

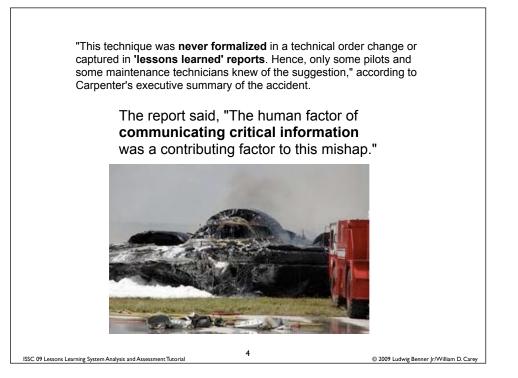


This is a B2 Spirit Bomber – cost estimates range from 1.4 – 2.2 billion depending on who you talk to. Most technologically fantastic ever &c. Flies from Missouri, can hit targets anywhere. We had 21 of them.



May have seen it on the news. Pilots eject in frame 3. Plane go boom.

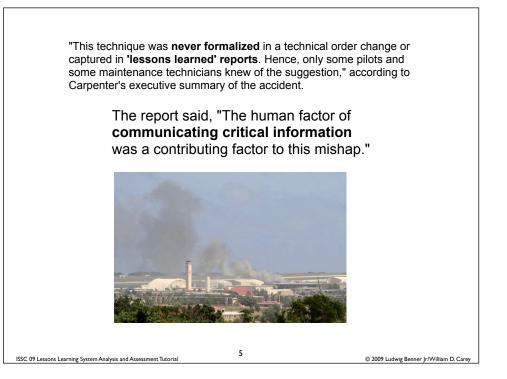
Now we have 20 bombers.



From Maj. Gen. Floyd L. Carpenter, who headed an accident investigation board. (AP News)

A lesson was not learned.

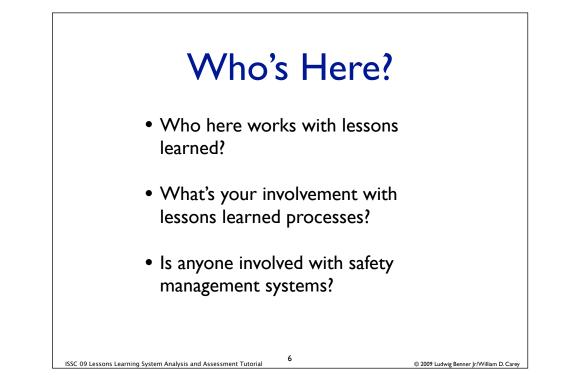
The Cost: well over \$1 billion + diminished capability



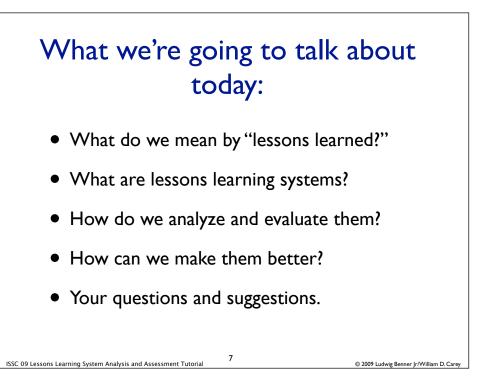
From Maj. Gen. Floyd L. Carpenter, who headed an accident investigation board. (AP News)

A lesson was not learned.

The Cost: well over \$1 billion + diminished capability



Poll class 1 Q at a time

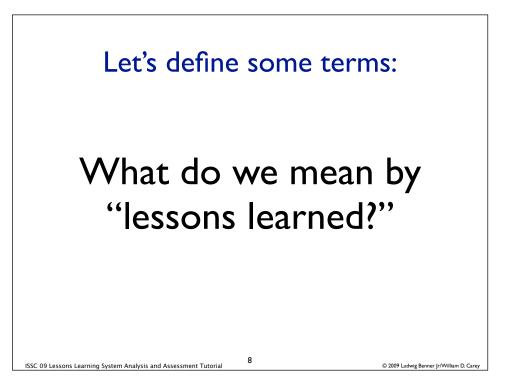


What do we mean by LL What systems produce them How do you evaluate yours? How you can make them better

We'll lead, but your willingness to share will pay dividends for everyone.

Scope of tutorial includes everything from the development of a lesson to the ultimate verification by an end user that a change inspired by the lesson has been learned and produced the desired results.

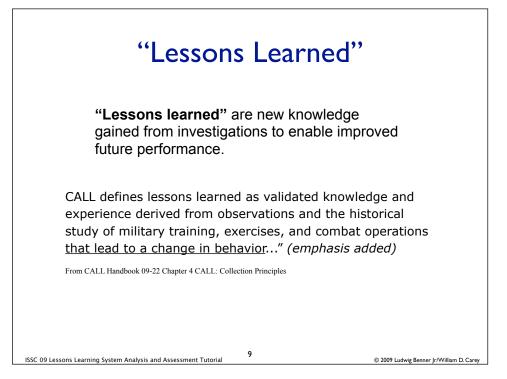
Do you work with LL of any kind in your jobs?



Hey, this is not just theoretical – it affects LL system design decisions – strategy and attributes, and resultant lessons learned system performance. (STOP)

Ask participants to write by No. 1 on our handout (blank paper) and jot down what they think lessons learned are,

Later, then ask them to read them back as time and participation dictates



What do we mean by Lessons Learned?

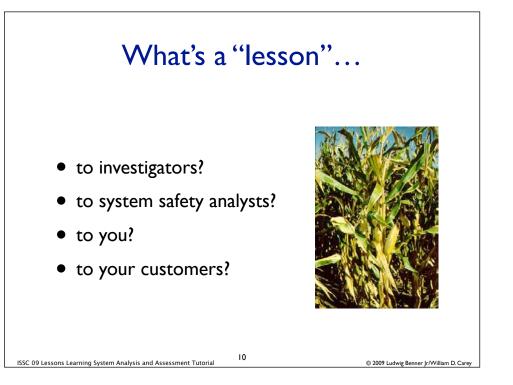
Actually, on the surface there's pretty general agreement about **meaning** of term – but abstraction masks differences in the actual lessons developed in practice, as we'll see.

Focus of tutorial is on lessons learned during mishap investigations – both **before and after they happen**, so think hazard and risk analyses which are "pre-mishap" investigations, as well as post-mishap investigations.

Both provide "new" knowledge from experience.

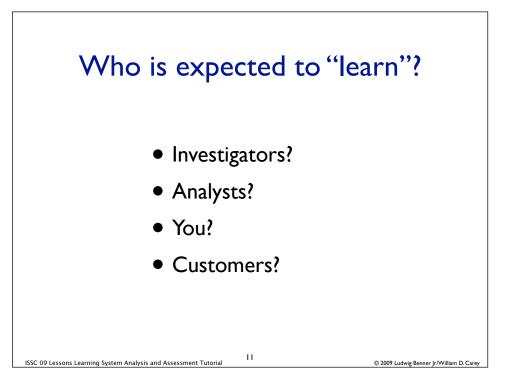
Pretty ambiguous, although with CALL, a little less. We like the part of call that says "change in behavior" but what follows is ambiguous –

What's doctrine, training or education behavior change? Aren't we aiming for design and operational behavior changes in people, objects or energies? Doesn't leadership behavior determine organization behavior? We can guess, but should we have to.....

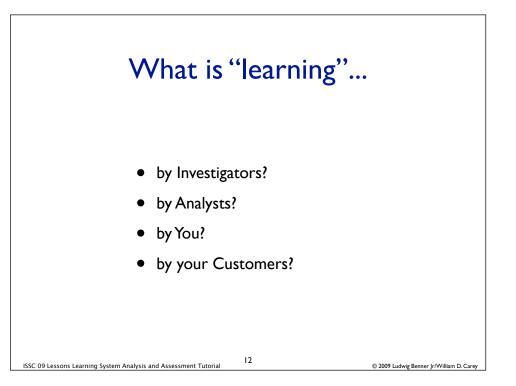


Shows different views of lessons.

- Investigators report causes, cause factors, root causes, issues, errors, failures, and recommendations, etc. as lessons learned, but don't report as lists "lessons" or "what was learned"
- Hazard analysts report hazards found, but no "lessons learned" per se
- for you personally ?
- knowledge they can act on –



- Investigators learn what happened and produce recommendations that if implemented are presumed to fix the reported causes, factors, issues or whatever
- Hazard analysts learn about past hazards and mishaps from ?
- You? What are you expected to learn?
- CALL (as a customer) wants "change in behavior...of doctrine, organization, training, material, leadership and education, personnel and facilities domains "



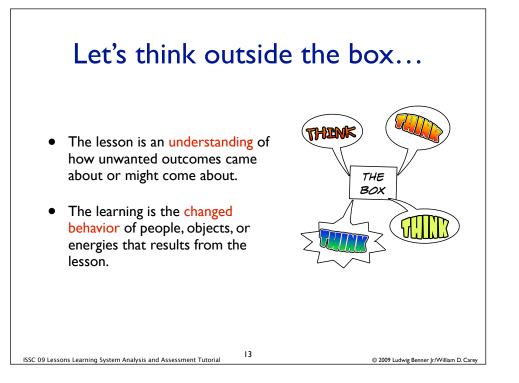
use to bring out different views of learning 2 kinds of learning - 1 = lessons from accidents, 2 = changes from lessons.

AI - learn new knowledge about what happened

Analysts - learn new knowledge about what might happen

by you - how things can go bad

by your customers - changes made due to lessons made available to them



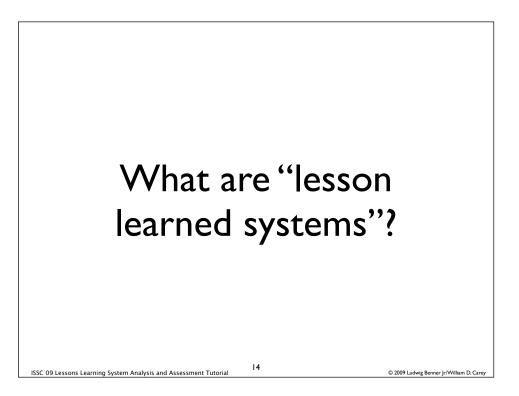
For this tutorial, let's adopt a couple of meanings that help us achieve our tutorial objectives LESSON

how unwanted outcome came about = investigation output

how unwanted outcome might come about = analysis output

LEARNING

Changed some behavior (s)

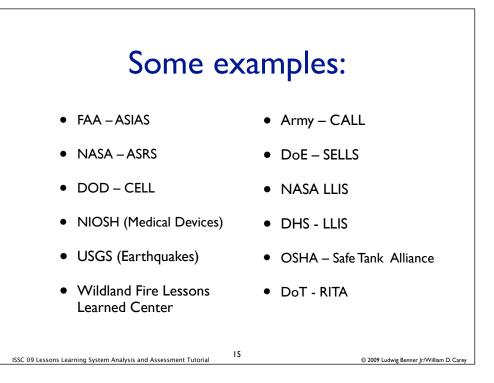


They are the systems – the interacting components – that produce the the changed behaviors learned from lessons derived from mishaps.

They are the systems

- the interacting components -

that produce changed behaviors learned from lessons generated during mishaps.



Here are just a few examples of governmental Lessons Learned systems, all of which strive to improve safety performance, based on mishap experience, through the acquisition and processing of Lessons Learned. Most private organizations have LL systems with widely varying degrees of sophistication.

ASIAS = Aviation Safety Information, Analysis and Sharing Center

NASA – ASRS = Aviation Safety Reporting System

DOD - CELL Center for Engineering Lessons Learned

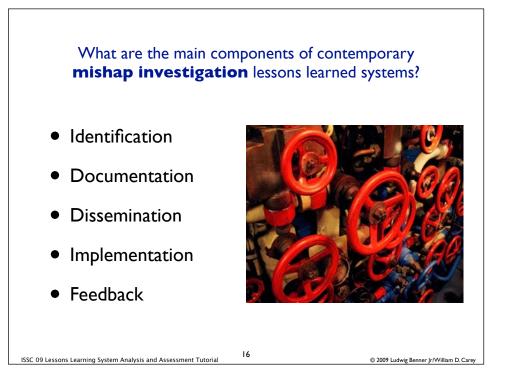
Army - CALL - Center for Army Lessons Learned

DoE – SELLS = Society for Effective Lessons Learned Sharing

NASA LLIS =Lesson Learning Information system

OSHA -Safe Tank Alliance

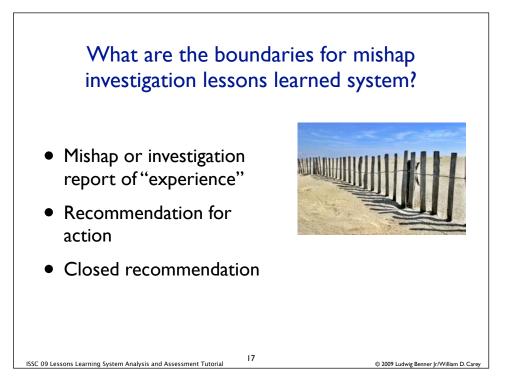
DoT – RITA Research and Innovative Technologies Lessons Learned reports for programs And they are all different, but with some common components.



