows sorted left-to-right as rarest-to-more-frequent;
- column one are SILAP variable names (see code)

e.g. the last line of the above table;

- Least common SS3 value = “1”: occurs 6% of the time;
- Most common SS3 value = “3”: occurs 57% of the time;
- “2” occurs 30% of the time;

Orbit = eo2 (265 records)

HS2	3 =0.01 5=0.01 2=0.02 4=0.08 1=0.41 0=0.48
AS2	5 =0.03 2=0.08 4=0.09 3=0.11 1=0.69
PF2	5 =0.09 2=0.10 4=0.19 1=0.25 3=0.37
EX3	5 =0.07 2=0.15 3=0.17 4=0.17 1=0.44
DO3	0.5=0.02 3=0.03 5=0.04 4=0.45 2=0.46
US3	5 =0.07 3=0.14 2=0.22 1=0.57
UC3	3 =0.03 2=0.16 1=0.81
CL3	2 =0.00 5=0.07 4=0.14 3=0.79
FR3	3 =0.00 2=0.09 1=0.91
DT3	4 =0.01 3=0.11 2=0.31 1=0.56
RM3	1 =0.16 2=0.30 3=0.54
RA3	3 =0.00 4=0.01 2=0.12 5=0.12 1=0.74
AM3	3 =0.19 2=0.20 1=0.61
CX3	5 =0.04 2=0.07 4=0.11 3=0.37 1=0.41
DI3	4 =0.01 3=0.01 2=0.08 1=0.91
SS3	1 =0.01 4=0.04 2=0.41 3=0.55

Orbit = xf3 (93 records)

HS2	2=0.02 1=0.10 0=0.88
AS2	5=0.03 2=0.07 3=0.09 4=0.14 1=0.67
PF2	2=0.08 1=0.10 5=0.11 4=0.15 3=0.57
EX3	4=0.12 2=0.14 3=0.24 1=0.50
DO3	1=0.18 5=0.37 4=0.45
US3	3=0.21 2=0.24 1=0.55
UC3	3=0.03 1=0.36 2=0.61
CL3	4=0.39 3=0.61
FR3	1=0.29 2=0.71
DT3	1=0.03 2=0.48 3=0.49
RM3	1=0.23 2=0.77
RA3	3=0.12 1=0.39 4=0.49
AM3	3=0.03 2=0.13 1=0.84
CX3	5=0.03 4=0.11 2=0.20 1=0.21 3=0.46
DI3	3=0.03 2=0.13 1=0.84
SS3	2=0.13 1=0.27 3=0.60

SILAP input parameters: distributions

“orbits” (cont.)

Introduction

Knowledge Farming

SILAP

mb1

❖ Components

❖ separation

❖ “All” and “Orbits”

❖ “Orbits” (cont.)

❖ “Prime”

❖ “Profile”

❖ “Profile” (cont.)

❖ Selections

❖ Separation

❖ Selected WBS

❖ Selected WBS (more)

❖ Weighted Frequencies

What's next?

Conclusion

Orbit = go4 (57 records)

HS2	0=0.43 1=0.57
AS2	2=0.04 4=0.04 3=0.11 5=0.27 1=0.55
PF2	4=0.09 2=0.09 1=0.11 5=0.29 3=0.43
EX3	5=0.02 4=0.04 3=0.09 2=0.41 1=0.45
DO3	5=0.05 1=0.30 4=0.64
US3	1=0.27 3=0.27 2=0.46
UC3	1=0.27 3=0.32 2=0.41
CL3	4=0.23 3=0.77
FR3	3=0.07 1=0.25 2=0.68
DT3	3=1.00
RM3	1=0.25 3=0.27 2=0.48
RA3	4=0.05 1=0.30 3=0.64
AM3	3=0.21 1=0.38 2=0.41
CX3	1=0.09 5=0.11 2=0.12 4=0.20 3=0.48
DI3	3=0.05 4=0.09 2=0.30 1=0.55
SS3	2=0.18 3=0.82

Orbit = gs1 (20 records)

HS2	4=0.05 2=0.05 5=0.26 0=0.63
AS2	5=0.16 4=0.21 1=0.63
PF2	4=0.05 2=0.05 1=0.11 5=0.26 3=0.53
CX3	4=0.11 1=0.16 2=0.32 3=0.42
EX3	3=1.00
DO3	2=1.00
US3	3=1.00
UC3	1=1.00
CL3	3=1.00
FR3	2=1.00
DT3	2=1.00
RM3	2=1.00
RA3	5=1.00
AM3	1=1.00
DI3	1=1.00
SS3	4=1.00

