

#### STRUCTURE AT A GLANCE

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Appendix: Past, Present, and Future

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Part VII – Failures: Computational Breaches



### **About This Presentation**

This presentation is intended to support the use of the textbook *Dependable Computing: A Multilevel Approach* (traditional print or on-line open publication, TBD). It is updated regularly by the author as part of his teaching of the graduate course ECE 257A, Fault-Tolerant Computing, at Univ. of California, Santa Barbara. Instructors can use these slides freely in classroom teaching or for other educational purposes. Unauthorized uses, including distribution for profit, are strictly prohibited. © Behrooz Parhami

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First	Sep. 2006	Oct. 2007	Dec. 2009	Nov. 2012	Nov. 2013
		Mar. 2015	Nov. 2015	Nov. 2018	Nov. 2019
		Nov. 2020			



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# **25 Failure Confinement**





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MARKETING FAILURE #137: GLASS PIÑATAS

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"Remember son, if at first you don't succeed,

assemble a team of lawyers to shield all involved, convince your investors that failure is an essential component of success, create a network for spin control to trivialize any negative consequences, and try, try again."

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Appendix: Past, Present, and Future



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# 25.1 From Failure to Disaster

• Computers are components in larger technical or societal systems

• Failure detection and manual back-up system can prevent disaster

Used routinely in safety-critical systems: Manual control/override in jetliners Ground-based control for spacecraft Manual bypass in nuclear reactors

Manual back-up and bypass systems provide a buffer between the failed state and potential disaster

Not just for safety-critical systems: Amtrak lost ticketing capability on Friday, Nov. 30, 1996, (Thanksgiving weekend) due to a communication system failure and had no up-to-date fare information in train stations to issue tickets manually

Manual system infeasible for e-commerce sites



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### Space Shuttle Challenger Disaster



Second shuttle after Columbia, first mission on April 4, 1983 Exploded 73 seconds after launch on its 10th mission (January 28, 1986) Seven crew members killed

Temperatures dipped below freezing on launch day

Engineers were concerned about the integrity of seals on the booster

Aftermath:

Presidential commission formed to look into the incident

Cultural problems at NASA, including ineffective communication channels Major changes enacted at NASA as a result

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# 25.2 Failure Awareness

Importance of collecting experimental failure data • Indicate where effort is most needed

• Help with verification of analytic models

System outage stats (%)*	Hardware	Software	Operations	Environment
Bellcore [Ali86]	26	30	44	
Tandem [Gray87]	22	49	15	14
Northern Telecom	19	19	33	28
Japanese Commercial	36	40	11	13
Mainframe users	47	21	16	16
Overall average	30	32	24	14

\*Excluding scheduled maintenance

### Tandem unscheduled outages

Power	53%
Communication lines	22%
Application software	10%
File system	10%
Hardware	5%

Tandem outages due to hardware

49%
24%
18%
9%
1%

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## System Failure Data Repositories

Usenix Computer Failure Data Repository

http://usenix.org/cfdr

"The computer failure data repository (CDFR) aims at accelerating research on system reliability by filling the nearly empty collection of public data with detailed failure data from a variety of large production systems. ...

You first need to register for full access to CFDR data/tools to obtain a user id/password. You can then go to the data overview page."

LANL data, collected 1996-2005: SMPs, Clusters, NUMAs <u>http://institute.lanl.gov/data/fdata/</u>

From the site's FAQs: "A failure record contains the time when the failure started (start time), the time when it was resolved (end time), the system and node affected, the type of workload running on the node and the root cause."





## Memory and Storage Failure Data

Storage failure data: "Disk Failures in the Real World: What does an MTTF of 1,000,000 hours mean to you?" (Schroeder & Gibson, CMU) <u>http://www.cs.cmu.edu/~bianca/fast07.pdf</u> From the abstract: "... field replacement is a fairly different process than one might predict based on datasheet MTTF."



OBSERVATION 1. Variance between datasheet MTTF and disk replacement rates in the field was larger than we expected. The weighted average ARR was 3.4 times larger than 0.88%, corresponding to a datasheet MTTF of 1,000,000 Hours. [Schr07]

Rochester Memory Hardware Error Research Project <u>http://www.cs.rochester.edu/research/os/memerror/</u> "Our research focuses on the characteristics of memory hardware errors and their implications on software systems." Both soft errors and hard errors are considered.

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### Software Failure Data

Promise Software Engineering Repository http://promise.site.uottawa.ca/SERepository/ "Here you will find a collection of publicly available datasets and tools to serve researchers [who build] predictive software models (PSMs) and software engineering community at large."

Software Forensics Centre failure data, Middlesex University <a href="http://www.cs.mdx.ac.uk/research/SFC/">http://www.cs.mdx.ac.uk/research/SFC/</a>

From the website: "The repository of failures is the largest of its kind in the world and has specific details of well over 300 projects (with links to another 2,000 cases)."

[Note: As of November 27, 2013, the large repository seems to have disappeared from the site.]





## Failure Data Used to Validate or Tune Models

- Indicate accuracy of model predictions (compare multiple models?)
- Help in fine-tuning of models to better match the observed behavior

**Example:** Two disks, each with MTTF = 50K hr (5.7 yr), MTTR = 5 hr Disk pair failure rate  $\approx 2\lambda^2/\mu$ Disk pair MTTF  $\approx \mu/(2\lambda^2)$ 

 $= 2.5 \times 10^8$  hr = 28,500 yr



In 48,000 years of observation (2 years  $\times$  6000 systems  $\times$  4 disk pairs), 35 double disk failures were reported  $\Rightarrow$  MTTF  $\approx$  1400 yr

Problems with experimental failure data:

- Difficult to collect, while ensuring uniform operating conditions
- Logs may not be complete or accurate (the embarrassment factor)
- Assigning a cause to each failure not an easy task
- Even after collection, vendors may not be willing to share data
- Impossible to do for one-of-a-kind or very limited systems

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# 25.3 Failure and Risk Assessment

- Minimum requirement: accurate estimation of failure probability
- Putting in place procedures for dealing with failures when they occur

Failure probability = Unreliability	
Reliability models are by nature pessimistic (provide lower bounds)	
However, we do not want them to be too pessimistic	

Risk	=	Frequency	×	Magnitude
Consequence / Unit time		Events / Unit time		Consequence / Event

Frequency may be equated with unreliability or failure probability Magnitude is estimated via economic analysis (next slide)

Failure handling is often the most neglected part of the process An important beginning: clean, unambiguous messages to operator/user Listing the options and urgency of various actions is a good idea Two way system-user communication (adding user feedback) helpful Quality of failure handling affects the "Magnitude" term in risk equation



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## How Much Is Your Life Worth to You?