Lets take a look at some practices that constitute the system within your organizations.

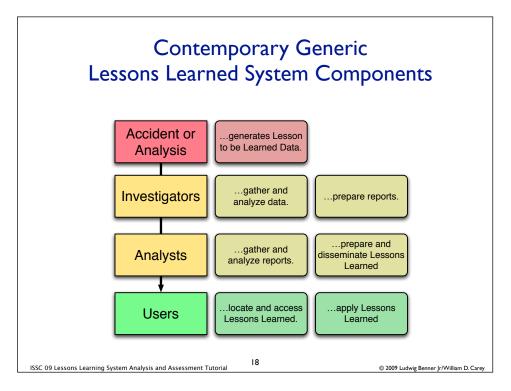
What are the components or elements that make up an **investigation** lessons learned System that you can discuss - your own, preferably, but anybody else's you care to mention.

First, lets try to walk through the system, starting with the occurrence of a mishap After that, we'll do the same for a hazard or risk analysis system - how are lessons learned developed and handled there?



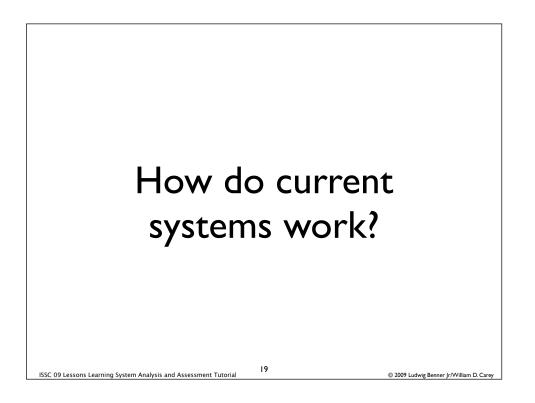
What do we mean by system boundaries?

how components of a constituting a system are limited by definition.



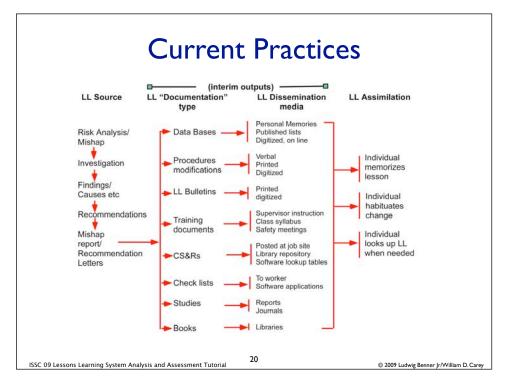
If you think of the components and functions that produce LL from mishaps as a system, here is a flow chart showing a generic system we synthesized from 7 selected accident investigation related LL processes. The role of analysts in the system operation is especially noteworthy, and reflects a strategic system design decision, as you will see later

We also found it useful to think of the data generated by an accident or incident as raw "lessons-to-be-learned" data from which lessons must be developed.



Poll class...

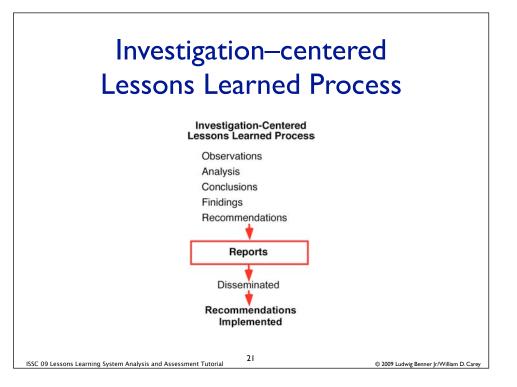
Well, lets take a look at current systems, and see how we found they work.



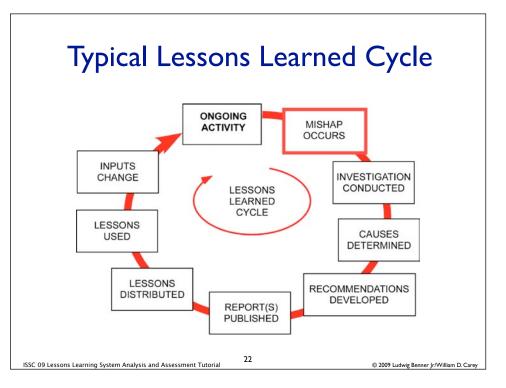
Here's a general representation of current mishap lessons learned system practices we put together before today, and how they fit together.

Note that there's a lot of action between the occurrence of the incident and the final assimilation of lessons learned.

(LL Process flow.png)



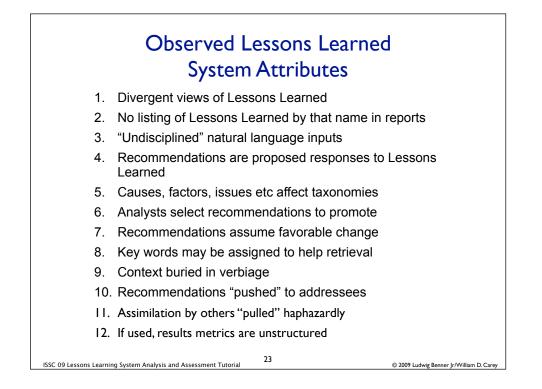
Here is how we thought to represent the way investigations lead to changes. This follows present ideas about investigations



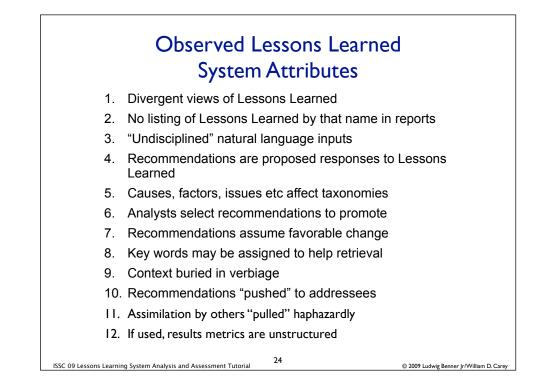
Now, if we look at an ongoing activity, the lessons learned system needs to be continuous if the organization is to be a "learning organization" of the kind envisioned by Senge in his book, The Fifth Discipline: The Art and Practice of the Learning Organization. That's an organization that joins adaptive learning with "generative" learning to enhance its capacity to create its future.

The mishap lessons learning system must be a continuous "loop" where experience changes inputs to the ongoing activity as it becomes available.

This graphic illustrates the nature of present mishap lessons learned cycles.



Here's a quick summary of the attributes we found during research of lessons learned systems. It's worth spending a moment highlighting some of them like 1, 2, 3, 5, 8, 9, 10

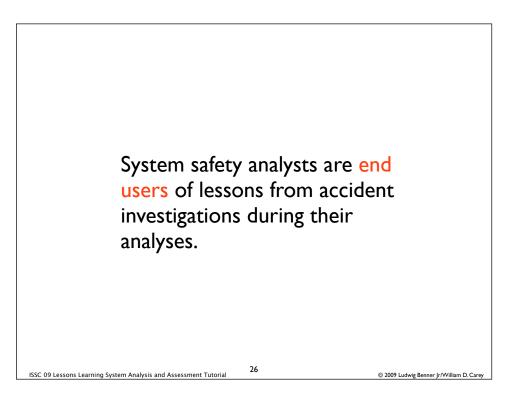


Here's a quick summary of the attributes we found during research of lessons learned from investigations. This is just how things are, not intended to be good or bad.

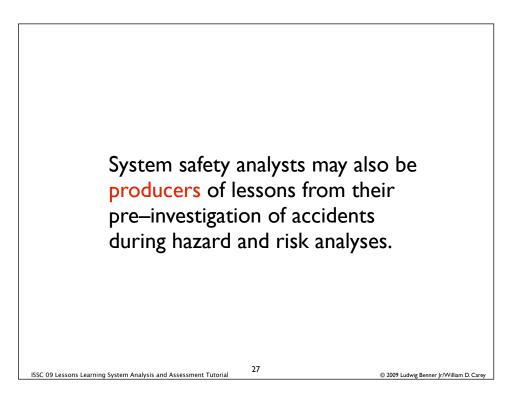
It's worth spending a moment highlighting some of them



let's talk about this a moment now – to introduce you to our way of looking at analysts role they might have, now or later

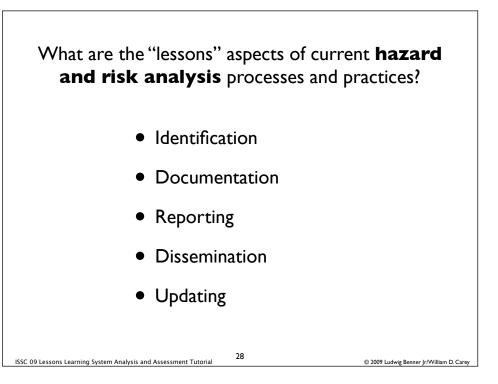


This is main thrust of tutorial. LL should be available to help you with your work.



Might this apply to analysts? Anybody want to comment on this notion

As tutorial progresses, we'd like to keep this possibility in mind as we get into criteria.



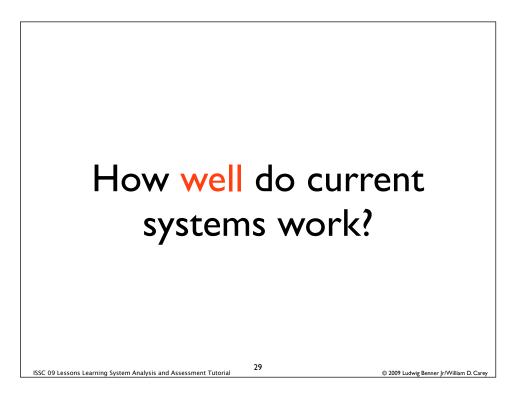
Aren't hazard analyses in reality mishap investigations before they happen?

Analysts just hypothesize what <u>might</u> happen rather than reconstructing what <u>did</u> happened.

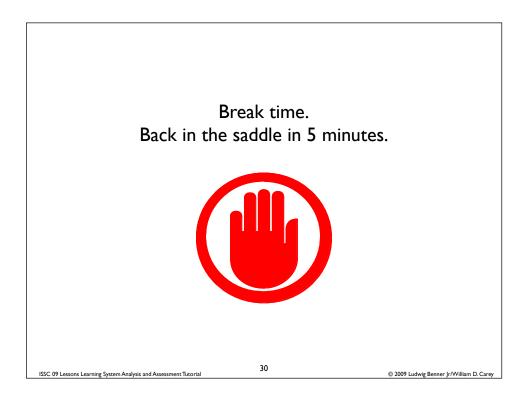
We could walk through each step of hazard and risk analysis lessons learned practices like we just did post-mishap investigations.

Introduces an approach for defining safety analyst's predictive Lessons Learned "system"

Could spend a lot of time looking into this little hummer – but we won't (later, Tutorial offers approach and questions you can take home)



Talk with people here.





Identification of lessons learning system components was derived in part from personal experiences with functions and actions required to bring about successful behavioral changes in people, objects and energies through accident investigations at the national level.

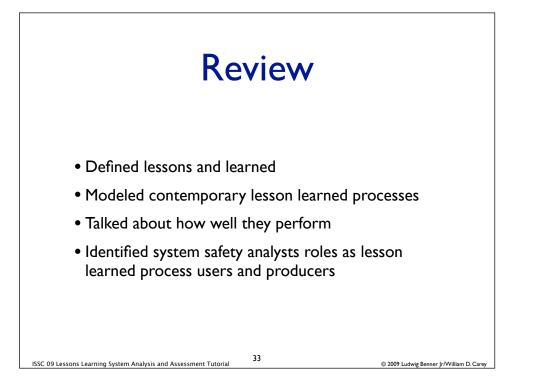
This accident killed a firefighter training officer. Our finding out what happened, and subsequent tasks, eventually led to major behavioral changes in the US fire services' responses.

Oh, there were some observations during unsuccessful efforts, too.