SILAP input parameters: distributions “prime”

- Introduction
- Knowledge Farming
- SILAP
- mb1
- ❖ Components
- ❖ separation
- ❖ “All” and “Orbits”
- ❖ “Orbits” (cont.)
- ❖ “Prime”
- ❖ “Profile”
- ❖ “Profile” (cont.)
- ❖ Selections
- ❖ Separation
- ❖ Selected WBS
- ❖ Selected WBS (more)
- ❖ Weighted Frequencies
- What’s next?
- Conclusion

Prime = p1 (182 records)

HS2	2 =0.01 1=0.05 0=0.94
AS2	5 =0.01 2=0.04 4=0.08 3=0.08 1=0.78
PF2	2 =0.08 5=0.10 1=0.15 4=0.16 3=0.50
EX3	4 =0.06 5=0.10 3=0.23 2=0.28 1=0.33
DO3	0.5=0.03 3=0.04 1=0.09 2=0.16 5=0.23 4=0.45
US3	5 =0.10 1=0.17 3=0.29 2=0.44
UC3	3 =0.04 2=0.42 1=0.54
CL3	5 =0.10 4=0.27 3=0.63
FR3	2 =0.49 1=0.51
DT3	4 =0.02 1=0.02 3=0.28 2=0.69
RM3	3 =0.02 1=0.28 2=0.70
RA3	3 =0.04 2=0.12 5=0.18 4=0.27 1=0.39
AM3	3 =0.04 2=0.36 1=0.60
CX3	5 =0.04 2=0.09 4=0.12 3=0.33 1=0.41
DI3	4 =0.01 3=0.01 2=0.17 1=0.81
SS3	1 =0.15 2=0.28 3=0.57

Prime = p3 (139 records)

HS2	5=0.01 3=0.01 2=0.04 4=0.14 1=0.78
AS2	5=0.05 3=0.12 2=0.13 4=0.14 1=0.56
PF2	5=0.07 2=0.09 4=0.21 3=0.31 1=0.32
EX3	3=0.01 4=0.33 1=0.65
DO3	4=0.33 2=0.67
US3	3=0.01 1=0.99
UC3	3=0.01 1=0.99
CL3	2=0.01 3=0.99
FR3	3=0.01 1=0.99
DT3	3=0.01 1=0.99
RM3	3=1.00
RA3	3=0.01 1=0.99
AM3	3=0.25 1=0.75
CX3	5=0.01 2=0.11 4=0.12 1=0.35 3=0.42
DI3	4=0.01 3=0.01 1=0.98
SS3	4=0.07 3=0.45 2=0.48

prime=p2 (96 records)

HS2	1=0.35 0=0.65
AS2	4=0.03 2=0.05 3=0.13 5=0.18 1=0.61
PF2	1=0.09 4=0.12 2=0.13 5=0.23 3=0.43
EX3	5=0.01 4=0.02 2=0.25 3=0.29 1=0.42
DO3	5=0.06 1=0.20 4=0.74
US3	3=0.20 2=0.27 1=0.53
UC3	3=0.23 1=0.31 2=0.46
CL3	4=0.39 3=0.61
FR3	3=0.05 2=0.40 1=0.55
DT3	2=0.02 1=0.12 3=0.86
RM3	3=0.17 1=0.29 2=0.54
RA3	5=0.01 4=0.03 2=0.12 3=0.41 1=0.43
AM3	2=0.25 3=0.25 1=0.49
CX3	1=0.08 5=0.11 2=0.12 4=0.16 3=0.54
DI3	4=0.05 3=0.05 2=0.20 1=0.69
SS3	2=0.14 3=0.86

Prime = p4 (20 records)

HS2	4=0.05 2=0.05 5=0.26 0=0.63
AS2	5=0.16 4=0.21 1=0.63
PF2	4=0.05 2=0.05 1=0.11 5=0.26 3=0.53
EX3	3=1.00
DO3	2=1.00
US3	3=1.00
UC3	1=1.00
CL3	3=1.00
FR3	2=1.00
DT3	2=1.00
RM3	2=1.00
RA3	5=1.00
AM3	1=1.00
CX3	4=0.11 1=0.16 2=0.32 3=0.42
DI3	1=1.00
SS3	4=1.00

SILAP input parameters: distributions “profile”

Introduction

Knowledge Farming

SILAP

mb1

❖ Components

❖ separation

❖ “All” and “Orbits”

❖ “Orbits” (cont.)