### **Thought experiment:**

You are told that you have a 1/10,000 chance of dying today

How much would you be willing to pay to buy out this risk, assuming that you're not limited by current assets (can use future earnings too)?

If your answer is \$1000, then your life is worth \$10M to you

Risk	=	Frequency	×	Magnitude
Consequence / Unit time		Events / Unit time		Consequence / Event

Can visualize the risk by imagining that 10,000 people in a stadium are told that one will be killed unless they collectively pay a certain sum

Consciously made tradeoffs in the face of well-understood risks (salary demanded for certain types of work, willingness to buy smoke detector) has been used to quantify the worth of a "statistical human life"



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## Very Small Probabilities: The Human Factor

Interpretation of data, understanding of probabilities, acceptance of risk

K
5K
8K
Μ
0B

Factors that increase risk of death by 1/10<sup>6</sup> (deemed acceptable risk)

Smoking 1.4 cigarettes Drinking 0.5 liter of wine Biking 10 miles Driving 300 miles Flying 1000 miles Taking a chest X-ray Eating 100 steaks

US causes of death	/ 10 <sup>6</sup> persons
Auto accident	210
Work accident	150
Homicide	93
Fall	74
Drowning	37
Fire	30
Poisoning	17
Civil aviation	0.8
Tornado	0.4
Bite / sting	0.2

Risk underestimation factors: Familiarity, being part of our job, remoteness in time or space **Risk overestimation factors:** Scale (1000s killed), proximity



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# 25.4 Limiting the Damage

Prompt failure detection is a prerequisite to failure confinement

In many cases dealing with mechanical elements, such as wing flaps, reaction time of 10s/100s of milliseconds is adequate (reason: inertia)

In some ways, catastrophic failures that are readily identified may be better than subtle failures that escape detection

**Example:** For redundant disks with two-way mirroring, detection latency was found to have a significant effect on the probability of data loss

See: http://www.hpdc.org/2004/papers/34.pdf

Failure detection latency can be made negative via "failure prediction" (e.g., in a storage server, increased error rate signals impending failure)



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# 25.5 Failure Avoidance Strategies

Advice for dependable system design and development

- Limit novelty [stick with proven methods and technologies]
- Adopt sweeping simplifications
- Get something simple working soon
- Iteratively add capability
- Give incentives for reporting errors
- Descope [reduce goals/specs] early
- Give control to (and keep it in) a small design team





# 25.6 Ethical Considerations

Risks must be evaluated thoroughly and truthfully

### IEEE Code of Ethics: As IEEE members, we agree to

- Accept responsibility in making decisions consistent with the safety, health and welfare of the public, and to disclose promptly factors that might endanger the public or the environment
- 6. Maintain and improve our technical competence and to undertake technological tasks for others only if qualified by training or experience, or after full disclosure of pertinent limitations;
- 7. Seek, accept, and offer honest criticism of technical work, to acknowledge and correct errors, and to credit properly the contributions of others

### ACM Code of Ethics: Computing professionals must

Minimize malfunctions by following generally accepted standards for system design and testing

Give comprehensive and thorough evaluations of computer systems and their impacts, including analysis of possible risks





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## National Society of Professional Engineers

Comprehensive code of engineering ethics

### I. Fundamental Canons

Engineers, in the fulfillment of their professional duties, shall:

- 1. Hold paramount the safety, health, and welfare of the public
- 2. Perform services only in areas of their competence
- 3. Issue public statements only in an objective and truthful manner
- 4. Act for each employer or client as faithful agents or trustees
- 5. Avoid deceptive acts
- 6. Conduct themselves honorably, responsibly, ethically, and lawfully so as to enhance the honor, reputation, and usefulness of the profession
- **II. Rules of Practice:** e.g., assignments to accept, whistle-blowing
- III. Professional Obligations: e.g., ack errors, be open to suggestions





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# **26 Failure Recovery**





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# 26.1 Planning for Recovery

Be prepared for recovery from both mild failures and disasters

- Have documented plans
- Have trained personnel





## **Recovery from Failures**

The recovery block scheme (originally developed for software)

ensure by else by	acceptance test primary module first alternate	e.g., sorted list e.g., quicksort e.g., bubblesort
÷		
else by else fail	last alternate	e.g., insertion sort

Computer system with manual backup may be viewed as a one-alternate recovery block scheme, with human judgment constituting the acceptance test

Before resorting to an alternate (hardware/software) module, one may reuse the primary module once or a few times

This scheme is known as "retry" or time redundancy and is particularly effective for dealing with transient or soft failures

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## 26.2 Types of Recovery



## Fail-Hard and Fail-Fast Systems

Subtle failures that are difficult to detect can be worse than overt ones that are immediately noticed

Design system so that their failures are difficult to miss (fail-hard system)

Corollary: Unneeded system complexity is to be avoided at all cost

Extended failure detection latency is undesirable because the occurrence of subsequent unrelated failures may overwhelm the system's defenses

Design system to support fast failure detection (fail-fast system)

Concurrent fault / error / malfunction detection better than periodic testing





## Fail-Stop and Fail-Over Strategies

**Fail-stop systems:** Systems designed and built in such a way that they cease to respond or take any action upon internal malfunction detection Such systems do not confuse the users or other parts of a distributed system by behaving randomly or, worse, maliciously upon failure A subclass of fail-safe systems (here, stopping is deemed a safe output)