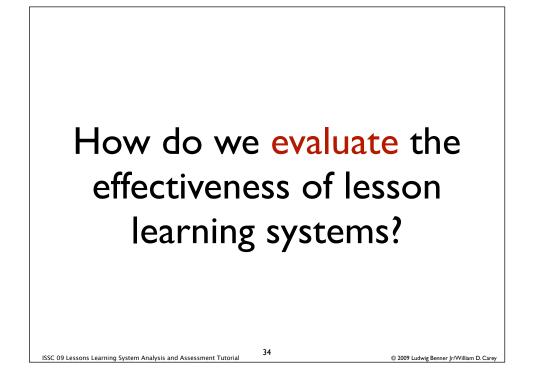


Have we left you out in the cold so far?

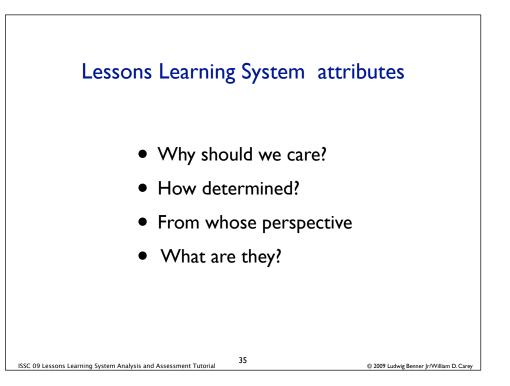
Well lets start poking into the details to see if that helps.



Here's where we are.



Talk with people here.

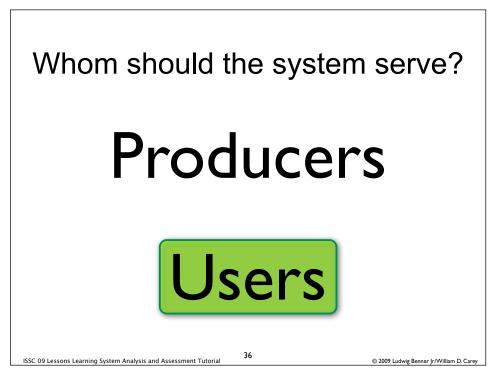


Why care? You need to know what is desirable to determine if what you have is OK, and to

distinguish relative quality of alternative strategies and systems

To determine them, we called on experiences in bringing about changes in the past, applied systems analysis tools to track the progression of lesson data through the system, and we looked at how contemporary processes worked and what problems they seemed to pose

(Possibly digression about Karl Popper)



The system must help lessons producers and users.

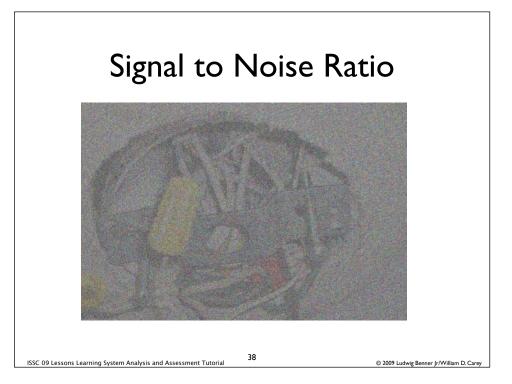
As we analyzed the necessary system operation, it soon became apparent to us that by differentiating between the learning and the lesson creation functions, the main "driver" for a well-performing lessons learning system design must be the users' perspective and resultant needs, rather that the investigators' or analysts perspectives of their own needs and outputs.

However both must be accommodated. So we divided the system into two parts- one part for user functions and the other part for developer functions.

- must be designed and optimized to provide lessons to users who can bring about better behaviors in people, objects or energies.
- design should NOT be driven by investigators' or analysts' perceptions of prevention

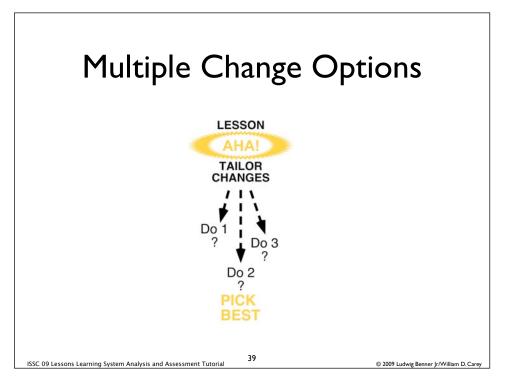


Let's take a what we expect from the lessons themselves. What attributes are important to users?

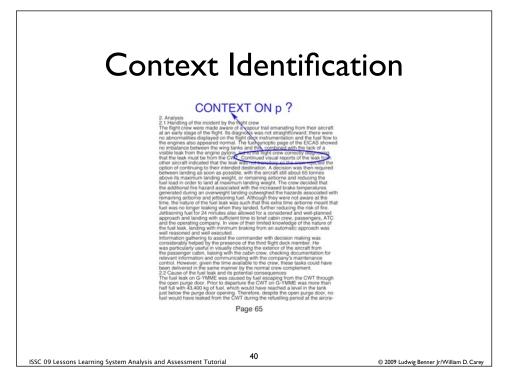


A frequent user complaint is the quantity of data that must be searched to find the morsel of interest to a potential LL user. It is difficult to acquire good messages from "noisy" data. This is mostly due to use of unstructured narrative form and vocabularies of LL. Lessons learning system design must address this signal-to-nose maximization challenge.

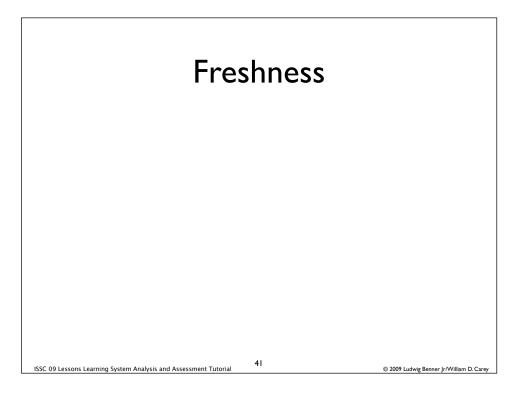
Another way to look at it is to think of "data density" of the documentation within which the LL are contained.



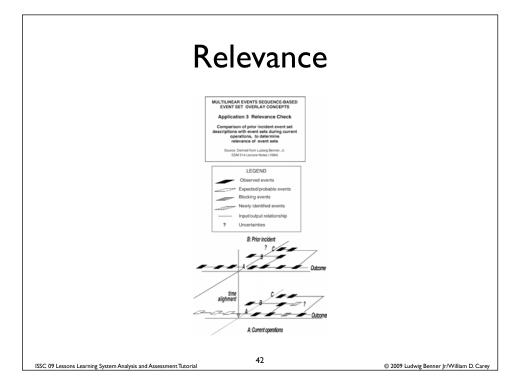
Recommendations by analysts rarely offer options for fixing "lessons learned" by investigations, so they can be <u>tailored to the specific</u> <u>activities of users</u>. There are exceptions, particularly in some engineering lessons learned processes. Ambiguously worded recommendations requiring interpretations, it might be argued, offer tailoring opportunities, but that poses other problems.



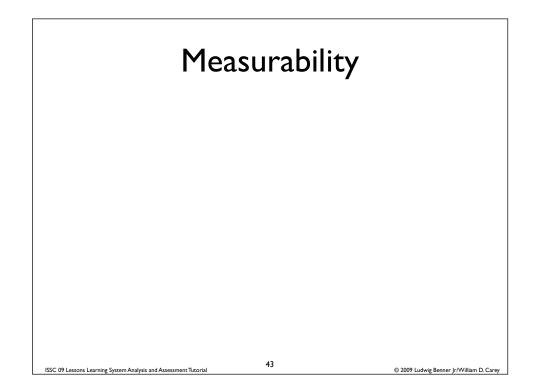
Documented lessons learned need to provide some form of context data for each lesson, to help users understand what happened, with the context in which it happened. Formal accident reports often contain the context, somewhere in the narrative if one has the time and skill to find it. Context identification should be directly discernible, not a treasure hunt.

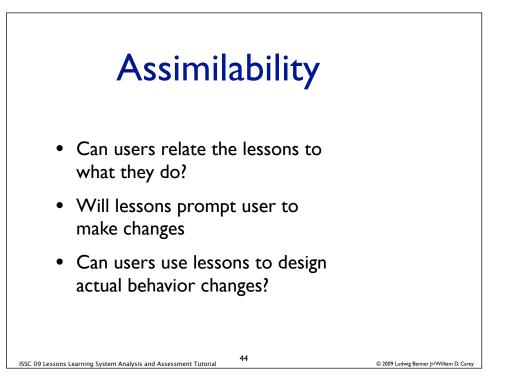


Overtaken by events?



Determining relevance of an accessed lesson is a subjective decision by a user. Users need to be able to "overlay" the lesson onto their activities. The longer this decision takes, the greater the disincentive for the user to use the system, so this is a significant design consideration.





Assimilation is the absorption or integration and use of lessons for one's benefit. The spotty record of assimilation and achievement of new behaviors, for contemporary processes, raises the question: what is the best way to ensure maximized assimilability of a lesson by a potential user?

Little research of lessons learned assimilability has been done to our knowledge, so the choice presently seems to depend on episodic observations or logic. But intuitively, simplicity and clarity seem to be essential.

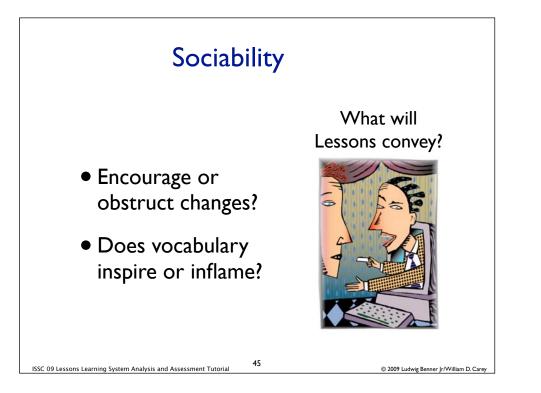
When a user finds lessons, how does a user know they're relevant to what he or she is doing now? (Relevance)

Is the context readily discernible?

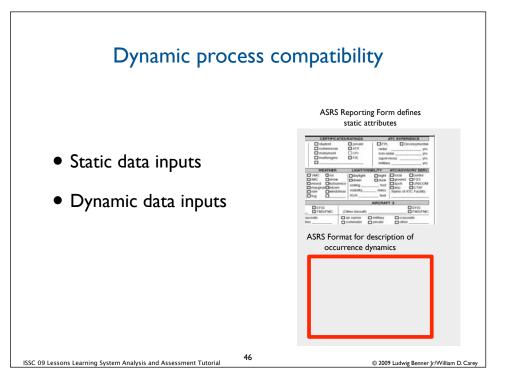
Will the lessons data actually prompt user to change what they are doing

Is lesson an actionable item for them - something they can change in how or what they do in their own tasks or activities?

Will they be there in the future if users want to refresh their memories until the desired behavior is a habituated?



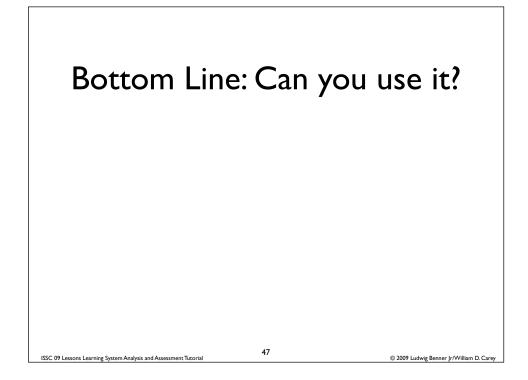
Sociability is a subtle system attribute to consider. Sociability of lessons, or how lessons fare in the social milieu after they are "published," poses at least two kinds of challenges - creating a climate to <u>encourage</u> the behavior changes, and avoiding a climate for <u>obstructing</u> changes. Restricted sharing of the lessons, for example, can <u>obstruct</u> changes. The vocabulary used to document lessons can <u>inflame</u> or <u>inspire</u> reactions to their documentation and dissemination, and also needs to be considered in system design decisions.

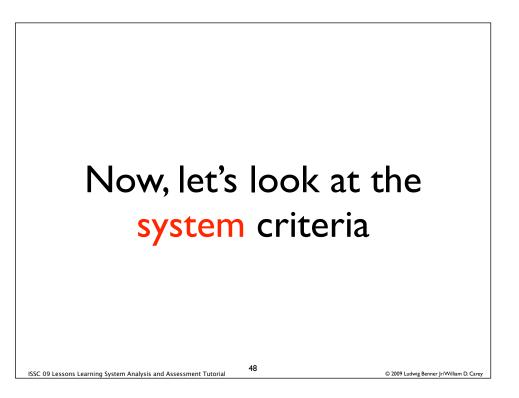


Another attribute is dynamic process compatibility. Relatively good data dictionaries and definitions of static data, such as that required by the FAA's and NASA's ASRS aviation reporting system now exist, but when describing the dynamics of an accident, we present blanks for writing unstructured narratives. Should it surprise anyone that you need an army of analysts to try to glean useful lessons learned out of such inputs? No wonder it is so difficult to develop lessons learned from such data, or to use those that are reported.