❖ “Prime”

❖ **“Profile”**

❖ “Profile” (cont.)

❖ Selections

❖ Separation

❖ Selected WBS

❖ Selected WBS (more)

❖ Weighted Frequencies

What’s next?

Conclusion

Profile = es (29 records)

HS2	0=1.00
AS2	2=0.04 4=0.04 3=0.21 1=0.71
PF2	4=0.04 1=0.07 5=0.18 2=0.18 3=0.54
EX3	3=0.14 2=0.39 1=0.46
DO3	4=1.00
US3	3=0.07 1=0.46 2=0.46
UC3	3=0.07 2=0.39 1=0.54
CL3	3=1.00
FR3	1=0.46 2=0.54
DT3	2=0.14 1=0.39 3=0.46
RM3	3=0.14 2=0.39 1=0.46
RA3	1=0.14 2=0.86
AM3	2=0.04 1=0.39 3=0.57
CX3	5=0.14 1=0.14 4=0.29 3=0.43
DI3	2=0.04 1=0.96
SS3	1=0.04 3=0.46 2=0.50

Profile = hs (158 records)

HS2	3=0.01 5=0.04 2=0.04 0=0.08 4=0.13 1=0.69
AS2	5=0.06 3=0.10 2=0.11 4=0.15 1=0.57
PF2	2=0.09 5=0.09 4=0.19 1=0.29 3=0.34
EX3	3=0.13 4=0.29 1=0.57
DO3	4=0.29 2=0.71
US3	3=0.13 1=0.87
UC3	3=0.01 1=0.99
CL3	2=0.01 3=0.99
FR3	3=0.01 2=0.12 1=0.87
DT3	3=0.01 2=0.12 1=0.87
RM3	2=0.12 3=0.88
RA3	3=0.01 5=0.12 1=0.87
AM3	3=0.22 1=0.78
CX3	5=0.01 4=0.11 2=0.13 1=0.32 3=0.42
DI3	4=0.01 3=0.01 1=0.98
SS3	4=0.18 3=0.39 2=0.42

Profile = op (85 records)

HS2	2=0.02 1=0.44 0=0.54
AS2	2=0.04 4=0.07 3=0.08 5=0.19 1=0.62
PF2	2=0.08 1=0.10 4=0.14 5=0.23 3=0.45
EX3	5=0.01 3=0.07 4=0.15 2=0.33 1=0.43
DO3	5=0.08 1=0.43 4=0.49
US3	1=0.23 3=0.38 2=0.39
UC3	1=0.23 3=0.23 2=0.55
CL3	4=0.15 3=0.85
FR3	3=0.06 1=0.21 2=0.73
DT3	2=0.02 1=0.04 3=0.94
RM3	3=0.19 2=0.38 1=0.43
RA3	5=0.01 4=0.21 1=0.25 3=0.52
AM3	3=0.18 2=0.27 1=0.55
CX3	5=0.08 1=0.08 4=0.18 2=0.20 3=0.45
DI3	3=0.04 4=0.06 2=0.35 1=0.56
SS3	1=0.12 2=0.12 3=0.76

Profile = ss (111 records)

HS2	1=0.05 0=0.95
AS2	5=0.02 2=0.05 3=0.07 4=0.09 1=0.77
PF2	2=0.09 5=0.09 4=0.13 1=0.19 3=0.50
EX3	5=0.17 3=0.19 2=0.30 1=0.34
DO3	5=0.27 4=0.73
US3	2=0.14 5=0.17 3=0.25 1=0.44
UC3	3=0.03 2=0.47 1=0.50
CL3	5=0.17 4=0.33 3=0.50
FR3	2=0.38 1=0.62
DT3	3=0.22 2=0.78
RM3	1=0.17 2=0.83
RA3	3=0.03 4=0.27 1=0.70
AM3	2=0.28 1=0.72
CX3	5=0.02 4=0.06 2=0.07 3=0.29 1=0.55
DI3	3=0.03 2=0.18 1=0.79
SS3	1=0.14 2=0.34 3=0.53

SILAP input parameters: distributions “profile” (cont.)

Introduction

Knowledge Farming

SILAP

mb1

❖ Components

❖ separation

❖ “All” and “Orbits”

❖ “Orbits” (cont.)