**Fail-over:** Upon failure detection, often of the fail-stop kind, requests and other tasks are redirected to a backup system

Example – Fail-over on long-running connections for streaming media: all that is needed is to know the file being streamed, and current location within the file, to successfully switch over to a different server

Fail-over software is available for Web servers as part of firewalls for most popular operating systems

It monitors resources and directs requests to a functioning server

Fail-over features of Windows XP: <a href="http://msdn2.microsoft.com/en-us/library/ms686091.aspx">http://msdn2.microsoft.com/en-us/library/ms686091.aspx</a>



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## Fail-Over Features of Windows Servers



If System B fails, the DTC transaction manager on System A takes over. It reads the DTC log file on the shared disk, performs recovery, and then serves as the transaction manager for the entire cluster.

http://msdn2.microsoft.com/en-us/library/ms686091.aspx



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# 26.3 Interfaces and the Human Link

Believability and helpfulness of failure warnings

"No warning system will function effectively if its messages, however logically arrived at, are ignored, disbelieved, or lead to inappropriate actions." Foster, H. D., "Disaster Warning Systems," 1987

### **Unbelievable failure warnings:**

Failure event after numerous false alarms Real failure occurring in the proximity of a scheduled test run Users or operators inadequately trained (May 1960 Tsunami in Hilo, Hawaii, killed 61, despite 10-hour advance warning via sirens)

### **Unhelpful failure warnings:**

Autos – "Check engine" Computer systems – "Fatal error"



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## Human Factors in Automated Systems

"A few motorists have even driven off a cliff or into oncoming traffic after following [GPS] directions explicitly." [Gree09]

### The curse of highly successful automation efforts:

Human operators tend to "switch off" mentally, so they are not prepared when something unexpected happens (e.g., commuter train operators text-messaging or dozing off)





# 26.4 Backup Systems and Processes

Backup (*n*): Copy of data made for use after computer system failure or other data loss events [(v) to back up, backing up ]

Question: What method do you use to back up your data files?

Use removable storage: external hard drive, flash drive, CD/DVD E-mail file copy to yourself (not usable for very large files) Run a backup utility periodically (full or incremental backup) Subscribe to an on-line backup service (plain or encrypted) Do not overwrite files, create new versions

Businesses also need backups for programs and computer resources





# 26.5 Blame Assessment and Liability

### **Computer forensics is used to:**

Analyze computers belonging to defendants or litigants in legal cases Gather evidence against those suspected of wrongdoing Learn about a system to debug, optimize, or reverse-engineer it Analyze compromised computer systems to determine intrusion method Analyze failed computer systems to determine cause of failure Recover data after computer failures

There are several journals on computer forensics, digital investigations, and e-discovery





## 26.6 Learning from Failures



"When a complex system succeeds, that success masks its proximity to failure. ... Thus, the failure of the *Titanic* contributed much more to the design of safe ocean liners than would have her success. That is the paradox of engineering and design."

Henry Petroski, *Success through Failure: The Paradox of Design*, Princeton U. Press, 2006, p. 95



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## Some Notorious Failed Systems

Source: http://web.mit.edu/6.033/lec/s25.pdf

Automated reservations, ticketing, flight scheduling, fuel delivery, kitchens, and general administration, United Airlines + Univac, started 1966, target 1968, scrapped 1970, \$50M

Hotel reservations linked with airline and car rental, Hilton + Marriott + Budget + American Airlines, started 1988, scrapped 1992, \$125M

IBM workplace OS for PPC (Mach 3.0 + binary compatibility with AIX + DOS, Mac OS, OS/400 + new clock mgmt + new RPC + new I/O + new CPU), started 1991, scrapped 1996, \$2B

US FAA's Advanced Automation System (to replace a 1972 system), started 1982, scrapped 1994, \$6B

London Ambulance Dispatch Service, started 1991, scrapped 1992, 20 lives lost in 2 days, 2.5M



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## More Systems that Were Doomed to Failure

Source: http://web.mit.edu/6.033/lec/s25.pdf

- Portland, Oregon, Water Bureau, \$30M, 2002
- Washington D.C., Payroll system, \$34M, 2002
- Southwick air traffic control system \$1.6B, 2002
- Sobey's grocery inventory, Canada, \$50M, 2002
- King County financial mgmt system, \$38M, 2000
- Australian submarine control system, 100M, 1999
- California lottery system, \$52M
- Hamburg police computer system, 70M, 1998
- Kuala Lumpur total airport management system, \$200M, 1998
- UK Dept. of Employment tracking, \$72M, 1994
- Bank of America Masternet accounting system, \$83M, 1988
- FBI virtual case, 2004
- FBI Sentinel case management software, 2006



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# **27 Agreement and Adjudication**



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Abe, I just noticed – our marriage license has an expiration date! And it's today! Do you know anything about this, Abe? . . . Abe? . . . Abe?



"Next case: the Internet economy versus millions of investors who should have known better."







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#### STRUCTURE AT A GLANCE





Appendix: Past, Present, and Future



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## 27.1 Voting and Data Fusion



#### Introduction to Voting

Voting schemes and associated terminology in dependable computing were originally derived from concepts in sociopolitical elections

With inputs drawn from a small set of integers, the similarity between the two domains is strong

**Example:** Radar image analysis used to classify approaching aircraft type as civilian (0), fighter (1), bomber (2).

If three independent units arrive at the conclusions  $\langle 1, 1, 2 \rangle$ , then the presence of a fighter plane may be assumed

Option or candidate 1 "wins" or garners a majority

With a large or infinite input domain, voting takes on a new meaning

**Example:** There is no strict majority when distance of an approaching aircraft, in km, is indicated as  $\langle 12.5, 12.6, 14.0 \rangle$ , even though the good agreement between 12.5 and 12.6 could lead to a 12.55 km estimate