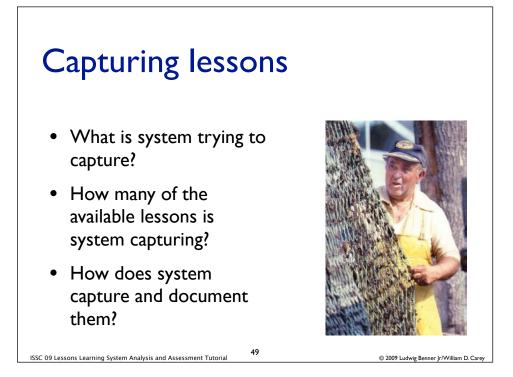
War story:

Formal reports are not much better: the Commercial Aviation and Helicopter Safety Teams had to glean and recast data from formal reports to get the information they needed to propose safety improvement actions. - volunteers spend 1/5 of their working hours doing that.





Next let's take a look at expectations for the lessons learning system...

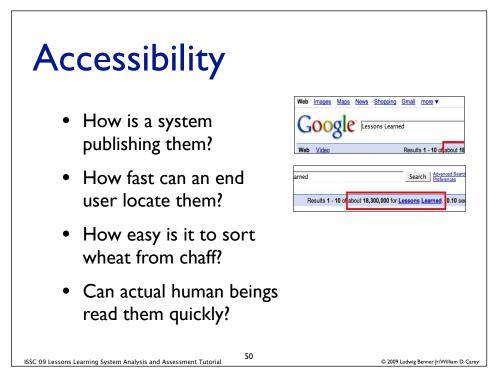


How and from what perspective has what the system is trying to capture been identified and defined?

Have the perspectives and goals been documented?

Is the system capturing all lessons generated by a mishap? Does system provide a way to determine that? (If this isn't possible, is there some benchmarking of how many?)

How concretely does system document lessons it has captured? Are lessons documented persuasively? Are the lessons actionable?



You might think accessibility is about making what you capture available. It's not. That's producers' perspective.

It's much more, from an individual user's perspective.

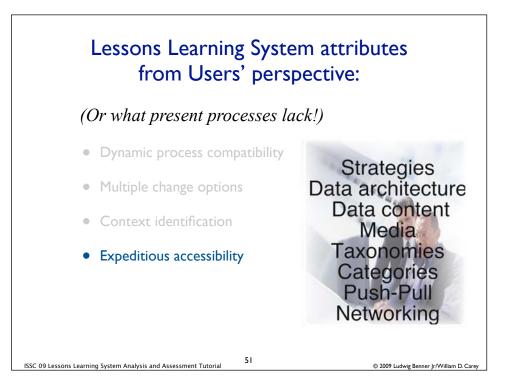
Users must be able to locate a source of relevant lessons quickly and easily. Are they being published so users can do that?

It can't take them forever to do that.

When users find a source, they've got to be able to filter lessons quickly to find those that will help them.

All this should require minimal read time for user – they have other things to do, too.

Demands high data density lessons learned data



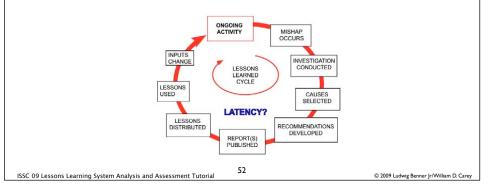
A major obstacle to use of LL in present processes is potential users' difficulty in accessing the lessons.

Locating and accessing LL is a challenge due to strategies chosen, data architecture, media, taxonomies, and other choices.

Data obsolescence and backward compatibility are additional concerns that must be taken into account while designing access to a lessons learning system.

Latency

- What's the actual system latency?
- How rapidly are the lessons disseminated
- How rapidly can the lessons be implemented?



How much time elapses between the generation of the lessons when the mishap occurs, and the changes they are intended to achieve? An important attribute of present

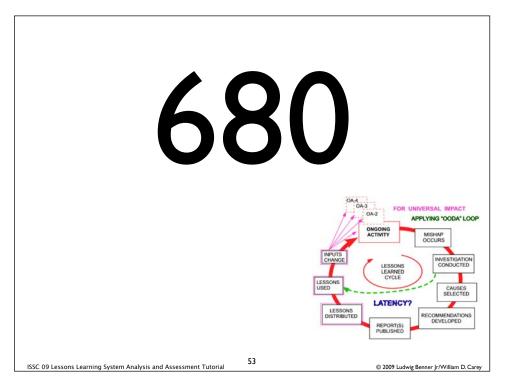
That's the latency.

Once identified, how rapidly are the lessons launched on their way to users?

Once accessed, how rapidly can the lessons be implemented by changing what people or objects do?

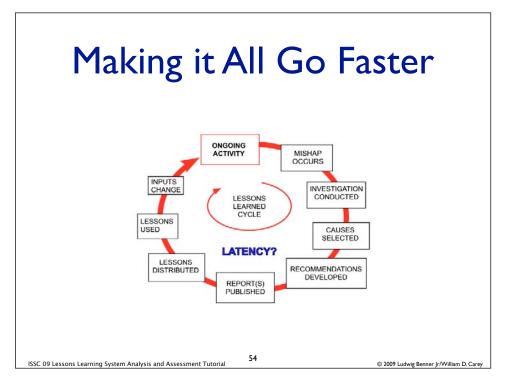
How quickly can operations be restarted with lessons implemented?

Totally quantifiable.



One sample in 2008 of 20 reports from a major investigation organization had a 680 day average latency period. A recent descriptive preliminary incident report by another organization had a 41 day latency period. Could even that be improved?

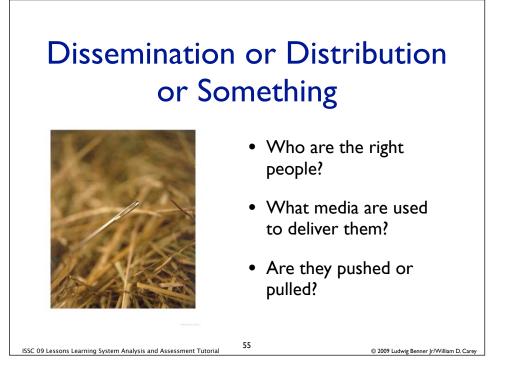
Learning system design could benefit from application of Boyd's OODA loop concepts. We had some thoughts, like bypassing analysts' functions by changing investigation and reporting of LL could dramatically reduce latency periods. This sounds like heresy, but we think its potential makes the possibility worth pursuing.



An important attribute of present Lessons Learned systems is the latency of the Lessons Learned from mishaps.

Latency is the term used to describe the period between the moment something is initiated and the moment its effects begin or become discernible.

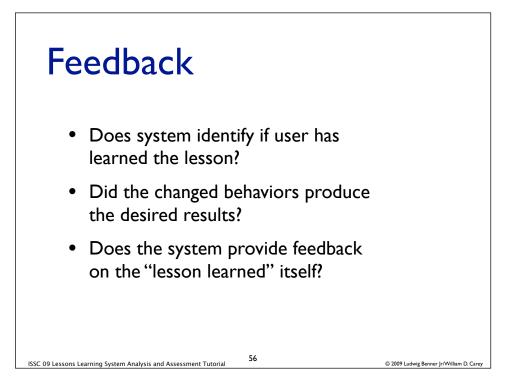
In mishaps, Lessons Learned are generated by the mishap. When a mishap occurs, it generates the data needed for Lessons Learned to be derived from the mishap. However, the **elapsed time** between the mishap and the time actual changes, indicated by the Lessons Learned, are accomplished can range from an almost instantaneous reaction, like pulling one's hand back from a hot stovetop, to years! In a sample of the 20 most recent reports published by the NTSB, for example, the latency period until the Lessons Learned were distributed (not implemented) averaged 689 days.



We've said that accessibility isn't about getting the data out there. Anyone can publish anything. 6 million kitten blogs.

Databases, Procedure Modifications, Lesson Learned Bulletins, Training Documents, Internet Repositories, CS&Rs, Checklists, Books, Studies & Research, Memories.

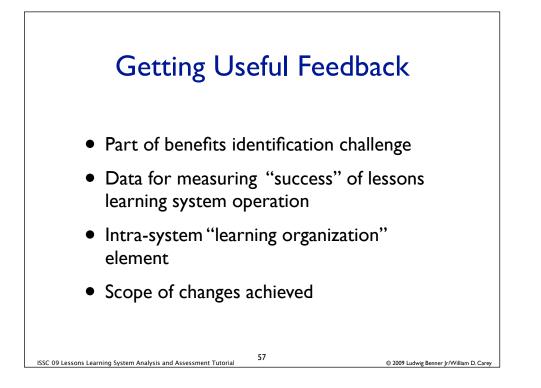
Everyone know what we mean when we talk about pushed or pulled?



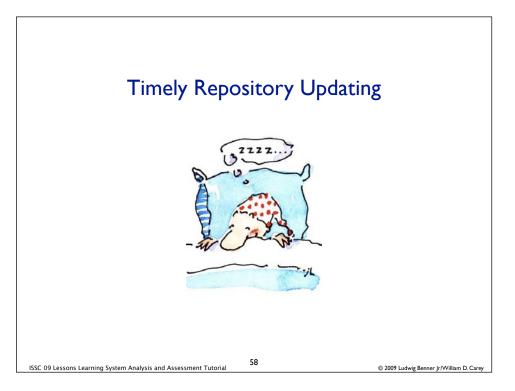
Does system provide feedback about what happened to lessons that were reported??

We're talking about validation, here – of both the assimilation of the lesson, and the efficacy of the lesson

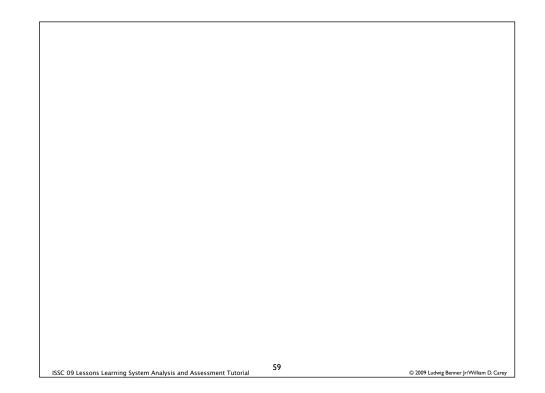
- Part of benefits identification challenge
- Data for measuring "success" of lessons learning system operation
- Intra-system "learning organization" element
- Scope of changes achieved

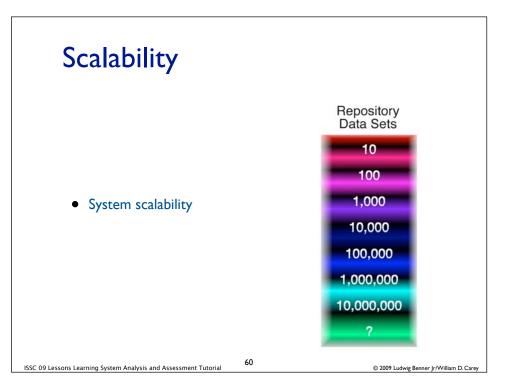


Systemic Loop Measure data access User investment in system Transition from a just-publish to a more interactive system.



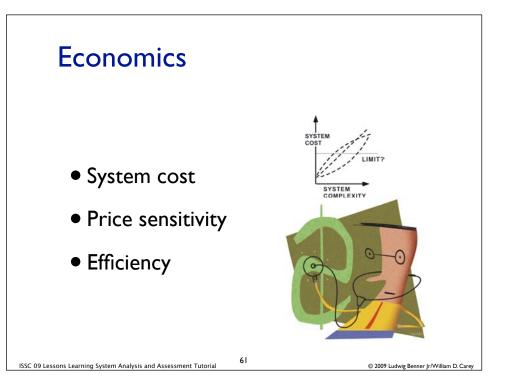
Repositories must be kept trustworthy, by purging lessons learned that didn't work or were misdefined otherwise unsuccessful, so users can sleep well after they use them.