❖ “Prime”

❖ “Profile”

❖ “Profile” (cont.)

❖ Selections

❖ Separation

❖ Selected WBS

❖ Selected WBS (more)

❖ Weighted Frequencies

What's next?

Conclusion

Profile = su (55 records)

HS2	1=0.05 0=0.95
AS2	5=0.02 2=0.05 3=0.07 4=0.09 1=0.77
PF2	2=0.09 5=0.09 4=0.13 1=0.19 3=0.50
EX3	5=0.17 3=0.19 2=0.30 1=0.34
DO3	5=0.27 4=0.73
US3	2=0.14 5=0.17 3=0.25 1=0.44
UC3	3=0.03 2=0.47 1=0.50
CL3	5=0.17 4=0.33 3=0.50
FR3	2=0.38 1=0.62
DT3	3=0.22 2=0.78
RM3	1=0.17 2=0.83
RA3	3=0.03 4=0.27 1=0.70
AM3	2=0.28 1=0.72
CX3	5=0.02 4=0.06 2=0.07 3=0.29 1=0.55
DI3	3=0.03 2=0.18 1=0.79
SS3	1=0.14 2=0.34 3=0.53

500 times * {pick SILAP inputs from known distributions; compute (EP1,CO1); select relevant WBS tasks}

Introduction

Knowledge Farming

SILAP

mb1

❖ Components

❖ separation

❖ “All” and “Orbits”

❖ “Orbits” (cont.)

❖ “Prime”

❖ “Profile”

❖ “Profile” (cont.)

❖ **Selections**

❖ Separation

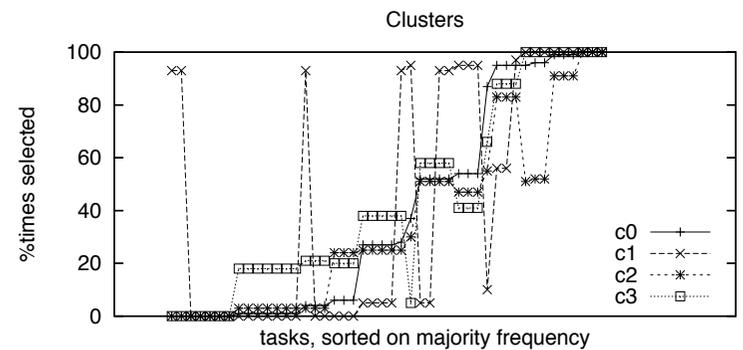
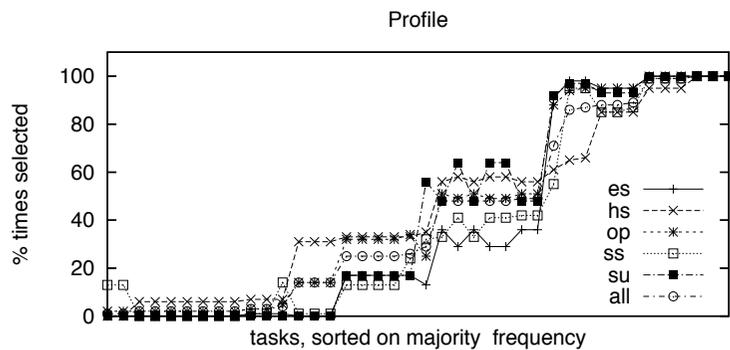
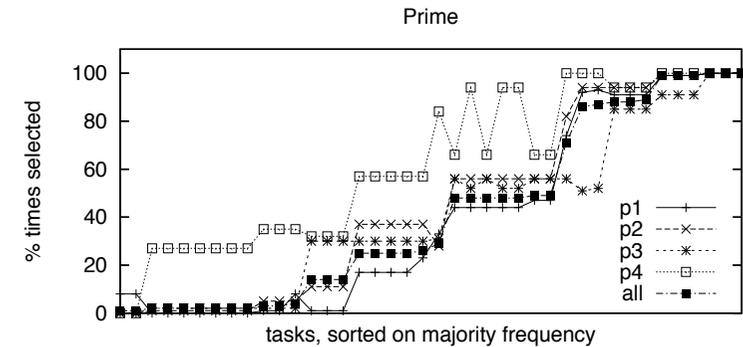
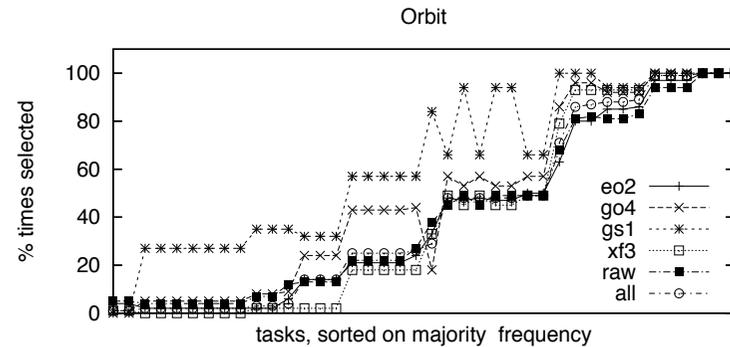
❖ Selected WBS