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#### Role of Voting Units in Dependable Computing

John von Neumann, 1956: "Probabilistic Logic and Synthesis of Reliable Organisms from Unreliable Components"

Hardware voting for multichannel computation

High performance (pipelined)

Software voting for multiversion programming

Imprecise results (approximate voting)

Consistency of replicated data

Weighted voting and weight assignment

TMR, NMR



Voting schemes in these three contexts share some properties General schemes have been devised to cover all these instances





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#### What We Already Know About Voting

Majority voting: Select the value that appears on at least  $\lfloor n/2 \rfloor$  + 1 of the *n* inputs

Number *n* of inputs is usually odd, but does not have to be

**Example:** vote(1, 2, 3, 2, 2) = 2



Majority fusers can be realized by means of comparators and muxes

This design assumes that in the case of 3-way disagreement any one of the inputs can be chosen

Can add logic to this fuser so that it signals three-way disagreement





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# 27.2 Weighted Voting

Virtually all voting schemes of practical interest can be formulated in terms of the generalized weighted voting model, as follows:

Given *n* input data objects  $x_1, x_2, ..., x_n$  and associated nonnegative real weights  $v_1, v_2, ..., v_n$ , with  $\sum v_i = V$ , compute output *y* and its weight *w* such that *y* is "supported by" a set of input objects with weights totaling *w*, where *w* satisfies a condition associated with the voting subscheme

Possible voting subschemes:

Unanimity	w = V	$\langle \mathbf{x}_1, \mathbf{v}_1 \rangle$
Majority	w > V/2	$\langle X_2, V_2 \rangle$
Supermajority	$w \ge 2V/3$	• $( \text{weighted} ) \rightarrow \langle y, w \rangle$
Byzantine	w > 2V/3	• fuser
Plurality	$(w \text{ for } y) \ge (w \text{ for any } z \ne y)$	
Threshold	<i>w</i> > a preset lower bound	$\langle \boldsymbol{X}_n, \boldsymbol{V}_n \rangle$



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#### There is More to Voting than Simple Majority

Plurality voting: Select the value that appears on the largest number of inputs

**Example:** vote(1, 3, 2, 3, 4) = 3

```
What should we take as the result of vote(1.00, 3.00, 0.99, 3.00, 1.01)?
```



It would be reasonable to take 1.00 as the result, because 3 participants agree or approximately agree with 1.00, while only 2 agree with 3.00

Will discuss approximate voting and a number of other sophisticated voting schemes later

Median voting: one way to deal with approximate values median(1.00, 3.00, 0.99, 3.00, 1.01) = 1.01

Median value is equal to the majority value when we have majority





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## Implementing Weighted Plurality Voting Units

Inputs: Data-weight pairs

Output: Data with maximal support and its associated tally



Source: B. Parhami, IEEE Trans. Reliability, Vol. 40, No. 3, pp. 380-394, August 1991



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#### Details of a Sorting-Based Plurality Voting Unit



The first two phases (sorting and combining) can be merged, producing a 2-phase design – fewer, more complex cells (lead to tradeoff)

Source: B. Parhami, IEEE Trans. Reliability, Vol. 40, No. 3, pp. 380-394, August 1991



#### **Threshold Voting and Its Generalizations**

Simple threshold (*m*-out-of-*n*) voting: Output is 1 if at least *m* of the *n* inputs are 1s

Majority voting is a special case of threshold voting:  $(\lfloor n/2 \rfloor + 1)$ -out-of-*n* voting

Weighted threshold (*w*-out-of- $\Sigma v_i$ ) voting: Output is 1 if  $\Sigma v_i x_i$  is *w* or more

Agreement or quorum sets  $\{x_1, x_2\}, \{x_2, x_3\}, \{x_3, x_1\}$  – same as 2-out-of-3  $\{x_1, x_2\}, \{x_1, x_3, x_4\}, \{x_2, x_3, x_4\}$ 

The 2nd example above is weighted voting with  $v_1 = v_2 = 2$ ,  $v_3 = v_4 = 1$ , and threshold w = 4



 $x_n$  Threshold gate

Agreement sets are more general than weighted voting in the sense of some agreement sets not being realizable as weighted voting

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#### Usefulness of Weighted Threshold Voting

Unequal weights allow us to take different levels of input reliabilities into account

Zero weights can be used to disable or purge some of the inputs (combined switching-voting)

Maximum-likelihood voting prob{ $x_1$  correct} = 0.9 prob{ $x_2$  correct} = 0.9 prob{ $x_3$  correct} = 0.8 prob{ $x_4$  correct} = 0.7

 $prob\{x_5 correct\} = 0.6$ 

Assume  $x_1 = x_3 = a$ ,  $x_2 = x_4 = x_5 = b$ 



Max-likelihood voting can be implemented in hardware via table lookup or approximated by weighted threshold voting (otherwise, we need software)

prob{a correct} =  $0.9 \times 0.1 \times 0.8 \times 0.3 \times 0.4 = 0.00864$ prob{b correct} =  $0.1 \times 0.9 \times 0.2 \times 0.7 \times 0.6 = 0.00756$ 





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#### Implementing Weighted Threshold Voting Units

**Ex.:** Implement a 4-input threshold voting unit with  $v_1 = v_2 = 2$ ,  $v_3 = v_4 = 1$ , and threshold w = 4

**Strategy 1:** If weights are small integers, fan-out each input an appropriate number of times and use a simple threshold voting unit



Strategy 2: Use table lookup based on comparison results

$x_1 = x_2$	$x_1 = x_3$	$x_2 = x_3$	$x_3 = x_4$	Result
1	Х	Х	х	<b>X</b> <sub>1</sub>
0	0	0	1	Error
0	1	0	1	<b>X</b> <sub>1</sub>
0	0	1	1	$X_2$
0	Х	X	0	Error

Is this table complete? Why, or why not?