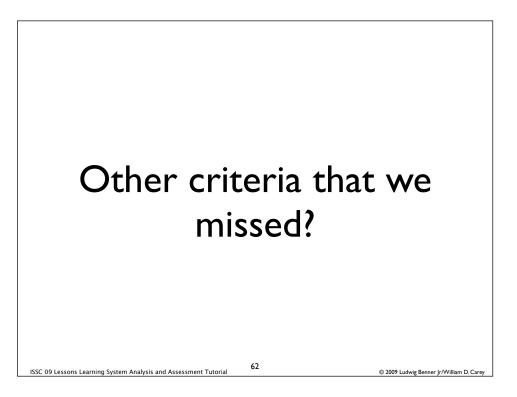
As system content grows, that growth should not sacrifice quality. Scalability needs to be designed into a lessons learning system so its growth does not discourage users from using it.

Retrieval problems with taxonomies, key words and categories suggest an alternative approach could be useful.

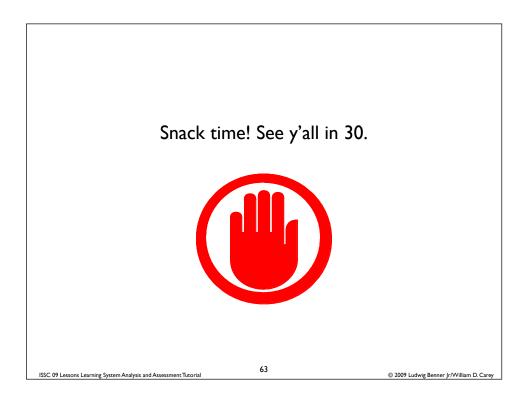


Lessons learning systems cost money. Resources devoted to lessons learning systems are not without limits. The sensitivity of price to performance of such systems must be a consideration in system design, which means maximum <u>efficiency</u> of the lesson development, dissemination and use functions is an attribute to seek in their design.

Example: NTSB cost per recommendation was almost 600K in 2008!



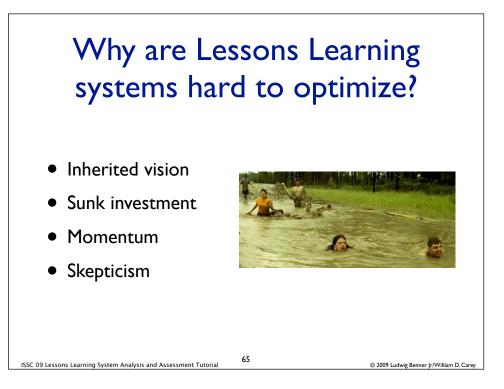
Any suggestions for us??





Lessons learned processes do not produce changed behaviors very well. Widely acknowledged underperformance of present processes, reflects deeply ingrained process attributes and design decisions. Therefore, we are going to suggest system attributes for what would be a better system for you.

Lessons Learned ≠ Changed Behavior



Basic practices have been in place a long time

Recent attempts have tinkered with existing systems, but little changed in underlying vision, assumptions, strategies, principles, design or data

WHY? Present systems conform to **inherited vision** of accidents, incidents, investigations, causes and lessons learned---a vision for a system that's better has been lacking

a lot of people and organizations have a large sunk investment in present practices that change would put at risk

status quo has a lot of momentum - comfortable doing what we know, in the groove, why bother looking for trouble

Skepticism that need exists - we're getting by ok

Why are Lessons Learning systems hard to optimize?

66

- Technical challenges
- Murky, incremental benefits
- Lack of metrics
- Fear of change

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Optimizing implies something better is available.

Technical challenges are not trivial. It's hard work to figure out what's better

Hesitancy is normal reaction when benefits of a change are ambiguous, viewed as incremental

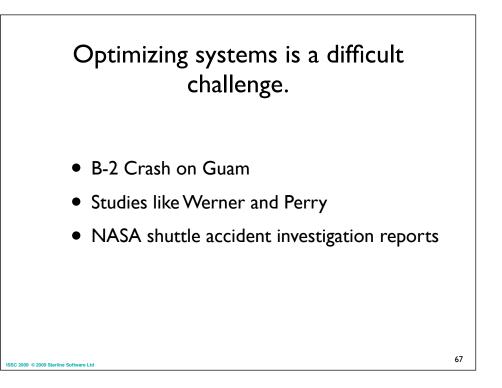
Then there's a lack of sound metrics that would reveal the poor performance of present processes

Then there is plain emotion-based fear - fear or losing job, fear or failing, fearing lack of competence, etc

What are some other constraints might you add?

To modify or supplant legacy systems, something better must be offered.

- Technical challenges
- Murky, incremental benefits
- Getting by.
- Fear of change



We've talked a bit about the B-2.

In their study, Werner and Perry cited a bunch of reasons why lessons learned are underutilized in the aviation community.

"NASA stated that it must do a better job of communicating the various lessons learned sources to employees, improving mechanisms to link these sources, and ensuring appropriate training for employees in order to maximize lessons learning." (United States General Accounting Office (2002) Report to the Subcommittee on Space and Aeronautics, Committee on Science, House of Representatives, NASA: *Better Mechanisms Needed for Sharing Lessons Learned*, GAO-02-195,)

Personal use of public lessons learned data is unknown, quantitatively, but interest in and use to generate new behaviors by individuals seems very limited. For example, one widely respected and emulated public incident lessons learned database with over 700,000 records (ASRS) had 88 search requests by individuals during a recent six year period.

How many individuals would buy a 334 page, \$80 book to find lessons learned that might apply to their tasks and then internalize all of them to change their behaviors? How frequently do individuals change their behaviors due to desired interpretations of generalized training, procedures, standards or regulations? We don't know. However, few would argue that present practices maximize investigation lessons learned dissemination and their use by all who might benefit from the data.

Natural Language Difficulties



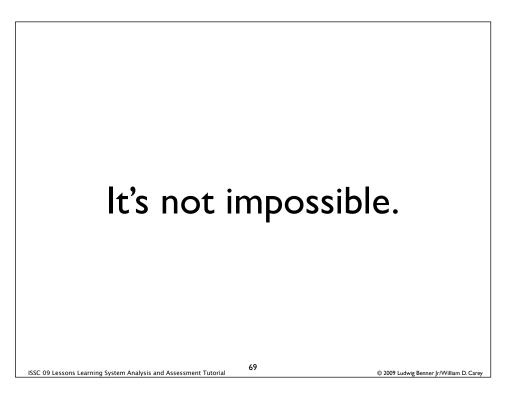
Natural language is a blessing and a curse. It can be very expressive and communicate meaning without being concrete. It can also frustrate the description of people, objects and energies and their actions because of its propensity to use and tolerate **ambiguity**, the great variety of ways it is possible to express something, **value-laden** vocabulary, and flexibility of grammar and syntax, for example.

Know about Hayakawa's ladder of **abstraction**, and the ambiguity introduced as objects rise on the ladder; abstractions can be "cover up" words chosen to cover up lack of specificity of understanding, and thus pose barriers to **objective** understanding of phenomena, **objectively** describing and explaining them, identifying lessons learned, and communicating that understanding and the lessons learned.

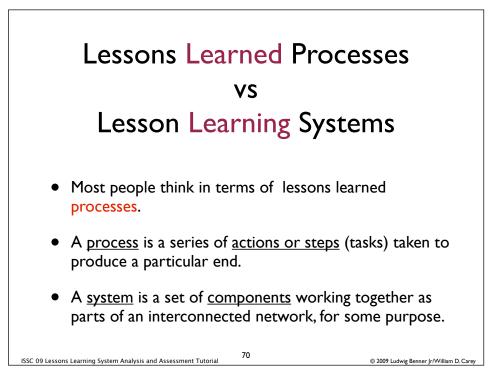
Descriptions can be enhanced by the definitive vocabulary, grammar, and structure.

The challenge is to work abstraction, ambiguity and subjectivity out of Lessons Learned systems.

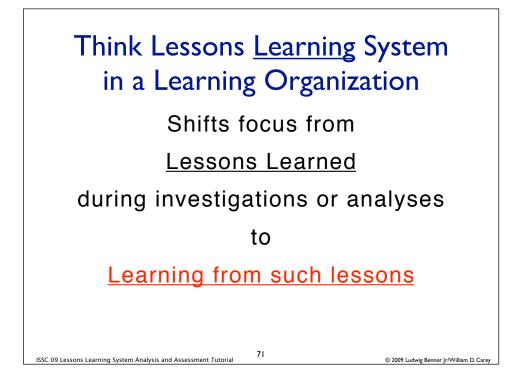
And never use the passive voice. Ever. Weasel wording.



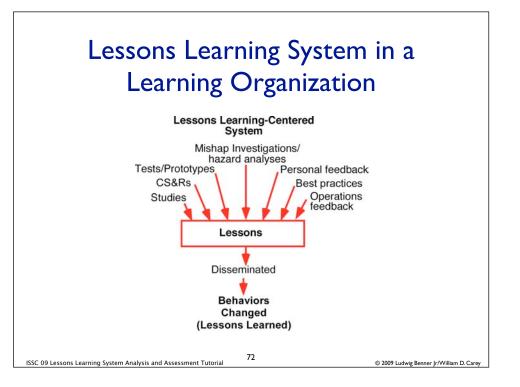
It may be hard, but not impossible. Our research has disclosed promising ideas.



We' re system safety people here – let's start by talking about a lessons learning system.

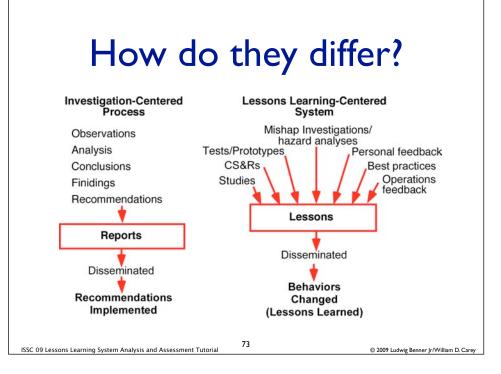


First lets shift the arena for discussion.



Viewed this way, focus of investigation is to provide the new knowledge that will enable a learning organization to bring about changed behaviors, rather than determination of causes, root causes, probable causes, proximate causes, remote causes, or cause factors, or hazards or recommendations.

Lessons learned from each source must be compatible to support a true Learning Organization's needs



	Old School	New School
Where's the focus of the system?	causes	description of behavior
Whose needs come first?	investigators	users
Who defines the lessons?	analysts	investigators
What do you report?	recommendations	lessons
How accessible are the lessons?	limited	universal
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Strategic Choices!

Focus on determining causes <u>or</u> on descriptions of behavioral interactions of people, objects, or energies that users can change to achieve improved performance.

Limited accessibility or universally accessible lessons for all who could benefit from their assimilation.

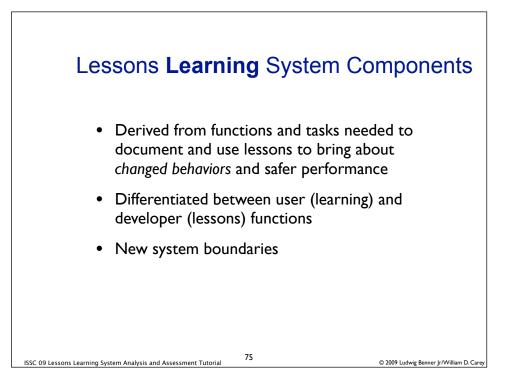
Focus on causes or descriptions of behavioral interactions?

Design to investigator or user needs?

Analysts or investigators define lessons?

Report recommendations or lessons?

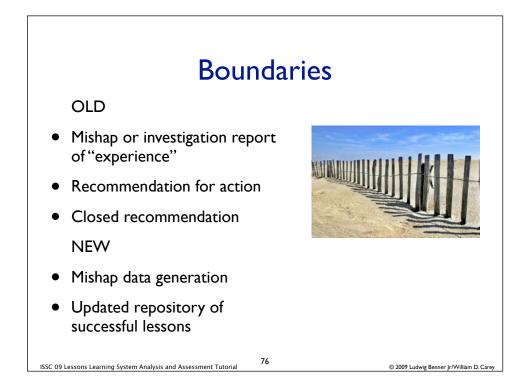
Limited or universal accessibility?



We analyzed the functions and tasks needed to covert data generated by an accident or mishap into outputs that would produced changed behaviors AND safer performance.

That led us to new system boundaries.