❖ Selected WBS (more)

❖ Weighted Frequencies

What's next?

Conclusion



- “Profile” offers least separation;
- “Orbit” and “Prime” very similar
- “Orbits”, “prime”, and “profile” select for very similar tasks.
- “Clusters” offers the most separation

Sanity check: sampling naive?

- The proceeding experiment drew values from probability distributions of each SILAP variable, *ignoring* any correlations *between* variables.
- Q: are there large correlations?
A: No

	HS2	AS2	PF2	EX3	DO3	US3	UC3	CL3	FR3	DT3	RM3	RA3	AM3	CX3	DI3	SS3	
HS2		0.41	0.08	-0.06	-0.18	-0.26	-0.22	-0.29	-0.09	-0.42	0.47	-0.16	-0.05	0.07	0	0.13	
AS2			0.44	-0.05	-0.06	-0.16	-0.05	-0.21	0.05	0.01	0.04	-0.01	-0.1	0.34	0.14	0.07	
PF2				0.01	0.01	-0.02	0.07	-0.07	0.08	0.26	-0.15	0.16	-0.12	0.4	0.12	0.08	
EX3					-0.19	0.45	0.21	0.47	-0.11	0.1	-0.19	0.05	0.41	-0.09	0.24	0.18	
DO3						0.13	0.35	0.18	0.2	0.23	-0.13	-0.07	-0.05	-0.04	0.09	-0.23	
US3							0.44	0.48	0.26	0.43	-0.47	0.29	0.03	-0.22	0.29	0.1	
UC3								0.39	0.49	0.68	-0.23	0.08	0.14	0.13	0.34	0.01	
CL3									-0.13	0.21	-0.31	0.03	0.24	-0.16	0.17	0.11	
FR3										0.5	-0.12	0.53	-0.02	0.15	0.21	-0.03	
DT3											-0.58	0.34	-0.12	0.24	0.3	0.09	
RM3												-0.28	0.02	-0.04	-0.23	-0.12	
RA3													0.02	0.15	0.06	0.2	
AM3														0.01	0.11	-0.01	
CX3															0.05	0.14	
DI3																0.18	
SS3																	

- Besides, when we run the raw data for “all” through SILAP, it picks tasks very similar to the simulation for “all” .

- Introduction
- Knowledge Farming
- SILAP
- mb1
- ❖ Components
- ❖ separation
- ❖ “All” and “Orbits”
- ❖ “Orbits” (cont.)
- ❖ “Prime”
- ❖ “Profile”
- ❖ “Profile” (cont.)
- ❖ Selections
- ❖ Separation
- ❖ Selected WBS
- ❖ Selected WBS (more)
- ❖ Weighted Frequencies
- What’s next?
- Conclusion

What Separates the Clusters?

Introduction

Knowledge Farming

SILAP

mb1

❖ Components

❖ separation

❖ “All” and “Orbits”

❖ “Orbits” (cont.)

❖ “Prime”

❖ “Profile”

❖ “Profile” (cont.)

❖ Selections

❖ Separation

❖ Selected WBS

❖ Selected WBS (more)

❖ Weighted Frequencies

What's next?

Conclusion

- “Tag” each record with its cluster identifier
- Using feature subset selection, learn which SILAP attributes are most important
 - ◆ 10-way, CFS, selected: **HS2, EX3, US3, CL3, FR3, DT3, RM3**
- Learn a decision procedure that identifies each clusters
- If SILAP performs differently for each cluster, then those clusters represent truly different project types.