Strategy 3: Convert the problem to agreement sets (discussed next)





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## 27.3 Voting with Agreement Sets

3-input majority voting has the agreement sets  $\{x_1, x_2\}, \{x_2, x_3\}, \{x_3, x_1\}$ 

If participants in any set agree, then their opinion is taken as the output (clearly, the agreement sets cannot include  $\{x_1, x_2\}$  and  $\{x_3, x_4\}$ , say, if we are seeking a unique output)





## Implementing Agreement-Set Voting Units

**Example:** Implement a voting unit corresponding to the agreement sets  $\{x_1, x_2\}, \{x_1, x_3, x_4\}, \{x_2, x_3, x_4\}$ 

Strategy 1: Implement as weighted threshold voting unit, if possible

Strategy 2: Implement directly

Find a minimal set of comparators that determine the agreement set



# 27.4 Variations in Voting

One can classify generalized weighted voting schemes into  $2^4 = 16$  categories based on dichotomies associated with input data ( $x_i$ s), output data (y), input weights ( $v_i$ s), and output weight (w)

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#### **Generalized Median Voting**

To find the median of a set of numbers, repeatedly remove largest and smallest numbers, until only one or two remain



If we replace "largest and smallest numbers" by "the two inputs that are furthest apart," we can use an arbitrary distance metric in our screening

A distance metric is any metric (mapping of pairs of inputs into real values) that satisfies the three conditions:

Isolation	d(x, y) = 0  iff  x = y
Symmetry	d(x, y) = d(y, x)
Triangle inequality	$d(x, y) + d(y, z) \ge d(x, z)$



For example, the Hamming distance satisfies these conditions



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## The Impossibility of Perfect Voting

Properties of an ideal voting scheme:

1. No big brother

(participants free to express preferences)

- 2. Independence of irrelevant alternatives (preference for one candidate over another is independent of all others)
- 3. Involvement (every outcome is possible)
- 4. No dictatorship or antidictatorship (outcome not always conforming to, or opposite of, one participant's view)

#### True majority voting scheme:

Each participant orders all the candidates; no circular preference allowed Choose a candidate who beats every other one in pairwise competitions (both simple majority and plurality rules fail to choose a candidate)

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## Briten

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Arrow's Theorem: No voting scheme exists that satisfies all four conditions



#### **Approximate Voting**

The notion of an input object "supporting" a particular output (akin to a hypothesis supporting an end result or conclusion) allows us to treat approximate and exact voting in the same way

**Example 1:** Input objects are points in the 2D space and the level of "support" between them is a function of their Euclidean distance

**Example 2:** Input objects are conclusions of character recognizers as to the identity of a character, with varying degrees of mutual support



## **Approval Voting**

Approval voting was introduced to prevent the splitting of votes among several highly qualified candidates from leading to the election of a less qualified candidate in plurality voting

In approval voting, a participant divides the candidates into two subsets of "qualified" and "not qualified" and indicates approval of the first subset



In the context of computing, approval voting is useful when a question has multiple answers or when the solution process is imprecise or fuzzy

Example question: What is a safe setting for a particular parameter in a process control system?

When the set of approved values constitute a continuous interval of real values, we have "interval" inputs and "interval" voting

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### **Interval Voting**

Inputs to the voting process are intervals, representing approved values

How should the voting result be derived from the input intervals?

Various combining rules can be envisaged If there is overlap among all intervals, then the decision is simple

Depending on context, it may make sense to consider greater levels of approval near the middle of each interval or to associate negative approval levels outside the approved intervals



## 27.5 Distributed Agreement

**Problem:** Derive a highly reliable value from multiple computation results or stored data replicas at multiple sites

**Key challenge:** Exchange data among nodes so that all healthy nodes end up with the same set of values; this guarantees that running the same decision process on the healthy nodes produces the same result

Errors are possible in both data values and in their transmission between sites

Agreement algorithms generally use multiple rounds of communication, with values held at each site compared and filtered, until the set of values held at all sites converge to the same set



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## **Byzantine Failures in Distributed Voting**

Three sites are to collect three versions of some parameter and arrive at consistent voting results



## The Interactive Consistency Algorithm

#### ICA(0) [no failure]

- 1. The transmitter sends its value to all other n 1 nodes
- 2. Each node uses the value received from the transmitter, or a default value  $\Phi$  if it received no value

#### **ICA(f)**, **f > 0** [*f* failures]

- 1. The transmitter sends its value to all other n 1 nodes
- 2. Let  $v_i$  be the value received by node *i* from the transmitter, or a default value  $\Phi$  if it received no value; node *i* then becomes the transmitter in its own version of ICA(f 1), sending its value to n 2 nodes

3. For each node *i*, let  $v_{i,j}$  be the value it received from node *j*, or a default

value  $\Phi$  if it received no value from node *j*. Node *i* then uses the value majority( $v_{i,1}, v_{i,2}, \ldots, v_{i,i-1}, v_{i,i+1}, \ldots, v_{i,n}$ ) O( $n^{f+1}$ ) messages needed, in f + 1 rounds, to tolerate *f* Byzantine failures







### **Building upon Consensus Protocols**



Message from a good node is eventually delivered to all good nodes unmodified

Source: M. Correia, N. Ferreira Neves, P. Veríssimo, "From Consensus to Atomic Broadcast: Time-Free Byzantine-Resistant Protocols without Signatures," *The Computer J.*, 2005.