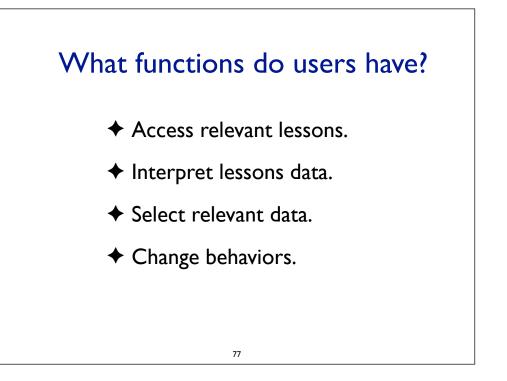
We also found it imperative to separate the functions of users who were the "learners" in the system from the "developers" who produced the lessons to be learned by the users.



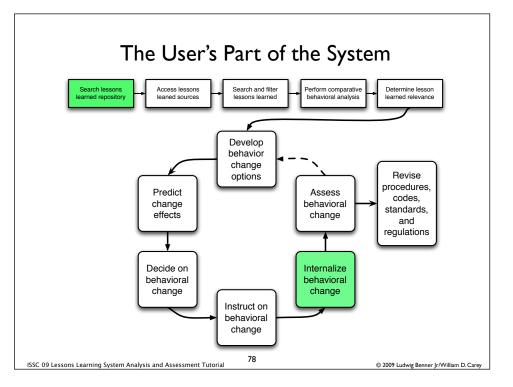
What do we mean by boundaries? What components does system include?

Old boundaries

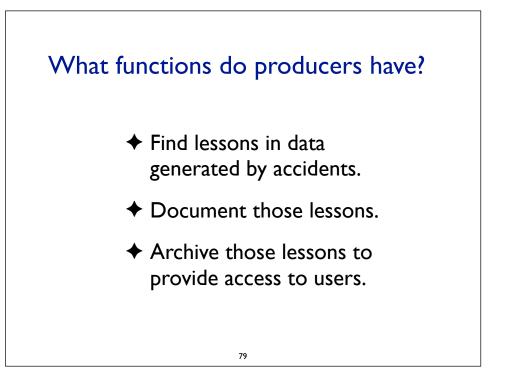
We suggest new boundaries – one – accident or mishap that generates new data from which lessons are extracted two – updated repository of successful lessons lessons learning learning system is everything in between.



Users must access lessons, interpret them for relevance and applicability, select relevant lessons, figure out what changes are needed to respond to lessons, and then produce the changed behavior needed

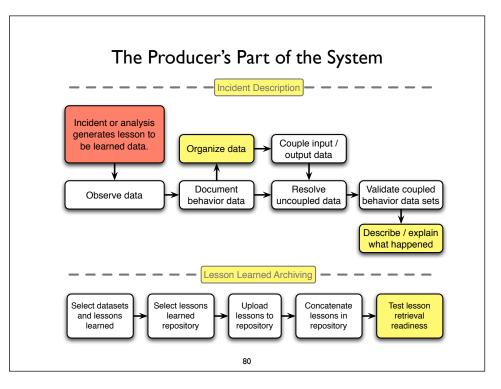


Let's look at user or learning part of system



We found it helpful to distinguish between the finding and documentation the lessons and the subsequent "archiving" functions involved in making the documented lessons accessible and assimilable for users.

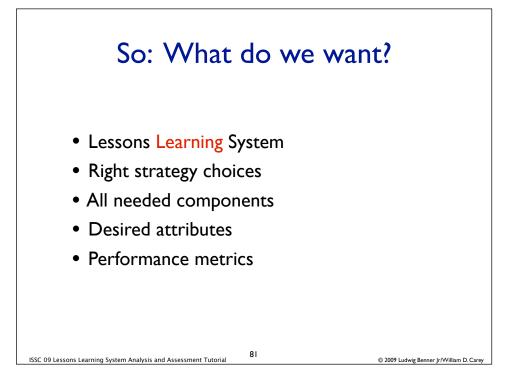
Investigation functions are needed to develop LTBL data and document all lessons Archiving functions are needed to make LL easily accessible and assimilable for users

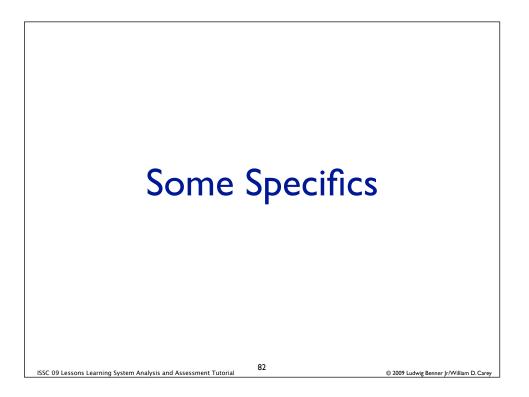


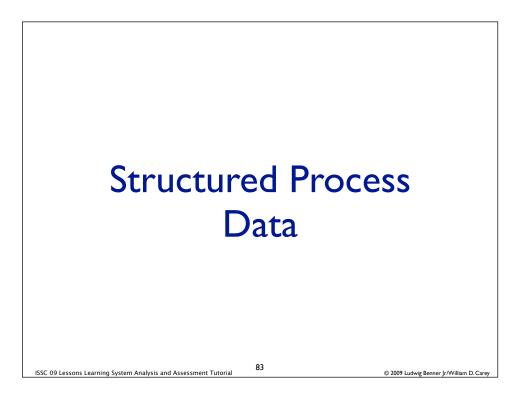
Lesson developers also have a lot to do. This developer part of the investigation learning system model reflects several strategic choices by us, based on previously reported work. For this model, we chose to

define the "lesson learned" as a description of what happened during the accident process,

document the lessons as coupled behavior sets in order to do that







VMC Cice daylight	non-ra superv military	yrs. adaryrs. visoryyrs. yyrs. ATC/ADVISORY SERV.
VMC ice daylight	night	
	-	local center
mixed turbulence ceiling marginal tstorm visibility rain windshear RVR	miles	ground FSS apch UNICOM dep CTAF Name of ATC Facility:
ł	AIRCRAFT	Т 2
EFIS (Other Aircraft)		EFIS
orporate air carrier her commuter		

Not how well the static data requested by the form is defined for the user.



Mishaps are dynamic processes.

This is the guidance for describing dynamic occurrence.

So what? Ambiguous, abstract, subjective, weasel wording.



Programmers don't want a narrative explanation of what went wrong. They want a stack trace.

Programmers have developed language (and code) to capture and record "mishaps".

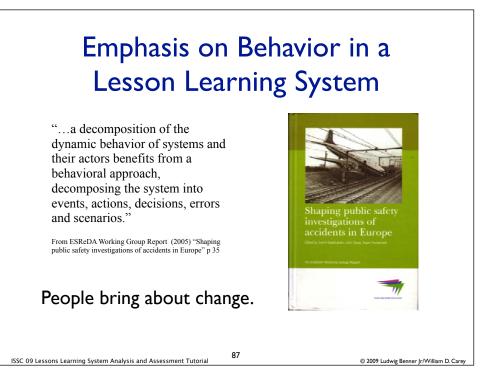
What data? Why? Formats and structure?

Define, Identify, Document, Validate?

Structural language vs dictionary approach - dictionary fine for objects and their description

Static data relatively easy to define, and being done.

Process data a little trickier.

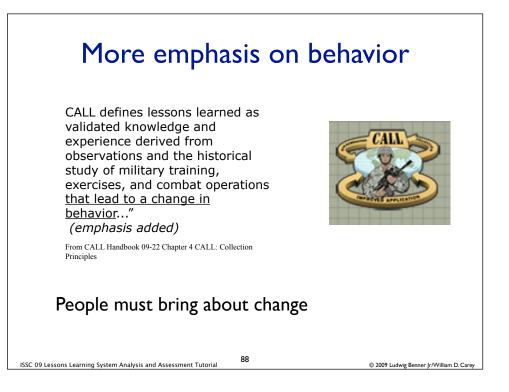


Focus on behaviors - who does what, when, where?

Note the third line -

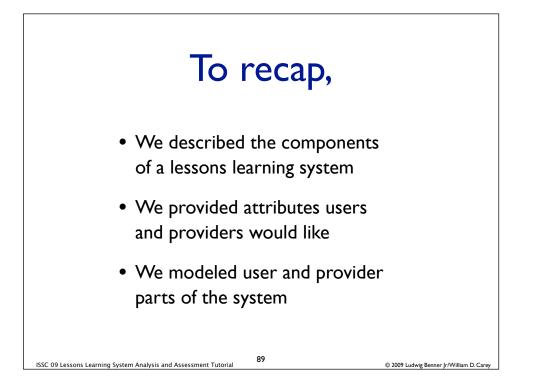
The focus on behaviors reported in our paper is not particularly original - the ESReDA working group recognized the benefits of the behavioral approach to decomposing the system in 2005.

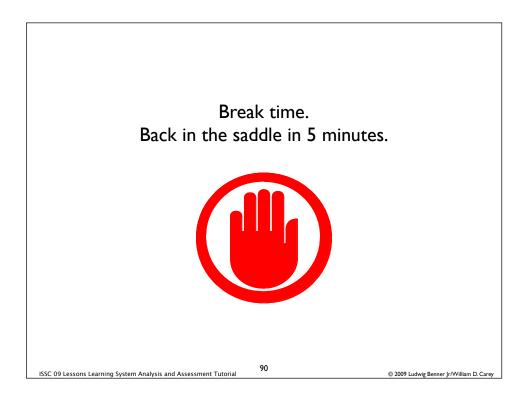
Our experiences support the benefits envisioned.

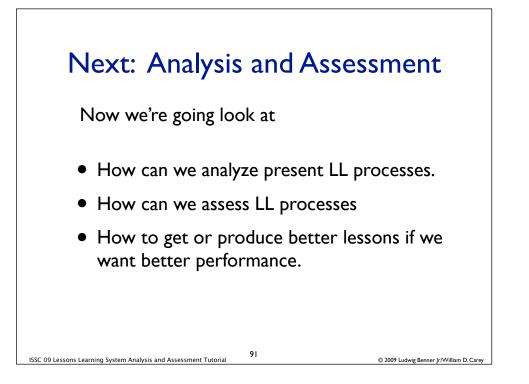


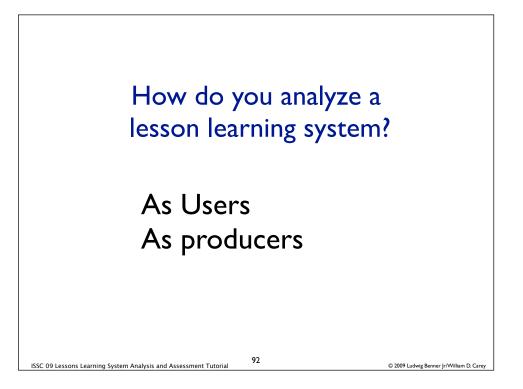
Focus on behaviors - who does what, when, where? Army CALL says behavior change is goal.

Consistency



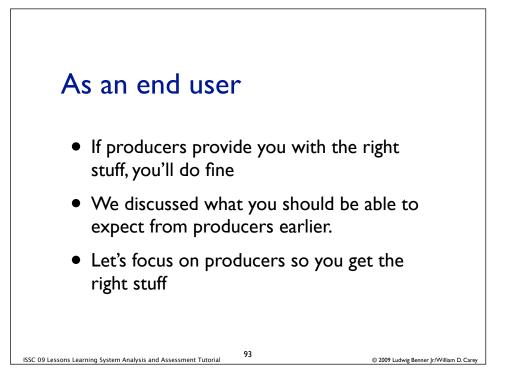




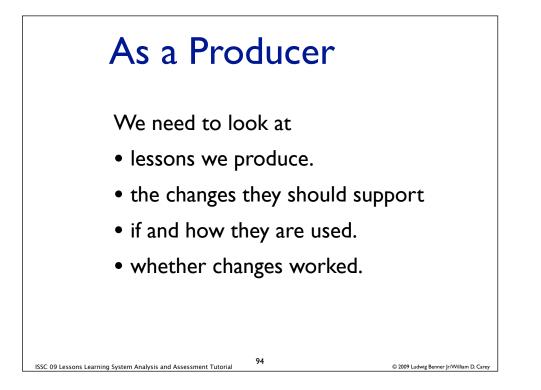


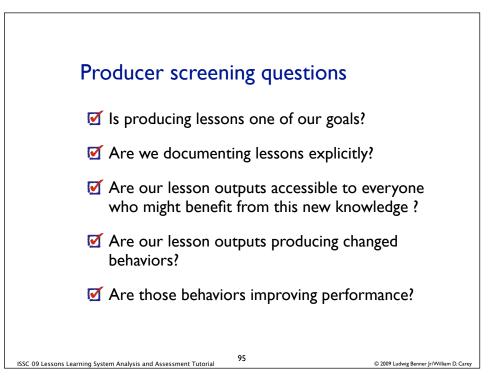
We need to analyze system as users or producers of lessons learned from investigations and analyses. How can you tell if your system is satisfactory. Let's start with criteria for an optimal Lessons Learned System, from User's or Customer's perspective

What attributes are worth measuring and monitoring to help you decide?