DT3= use of defect tracking; CL3= CMM level; US3= use of standards;
EX3= experience; HS2= human safety

```
DT3 <= 1: cluster2 (150.0/4.0)
DT3 > 1
|   CL3 <= 4
|   |   US3 <= 2: cluster0 (170.0)
|   |   |   US3 > 2
|   |   |   |   DT3 <= 2
|   |   |   |   |   EX3 <= 2: cluster3 (25.0)
|   |   |   |   |   |   EX3 > 2
|   |   |   |   |   |   |   HS2 <= 3: cluster0 (14.0)
|   |   |   |   |   |   |   |   HS2 > 3: cluster3 (6.0)
|   |   |   |   |   |   |   |   |   DT3 > 2: cluster0 (42.0)
|   |   |   |   |   |   |   |   |   |   CL3 > 4: cluster1 (19.0)
```

In a 10-way cross-val, accuracy=99.061% (!!).

“The variables in the tree are all about how much a project knows about itself and how much it is willing to share that knowledge with others.”

What WBS tasks are Selected by the Clusters?

- Introduction

- Knowledge Farming

- SILAP

- mb1

- ❖ Components
- ❖ separation
- ❖ “All” and “Orbits”
- ❖ “Orbits” (cont.)
- ❖ “Prime”
- ❖ “Profile”
- ❖ “Profile” (cont.)
- ❖ Selections
- ❖ Separation
- ❖ Selected WBS
- ❖ Selected WBS (more)
- ❖ Weighted Frequencies

- What’s next?

- Conclusion

id	task	W=231	X=21	Y=148	Z=32
		c0	c1	c2	c3
1.1	Management and Planning of IV&V	100	100	100	100
1.2	Issue and Risk Tracking	99	100	91	100
1.3	Final Report Generation	99	100	91	100
1.4	IV&V Tool Support	99	100	91	100
1.5	Management & Technical Review Support	100	100	100	100
1.6	Criticality Analysis	100	100	100	100
2.1	Reuse Analysis*	87	10	55	66
2.2	Software Architecture Assessment	52	5	51	58
2.3	System Requirements Review	52	93	51	58
2.4	Concept Document Evaluation	6		24	20
2.5	SW/User Requirements Allocation Analysis	6		24	20
2.6	Traceability Analysis	6		24	20
3.1	Traceability Analysis - Requirements	95	97	83	88
3.2	Software Requirements Evaluation	52	93	51	58
3.3	Interface Analysis - Requirements	54	95	47	41
3.4	System Test Plan Analysis	52	5	51	58
3.5	Acceptance Test Plan Analysis	4		3	21
3.6	Timing and Sizing Analysis				
4.1	Traceability Analysis - Design	95	100	51	100
4.2	Software Design Evaluation	28	93	25	38
4.3	Interface Analysis - Design	4	93	3	21
4.4	Software FQT Plan Analysis	95	56	83	88
4.5	Software Integration Test Plan Analysis		93		
4.6	Database Analysis	37	95	30	5
4.7	Component Test Plan Analysis				

What WBS tasks are Selected by the Clusters? (more)

Introduction

Knowledge Farming

SILAP

mb1

❖ Components

❖ separation

❖ “All” and “Orbits”

❖ “Orbits” (cont.)

❖ “Prime”

❖ “Profile”

❖ “Profile” (cont.)

❖ Selections

❖ Separation

❖ Selected WBS

❖ Selected WBS (more)

❖ Weighted Frequencies

What's next?

Conclusion

id	task	W=231	X=21	Y=148	Z=32
		c0	c1	c2	c3
5.1	Traceability Analysis - Code	96	100	52	100
5.10	Component Test Case Analysis	96	100	52	100
5.11	System Test Procedure Analysis	1		3	18
5.12	Software FQT Procedure Analysis	1		3	18
5.2	Source Code and Documentation Evaluation	54	95	47	41
5.3	Interface Analysis - Code	54	95	47	41
5.4	System Test Case Analysis	27	5	25	38
5.5	Software FQT Case Analysis	27	5	25	38
5.6	SW Integration Test Case Analysis				
5.7	Acceptance Test Case Analysis	4		3	21
5.8	SW Integration Test Procedure Analysis				
5.9	SW Integration Test Results Analysis		93		
6.1	Traceability Analysis - Test	95	56	83	88
6.2	Regression Test Analysis				
6.3	Simulation Analysis	1		3	18
6.4	System Test Results Analysis	27	5	25	38
6.5	Software FQT Results Analysis	27	5	25	38
7.1	Operating Procedure Evaluation	1		3	18
7.2	Anomaly Evaluation	1		3	18
7.3	Migration Assessment	1		3	18
7.4	Retirement Assessment	1		3	18

But what *good* is any of this? Well...