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#### **Correctness and Performance of ICA**

**Theorem 1:** With ICA(f), all nonfailed nodes will agree on a common value, provided that  $n \ge 3f + 1$  (proof is by induction on f)

ICA works correctly, but it needs an exponential number of messages:  $(n-1) + (n-1)(n-2) + (n-1)(n-2)(n-3) + \ldots + (n-1)(n-2) \ldots (n-m)$ 

More efficient agreement algorithms exist, but they are more difficult to describe or to prove correct; f + 1 rounds of message exchange is the least possible, so some algorithms trade off rounds for # of messages

**Theorem 2:** In a network *G* with *f* failed nodes, agreement is possible only if the connectivity is at least 2f + 1



#### The Byzantine Generals Problem

A general and n - 1 lieutenants lead n divisions of the Byzantine army camped on the outskirts of an enemy city

The *n* divisions can only communicate via messengers

We need a scheme for the generals to agree on a common plan of action (attack or retreat), even if some of the generals are traitors who will do anything to prevent loyal generals from reaching agreement

The problem is nontrivial even if messengers are totally reliable

With unreliable messengers, the problem is very complex





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#### Byzantine Generals with Reliable Messengers



# 27.6 Byzantine Resiliency

To tolerate *f* Byzantine failures:

We need 3f + 1 or more FCRs (fault containment regions)

FCRs must be interconnected via at least 2f + 1 disjoint paths

Inputs must be exchanged in at least f + 1 rounds

Corollary 1: Simple 3-way majority voting is not Byzantine resilient

Corollary 2: Because we need 2f + 1 good nodes out of a total of 3f + 1 nodes, a fraction (2f + 1)/(3f + 1) = 2/3 + 1/(9f + 3) of the nodes must be healthy

This is greater than a supermajority (2/3) requirement



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# 28 Fail-Safe System Design



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"Have you seen the use-by date on this porridge?"



"Damn those Health & Safety guys."







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#### STRUCTURE AT A GLANCE





Appendix: Past, Present, and Future



Part VII – Failures: Computational Breaches



## 28.1 Fail-Safe System Concepts

**Fail-safe:** Produces one of a predetermined set of safe outputs when it fails as a result of "undesirable events" that it cannot tolerate

Fail-safe traffic light: Will remain stuck on red

Fail-safe gas range/furnace pilot flame: Cooling off of the pilot assembly due to the flame going out will shut off the gas intake valve

A fail-safe digital system must have at least two binary output lines, together representing the normal outputs and the safe failure condition Reason: If we have a single output line, then even if one value (say, 0) is inherently safe, the output stuck at the other value would be unsafe Two-rail encoding is a possible choice: **0**: 01, **1**: 10, **F**: 00, 11, or both

**Totally fail-safe:** Only safe erroneous outputs are produced, provided another failure does not occur before detection of the current one **Ultimate fail-safe:** Only safe erroneous output is produced, forever



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# 28.2 Principles of Safety Engineering

#### Principles for designing a safe system (J. Goldberg, 1987)

- 1. Use barriers and interlocks to constrain access to critical system resources or states
- 2. Perform critical actions incrementally, rather than in a single step
- 3. Dynamically modify system goals to avoid or mitigate damages
- 4. Manage the resources needed to deal with a safety crisis, so that enough will be available in an emergency
- 5. Exercise all critical functions and safety features regularly to assess and maintain their viability
- 6. Design the operator interface to provide the information and power needed to deal with exceptions
- 7. Defend the system against malicious attacks





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## 28.3 Fail-Safe Specifications

Amusement park train safety system

Signal  $s_B$  when asserted indicates that the train is at beginning of its track (can move forward, but should not be allowed to go back)

Signal  $s_E$  when asserted indicates that the train is at end of its track (can go back, but should not move forward)



Is the specification above consistent and complete?

No, because it does not say what happens if  $s_B = s_E = 1$ ; this would not occur under normal conditions, but because such sensors are often designed to fail in the safe mode, the combination is not impossible

Why is this a problem, though? (Train simply cannot be moved at all)

Completeness will prevent potential implementation or safety problems

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#### Simple Traffic Light Controller

Six signals: red, amber, and green light control for each direction (no left turn signals)

Safety flip-flop, when set, forces flashing red lights in both directions

$$S = (A_1 \vee G_1) \land (A_2 \vee G_2)$$

Let  $g_1$  be the computed green light signal for direction 1, and so on for other signals, and  $G_1$  be the signal that controls the light

$$G_1 = \overline{S} \wedge g_1 \wedge (\overline{A_2 \vee G_2})$$



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## 28.4 Fail-Safe Combinational Logic

Similar to the design of self-checking circuits: Design units so that for each fault type of interest, the output is either correct or safe

Totally fail-safe: Fail-safe and self-testing [Nico98]

Strongly fail-safe: Intermediate between fail-safe and totally fail-safe For each fault *x* it is either totally fail-safe *or* it is fail-safe with respect to *x* and strongly fail-safe with respect to  $x \cup y$ , where *y* is another fault





### Fail-Safe 2-out-of-4 Code Checker

Input: 4 bits *abcd*, exactly 2 of which must be 1s



Output will become permanently 00 upon the first unsafe condition



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## 28.5 Fail-Safe State Machines

Use an error code to encode states

Implement the next-state logic so that the machine is forced to an error state when something goes wrong

Possible design methodology:

Use Berger code for states, avoiding the all 0s state with all-1s check, and vice versa

Implement next-state logic equations in sum-of-products form for the main state bits and in product-of-sums form for the check state bits

The resulting state machine will be fail-safe under unidirectional errors

	Input			
<u>State</u>	<u>x=0</u> .	<u>x=1</u>	<u>State</u>	Encoding
A	E	В	А	001 10
В	С	D	В	010 10
С	A	D	С	011 01
D	E	D	D	100 10
E	А	D	E	101 01

Hardware overhead for *n*-state machine consists of O(log log *n*) additional state bits and associated next-state logic, and a Berger code checker connected to state FFs

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## 28.6 System- and User-Level Safety

Principle: False alarms are tolerable, but missed failures are not

A binary test is characterized by its sensitivity and specificity

**Sensitivity** is the fraction of positive cases (e.g., people having a particular disease) that the test identifies correctly

**Specificity** is the fraction of negative cases (e.g., healthy people) that are correctly identified

There exists a fundamental tradeoffs between sensitivity and specificity

We want to err on the side of too much sensitivity



### Interlocks and Other Safety Mechanisms

**Dead-man's switch (kill switch):** Used in safety-critical systems that should not be operable without the presence of an operator

Usually, takes the form of a handle or pedal that the operator must touch or press continuously

First used in trains to stop them in the event of operator incapacitation





# **A** Past, Present, and Future





Part VII – Failures: Computational Breaches





"I am not disorganized — I know *exactly* where everything is! The newer stuff is on top and the older stuff is on the bottom."