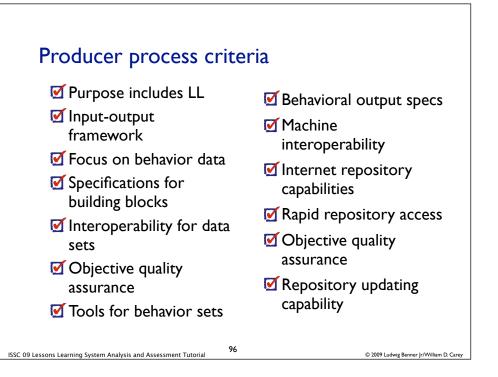


Let's keep user's analysis short.

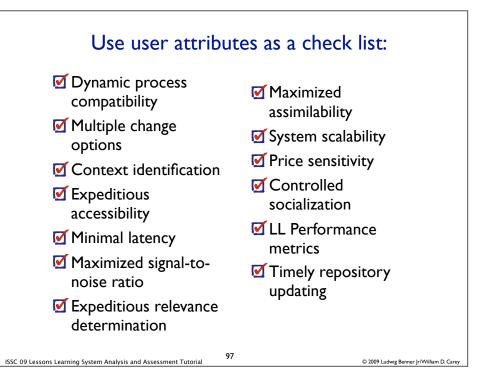




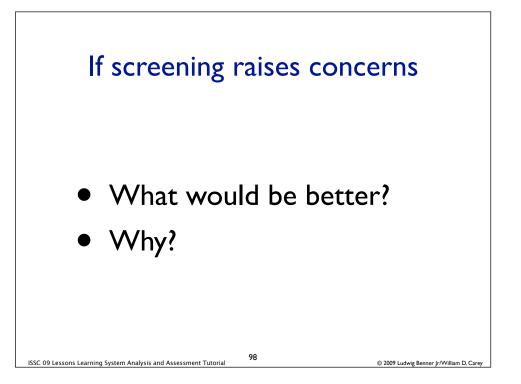
here's a quick check list we offer to get an overview of how things are going



Here's a quick check list of desired process attributes we would offer to anyone who wants to review an investigation-based lessons learning system

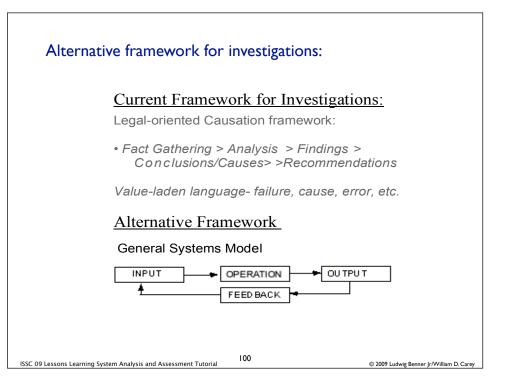


If you are a user, here's a quick check list we developed to help determine if lessons learned process is providing useful new knowledge...





We tried to Re-think how things were being done, and how that might be improved, and we're going to share what we came up with next.



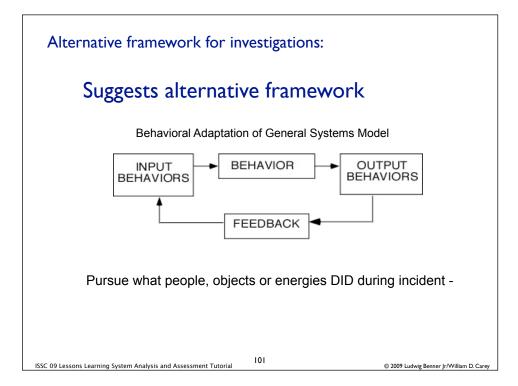
What might be an alternative to the current **inherited** investigation framework of accident causation models?

We tried the General Systems Model

Could we use it for a new framework for thinking about investigations - by adapting it as an investigation process framework?

What would it do?

- •Well, it seemed to fit a learning organization better
- •it can change how we think about investigations to I/O framework
- •it could potentially minimize value-laden language,

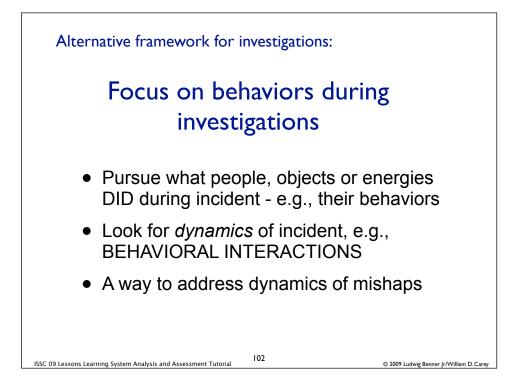


But if we want to focus on behaviors, how might we describe our framework?

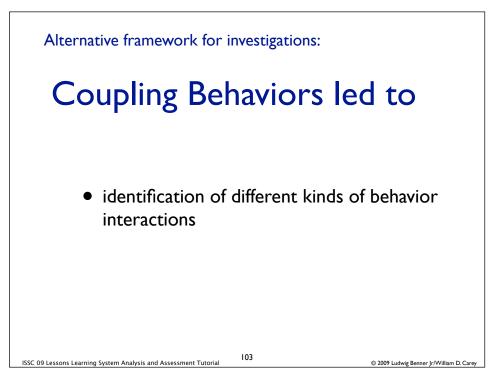
Well, here's what we came up with.

How would that change investigations? Changes what investigators look for

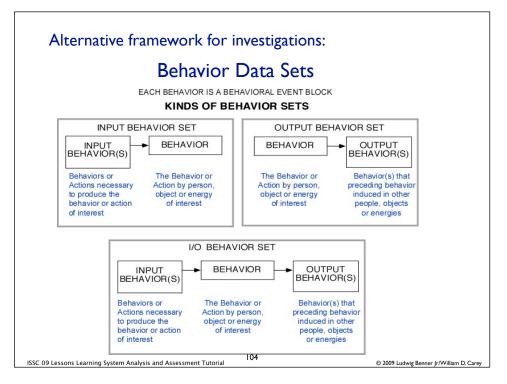
Look for dynamics of incident, e.g., BEHAVIORAL INTERACTIONS or COUPLED BEHAVIOR SETS



With the modified general systems model, we have a way to address the DYNAMICS of accident during investigations.

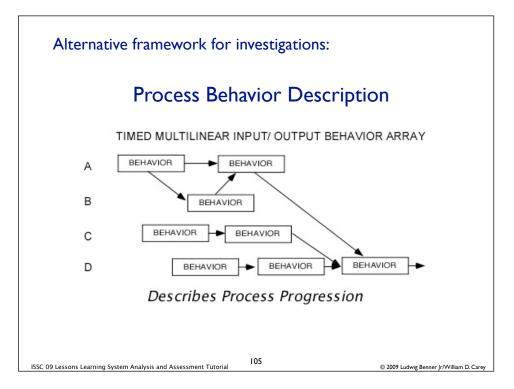


When we started coupling behaviors we discovered we could get different kinds of behavior interactions or sets



Here's a look at some of the different kinds of behavioral interactions this approach might provide for us. Don't have to study – but this is what we went through

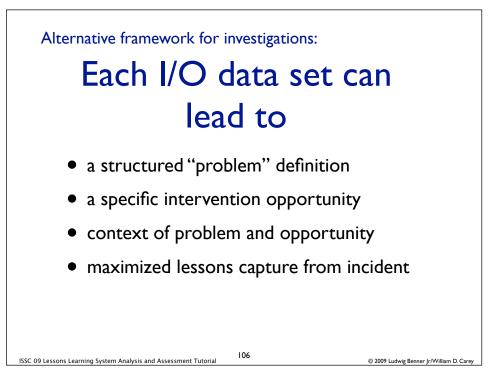
(Step thru sections)



How do we organize these behaviors after we capture them?

Well, one way is to set up data on a matrix so you can follow who did what,

But also what behaviors they influenced, by showing arrows to represent interactions. This is the behavior coupling step – establishing interactions among everyone and every thing involved in producing the eventual outcome.



now, if we do it this way, look at what it could do for us.

Each I/O data set constitutes a "problem definition" Think of this way.

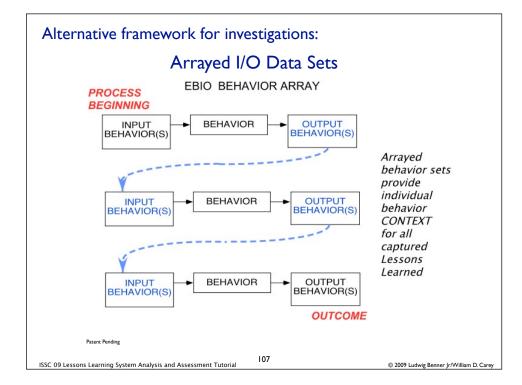
if the interactions had not existed, the accident would not have progressed the way it did. So if you eliminate a set, there would have been a different outcome.

But gee, the accident happened. How about if you eliminate that behavior interaction from future operations? That would reduce likelihood of similar accidents where that set would have to be present.

And when you couple behaviors before and after the set you are working with, you have indications of the context of the problem behavior set, and of the opportunity changing it might present.

And lastly, you capture ALL the behavioral lessons from the accident.

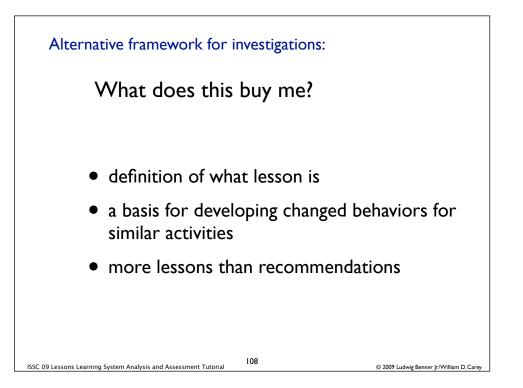
Not too shabby.



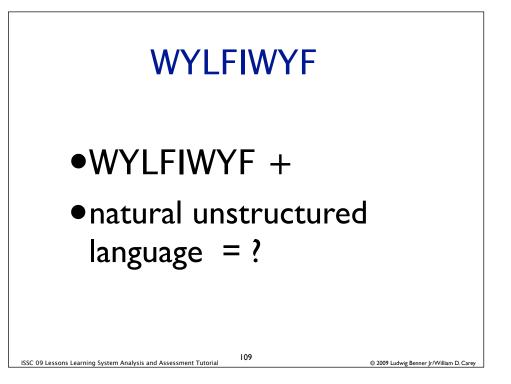
How investigators organize their investigation data affects the story they tell. Often it's put into logic trees, or EC&F charts. Not bad, as far as it goes, but because they use unstructured data, couplings are not very compelling.

Here's what we came up with to show the dynamics of the behavioral interactions, as part of a continuous series of changes that led to the outcome.

Can you see how this "flows" the behavioral inputs, actions and outputs??



- New knowledge is understanding of behavioral interactions in form of coupled input/output behavior sets
- Behavioral interaction sets provide basis for developing changed behaviors for similar activities
- Enables broader use than recommendations



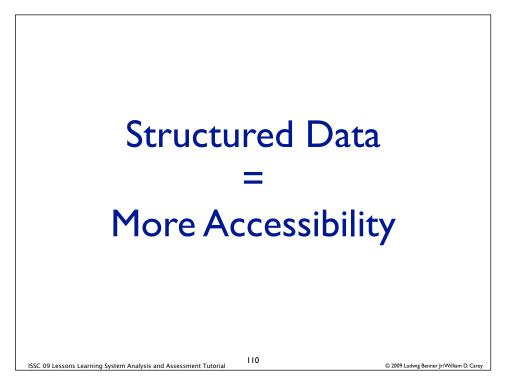
We have observed that what investigators look for during investigations depends heavily on their perspective of what an accident is. That's called WYLFIWYF (Wilfi wif) And then findings are reported in natural language.

results: today's performance.

how about hazard analyses ...? Simiilar problems???