- Check the above. Is it reasonable? If not- should SILAP be changed?
- Don't make “blunt” business distinctions;
 - ◆ e.g. don't plan IV&V around “mission type” but, rather, around the process maturity of the developers.
- Make business decisions that are sensitive to the kinds of current business; e.g.
- What are the *current* and *expected future* frequencies of {W,X,Y,Z}?
- Weight the above numbers by those frequencies.

Weighted Frequencies

<u>Introduction</u>	1.1	Management and Planning of IV&V	100	*****
<u>Knowledge Farming</u>	1.5	Management & Technical Review Support	100	*****
<u>SILAP</u>	1.6	Criticality Analysis	100	*****
<u>mb1</u>	1.2	Issue and Risk Tracking	96	*****
❖ Components	1.3	Final Report Generation	96	*****
❖ separation	1.4	IV&V Tool Support	96	*****
❖ “All” and “Orbits”	3.1	Traceability Analysis - Requirements	90	*****
❖ “Orbits” (cont.)	4.4	Software FQT Plan Analysis	88	*****
❖ “Prime”	6.1	Traceability Analysis - Test	88	*****
❖ “Profile”	4.1	Traceability Analysis - Design	81	*****
❖ “Profile” (cont.)	4.1	Traceability Analysis - Code	81	*****
❖ Selections	5.1	Traceability Analysis - Code	81	*****
❖ Separation	5.10	Component Test Case Analysis	81	*****
❖ Selected WBS	2.1	Reuse Analysis*	71	*****
❖ Selected WBS (more)	2.3	System Requirements Review	54	*****
❖ Weighted Frequencies	3.2	Software Requirements Evaluation	54	*****
<u>What's next?</u>	3.3	Interface Analysis - Requirements	53	*****
<u>Conclusion</u>	5.2	Source Code and Documentation Evaluation	53	*****
	5.3	Interface Analysis - Code	53	*****
	2.2	Software Architecture Assessment	50	*****
	3.4	System Test Plan Analysis	50	*****
	4.6	Database Analysis	35	*****
	4.2	Software Design Evaluation	31	****
	5.4	System Test Case Analysis	26	****
	5.5	Software FQT Case Analysis	26	****
	6.4	System Test Results Analysis	26	****
	6.5	Software FQT Results Analysis	26	****
	2.4	Concept Document Evaluation	13	*
	2.5	SW/User Requirements Allocation Analysis	13	*
	2.6	Traceability Analysis	13	*
	4.3	Interface Analysis - Design	9	

Weighted Frequencies (more)

Introduction

Knowledge Farming

SILAP

mb1

- ❖ Components
- ❖ separation
- ❖ “All” and “Orbits”
- ❖ “Orbits” (cont.)
- ❖ “Prime”
- ❖ “Profile”
- ❖ “Profile” (cont.)
- ❖ Selections
- ❖ Separation
- ❖ Selected WBS
- ❖ Selected WBS (more)
- ❖ **Weighted Frequencies**

What's next?

Conclusion

3.5	Acceptance Test Plan Analysis	5
4.5	Software Integration Test Plan Analysis	5
5.7	Acceptance Test Case Analysis	5
5.9	SW Integration Test Results Analysis	5
5.12	Software FQT Procedure Analysis	3
5.11	System Test Procedure Analysis	3
6.3	Simulation Analysis	3
7.1	Operating Procedure Evaluation	3
7.2	Anomaly Evaluation	3
7.3	Migration Assessment	3
7.4	Retirement Assessment	3
3.6	Timing and Sizing Analysis	0
4.7	Component Test Plan Analysis	0
5.6	SW Integration Test Case Analysis	0
5.8	SW Integration Test Procedure Analysis	0
6.2	Regression Test Analysis	0

- Assess current capabilities:
 - ◆ how well do we do the most frequent tasks seen in current practice?
- Gap analysis:
 - ◆ what aren't we doing now that we will be doing more of in the future? Are we ready for that jump?

Introduction

Knowledge Farming

SILAP

mb1

What's next?

❖ Support

❖ Extensions

Conclusion

What's next?

Supporting Process Improvement

Introduction

Knowledge Farming

SILAP

mb1

What's next?