"We couldn't afford faster computers, so we just made them sound faster."



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#### STRUCTURE AT A GLANCE





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## A.1 Historical Perspective

Antonin Svoboda (1907-80), built SAPO, the first fault-tolerant computer—it used triplication with voting for error correction; necessitated by poor component quality



Motivated by NASA's planned Solar System exploration taking 10 years (the Grand Tour), Algirdas Avizienis built Jet Propulsion Lab's self-testing-and-repairing (STAR) computer





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#### Dependable Computing in the 1950s

SAPO, built in Czechoslovakia in 1951 by Antonin Svoboda Used magnetic drums and relays

SAPO fault tolerance features were motivated by [Renn84]: Low quality of components available to designers Severe (political) pressure to get things right

With the advent of transistors, which were much more reliable than relays and vacuum tubes, redundancy methods took a back seat (used only for exotic or highly critical systems)



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### Dependable Computing in the 1960s

NASA began extensive development in long-life, no-maintenance computer systems for (manned) space flight

Orbiting Astronomical Observatory (OAO), transistor-level fault masking

The Apollo Guidance System, triplication and voting

Mars Voyager Program (late 1960s, cancelled)

Deep space missions (e.g., the cancelled Solar System Grand Tour) JPL self-testing-and-repairing computer (STAR)

Also:

AT&T Electronic Switching System (ESS), down time of minutes/decade Serviceability features in mainframes, such as IBM System/360



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### Dependable Computing in the 1970s

Two influential avionics systems:

Software-implemented fault tolerance (SIFT), SRI, Stanford Univ. Spun off a commercial venture, August Systems (process control) Fault-tolerant multiprocessor (FTMP), Draper Lab., MIT

The Space Shuttle (4 identical computers, and a 5th one running different software developed by a second contractor)

Redundancy in control computers for transportation (trains)

Also: First Fault-Tolerant Computing Symposium, FTCS (1971) Systems for nuclear reactor safety Tandem NonStop (1976), transaction processing



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### Dependable Computing in the 1980s

1980: IFIP WG 10.4 on Dependable Computing and Fault Tolerance

Continued growth in commercial multiprocessors (Tandem)

Advanced avionics systems (Airbus)

Automotive electronics

Reliable distributed systems (banking, transaction processing)





### Dependable Computing in the 1990s

DCCA conferences, started in 1989, continued until 1999 (DCCA-7); subsequently merged with FTCS to become DSN Conf.

SEU tolerance

BIST

IBM G4 fault-tolerant mainframe

Robust communication and collaboration protocols (routing, broadcast, f-t CORBA)





### Dependable Computing in the 2000s

2000: First DSN Conf. (DSN-2000, 30th in series going back to FTCS) Now composed of two symposia DCCS: Dependable Computing and Communications Symp. PDS: Performance and Dependability Symp.

2004: IEEE Trans. Dependable and Secure Computing

Deep submicron effects

F-TARM, Sparc

Intrusion detection and defense

Widespread use of server farms for directory sites and e-commerce (Google, e-Bay, Amazon)



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#### Dependable Computing in the 2010s

2010: DSN Conf. held its 40th in series, going back to FTCS

Cloud computing reliability focus of an IEEE TDSC 2013 special issue

Greater integration of reliability and security concerns

End of decade: Half-century mark of the field and of the DSN Conf.





# A.2 Long-Life Systems

Genesis: Computer systems for spacecraft on multiyear missions, with no possibility of repair

Today: More of the same, plus remotely located, hard-to-access systems for intelligence gathering or environmental monitoring

Typical systems for case studies: NASA OAO, Galileo, JPL STAR, . . . , Space Station Communication satellites Remote sensor networks





## The JPL STAR Computer

Became operational in 1969, following studies that began in 1961 Standby redundancy for most units, 3 + 2 hybrid redundancy for TARP mod-15 inverse residue code to check arithmetic ops and bus transfers Also used other codes, plus various hardware/software sanity checks



## A.3 Safety-Critical Systems

Genesis: Flight control, nuclear reactor safety, factory automation

Today: More of the same, plus high-speed transportation, health monitoring, surgical robots

Typical systems for case studies: CMU C.vmp Stanford SIFT MIT FTMP August Systems High-speed train controls Automotive computers



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### Avionic Fly-by-Wire Systems



### Automotive Drive-by-Wire Systems



## A.4 High-Availability Systems

Genesis: Electronic switching systems for telephone companies

Today: More of the same, plus banking, e-commerce, social networking, and other systems that can ill-afford even very short down times

Typical systems for case studies: AT&T ESS 1-5, telephone switching, 1965-1982 Tandem NonStop I/II- . . . -Cyclone-CLX800, 1976-1991 Stratus FT200-XA2000, 1981-1990 Banking systems Portals and e-commerce (Google, Amazon)





## Tandem NonStop Cyclone

Announced in 1989 for database and transaction-processing applications (descendant of the first NonStop system announced in 1976)



## Google Data Center Architecture

Uses commodity (COTS) processing and communication resources Data replication and software used to handle reliability issues Massive hardware replication, needed for capacity and high performance, also leads to fault tolerance

Consistency problems are simplified by forcing read-only operation most of the time and shutting down sections of the system for infrequent updates





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## A.5 Commercial and Personal Systems

Genesis: Due to highly unreliable components, early computers had extensive provisions for fault-masking and error detection/correction

Today: Components are ultrareliable, but there are so many of them that faults/errors/malfunctions are inevitable

Typical systems for case studies: SAPO IBM System 360 IBM POWER6 microprocessor



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### The IBM Power6 Microprocessor



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## A.6 Trends, Outlook, and Resources

- Nobel Laureate Physicist Niels Bohr said: "Prediction is very difficult, especially if it's about the future." [Paraphrased by Yogi Berra in his famous version]
- Anonymous quotes on the perils of forecasting: "Forecasting is the art of saying what will happen, and then explaining why it didn't."
   "There are two kinds of forecasts: lucky and wrong."
   "A good forecaster is not smarter than everyone else; he merely has his ignorance better organized."
- Henri Poincare was more positive on prediction: "It is far better to foresee even without certainty than not to foresee at all."