- View accidents as processes generating lessons-to-be-learned data for investigators
- Investigate to gain understanding of behaviors that produced the outcomes
- Structure inputs and outputs to guide investigations?
- Here are some ideas we think will help overcome that for

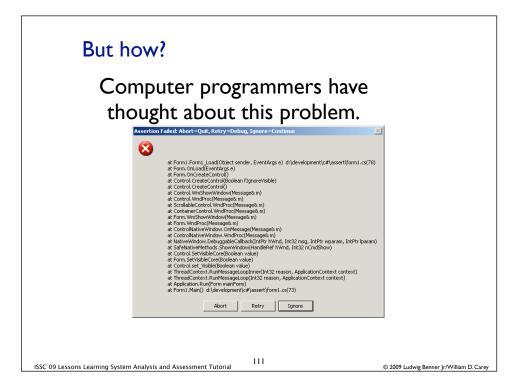
investigations



But only for the static data. ASRS does this well.

For dynamic data? Harder? Ham-fisted right now

How do you solve that problem?

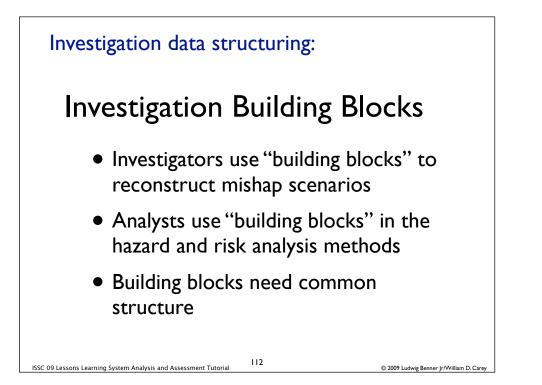


As we mentioned,

Programmers don't want a narrative explanation of what went wrong. They want a stack trace.

Programmers have developed language (and code) to capture and record "mishaps".

So we tried to learn from them...



we started with the investigation input data, which leads to building blocks.

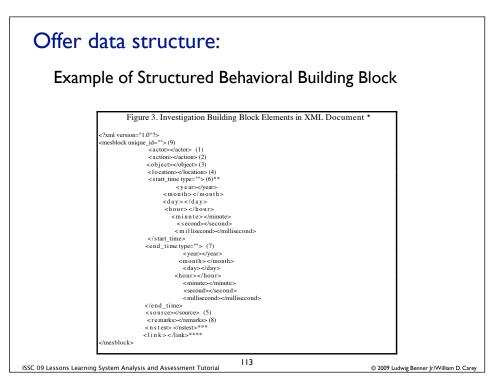
So, we structured the building blocks to require structured input data,

to provide guidance for investigators and analysts

to facilitate machine interoperability for data

to enhance search and retrieval of lessons

to optimize assimilation of lessons



Here's an example of a well defined structure for a behavioral building block using data tags (This is in use)

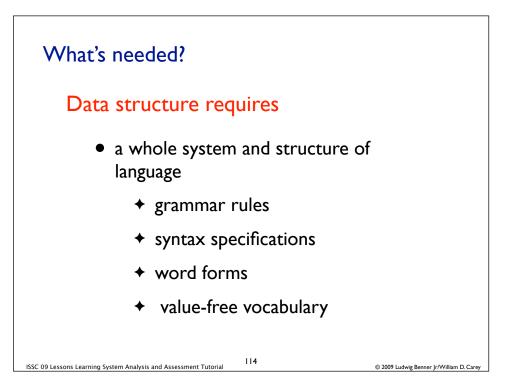
Times permit machine ordering of behaviors

Enables investigators to couple behavioral inputs and outputs

Structure helps reduce natural language problems of ambiguity, abstraction and inconsistency

Readily expandable to accommodate addition data elements if needed

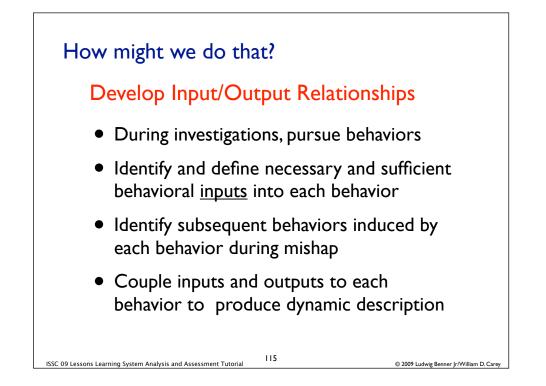
enhances machine interoperability



to optimize consistency to optimize machine interoperability to reduce conflict following investigations

We think we really need to change investigation vocabulary to value-free, nonjudgmental, words - eliminate accusatory or pejorative words, like cause, fault, failed to, error, failure,

Judgments about what happened are the province of different societal entities, not SS or AI

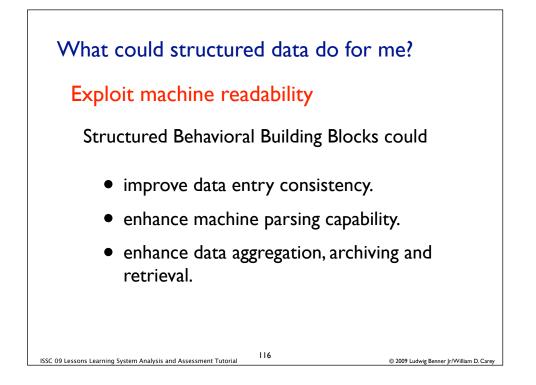


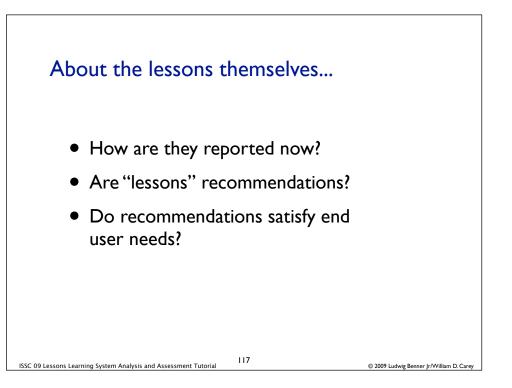
* The coupling of the inputs, behaviors and outputs produces behavioral sets which can be manipulated by machine. Those sets can provide context for lessons learned.

This lets you do a more readily defined search.

Capture the static attributes related to the incident, as in the ASRS system, for statistical analyses and potential search keys.

Capture the dynamics with coupled structured behavior data elements.





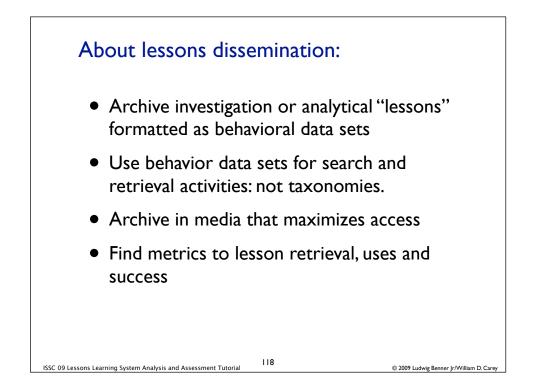
Remember our earlier discussion?

We realized that recommendations which are common investigation outputs are not lessons but FIXES for problems defined by the investigators.

So we thought it would be a good idea to call lessons lessons, not recommendations, factors, causes, failures, etc

we wondered why we don't see lessons listed as lessons

That's when we started thinking old school vs new school again

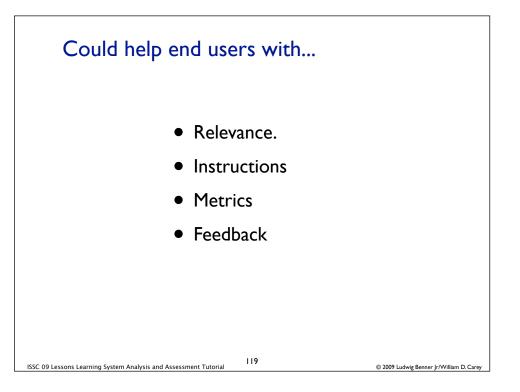


The answers weren't obvious to us at first, but gradually some ideas took hold.

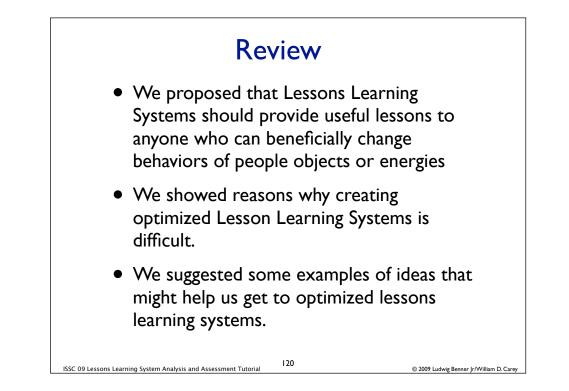
If we used behavior sets as the lessons, that's what we could archive, and it could improve search and find capabilities because of their machine interoperability characteristics – no taxonomies to second guess.

New media exist to maximize unfiltered, uninterpreted access -the internet,

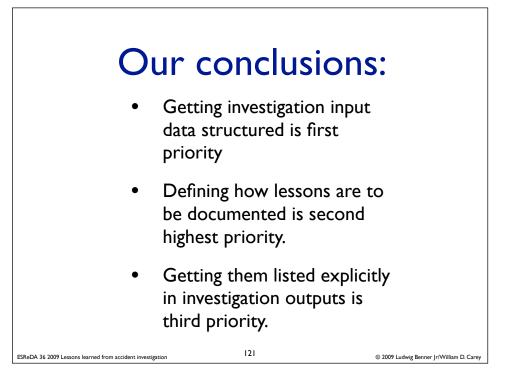
Behavior sets could be observed, and thus monitored before and after they are changed.



We think I/O behavior sets could definitely help users with certain tasks they are faced with when an incident reveals sets they have in their operations. Those tasks include lessons relevance determination, converting those lessons into task changes in their organizations, developing metrics to assess lessons and changes made, and a format for offering feedback to lesson suppliers.



SS practitioners can demonstrate a way to go with their own outputs. Start by cleaning up fault trees we create by using structured input behavioral data. Do parallel outputs if necessary to test ideas, and provide basis for comparison of results.



Here are the priorities we think are in order to develop optimized lessons learning system performance.

To that end,

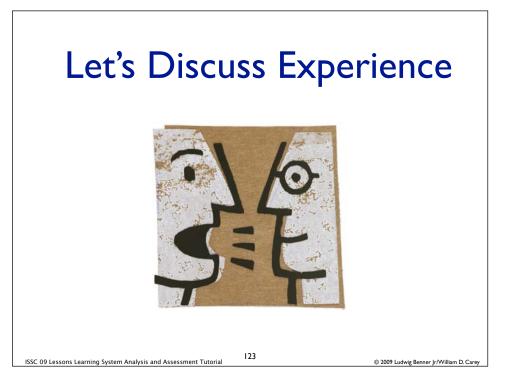
Open Source (LGPL) Library:			
http://code.google.com/p/meslib/			
Complete OS X sample app in Objective C, early alpha development library in platform independent C++. Some sample PHP for online stuff too.			
Contribute your ideas!			
Contact:	luben@starlinesw billcarey@mac.co		
DA 36 2009 Lessons learned from accident investigation	122	© 2009 Ludwig Benner Jr/William D. Carey	

We believe very strongly in our approach, and the potential for evolutionary development of improved systems. So strongly that we are making publicly available on line an Open Source Library of software we developed, to launch the first steps toward needed changes. The Software Library includes a royalty-free license for use by anyone who wants to redesign their investigation data inputs and lessons documentation to support lessons learning system improvements.

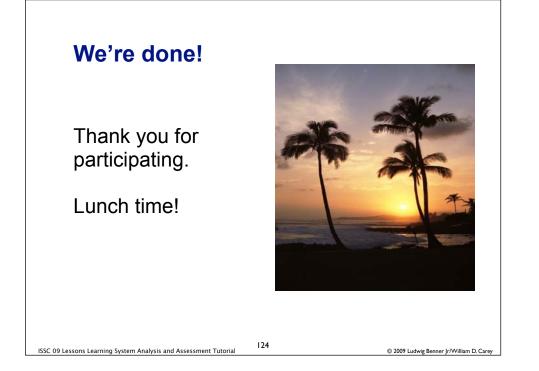
Complete OS X sample app in Objective C, early development library in platform independent C++. Some sample PHP for online stuff too.

(Library GPL is the license) (work in progress) (C++, Objective C)

ESRe



Free discussion for about 10 minutes, plus more after session if desired.



Open Source	(LGPL) Libı	rary:	
http://code.google.com/p/meslib/			
Complete OS X sample app in Objective C, early alpha development library in platform independent C++. Some sample PHP for online stuff too.			
Contribute your ideas!			
(See handout for details)			
Contact:	luben@starlinesw.cor billcarey@mac.com	n	
ESReDA 36 2009 Lessons learned from accident investigation	125	© 2009 Ludwig Benner Jr/William D. Carey	

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