❖ Support

❖ Extensions

Conclusion

The IV&V planning & scoping team are continually and actively reviewing and improving SILAP. Current activities include:

- Review/update SLP 9-1 (the WBS)
- Revising SILAP documentation
- identified areas for ?change:
 - ◆ right tasks selected by scores?
 - ◆ are factors the best selectors?
- Is the criteria sufficient/ correct?
 - ◆ Map factors directly to tasks?
 - ◆ Study planned vs actual to find a “best” or most common architecture break out?

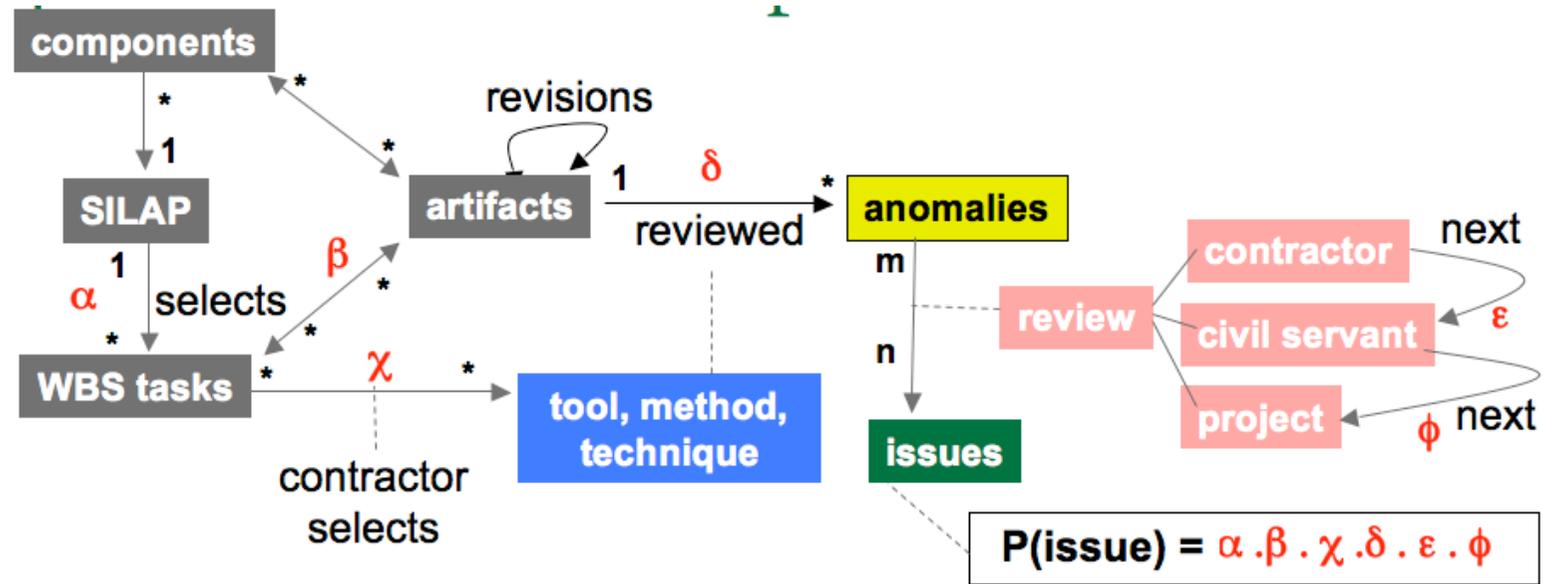
Minimum set of tasks needed to add IV&V value?

- ◆ Defining sets of ”common” tasks for specific types of functions?
- ?? break code analysis out into
 - ◆ tool execution only?
 - ◆ tool execution + review of results?
 - ◆ full-up code inspection?
- Etc etc

Can knowledge farming simplify, optimize any of these dialogues?

The Future

- Introduction
- Knowledge Farming
- SILAP
- mb1
- What's next?
- ❖ Support
- ❖ Extensions
- Conclusion



- This talk has been all about learning α .
- The rest of the alphabet awaits.

Conclusion

Introduction

Knowledge Farming

SILAP

mb1

What's next?

Conclusion

- If puzzled, then poke:
 - ◆ Model a little,
 - ◆ Simulate a little,
 - ◆ Summarize a little
- If certain, then check:
 - ◆ Model a little,
 - ◆ Simulate a little,
 - ◆ Summarize a little
- There are surprises hiding in your business.
- Use knowledge farming to find them.