## 2020s | 2030s | 2040s | 2050s







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### **Current Challenges in Dependable Computing**

Dependable computer systems and design methods continue to evolve

Current trend: Building dependable systems from mass-produced, commercial off-the-shelf (COTS) subsystems, not custom-made units (that is, dependability is viewed as a layer put around otherwise untrustworthy computing elements)

This is similar to the RAID approach to dependable mass storage

Challenge 1: The curse of shrinking electronic devices (nanotechnology) Challenge 2: Reliability in cloud computing (opportunities & problems) Challenge 3: Smarter and lower-power redundancy schemes (brainlike) Challenge 4: Ensuring data longevity over centuries, even millennia Challenge 5: Dependability verification (reasoning about uncertainty) Challenge 6: Counteracting combined effects of failures and intrusions Challenge 7: Reliability as roadblock for exascale systems (reliability wall)

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### The Curse of Shrinking Electronic Devices

Size shrinkage leads to

Unreliable operation due to "crummy" devices and interconnects Need for modeling parameter fluctuations due to process variations Difficulties in testing, particularly with regard to delay faults

Reliability issues

Some of the area/power gains must be reinvested in reliability features

Circuit modeling Can no longer model by considering only worst-case latencies

Testing challenges

Need for more tests to cover combinations of parameter values

http://webhost.laas.fr/TSF/WDSN10/WDSN10\_files/Slides/WDSN10\_Wunderlich-Becker-Polian-Hellebrand.pdf





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### Computing with Majority Elements

Several new technologies offer simple and power-efficient majority gates

In quantum-dot cellular automata (QCA), each cell contains 2 electrons that can tunnel between corner spots (dots), but can't leave the cell



### **Cloud Computing Overview**



### Reliability in Cloud Computing

Theoretically, the cloud can be highly reliable (no single point of failure) Challenge of integrating diverse resources from multiple providers Identifying weakest links; assigning blame in the event of outages Risk assessment in e-commerce and safety-critical processing

Planning for and modeling outages Accidental or deliberate

Balance and accountability Recall Amdahl's law of reliability

**Risk assessment** 



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Smarter and Lower-Power Redundancy Schemes

Brain-inspired computing

 Human brain has a very inelegant design, but it operates on 20 W (a supercomputer doing a small fraction of the brain's tasks needs MWs)

Heor

- Ad-hoc additions over eons
- Functional redundancy
- Layered improvement
- Electrochemical signals
- Slow signaling

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• Yet, functionally effective!



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Limbic

system

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Reptilian

brain

#### **Engineering the Future: System Attributes**

Toward Self-Organizing, Self-Improving, Self-Healing, and Self-Sustaining Systems



#### Example: Learning Voter

Suppose the voter/fuser keeps track of disagreements by units 1-3 so that it learns which units are more trustworthy

If unit 1, say, disagrees often, then the first disagreement between units 2 and 3 may be cause for the system to shut itself down





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### **Ensuring Data Longevity**

Typical media used for storing data have lifespans of 3-20 years

We can lose data to both media decay and format obsolescence

We're not yet aware of, or sensitive to, decay and obsolescence, because we move data to newer devices every few years as part of routine upgrades



#### **Reference:**

Springer Encyclopedia of Big Data Technologies, Article on "Data Longevity and Compatibility," 2019



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#### Reasoning About Uncertainty

Uncertainty can exist in data (facts)

Data not available

Data unreliable (measurement / representation errors, estimation

Uncertainty can exist in rules (knowledge)

Rules may be empirical

Rules may be based on typical or average cases

Reasoning about uncertainty requires Representing uncertain data or rules Combining two or more items uncertain data items Drawing inferences based on uncertain data and rules

Categories of approaches Probabilistic analysis (Bayesian logic) Certainty/Confidence factors Evidence/Belief theory (Dempster-Shafer theory) Continuous truth values (Fuzzy logic)

#### **Reference:**

*Int'l J. Approx. Reasoning*, Special Issue 40 Years of Research Dempster-Shafer Theory, December 2016

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#### Interaction Between Failures and Intrusions

Intruders can cause failures as a way of disrupting systems Self-driving cars Fly by wire Electronic terrorism

Failures may facilitate security breaches





#### Overcoming the Reliability Wall

We have known about memory wall limiting performance for some time

Reliability wall limits the scalability of parallel applications

Yang et al's two case studies [Yang12] Intrepid supercomputer ASCI White supercomputer

The notion of reliability wall is introduced in Problem 3.17 Amdahl's constant-task speed-up formula: s = p/[1 + f(p - 1)]Speed-up is upper-bounded by 1/*f* Gustafson's constant-running-time speed-up formula: s = f + p(1 - f)Speed-up is unbounded for any f < 1

If the reliability overhead is a superlinear function of p, then a reliability wall exists under Amdahl's interpretation but not under Gustafson's





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# Dependable Computing Through the Decades



## **Resources for Dependable Computing**

IFIP Working Group 10.4 on Dependable Computing and Fault Tolerance IFIP WG 10.4 — http://www.dependability.org/wg10.4/

IEEE/IFIP Int'l Conf. Dependable Syst's and Networks (2021, Taipei) DSN — http://www.dsn.org/

European Dependable Computing Conf. (2020, Munich) EDCC — http://edcc.dependability.org/

IEEE Pacific Rim Int'l Symp. Dependable Computing (2021, Perth) PRDC — http://prdc.dependability.org/

Int'l Conf. Computer Safety, Reliability and Security (2021, York, UK) SAFECOMP — http://www.safecomp.org/

IEEE Trans. Dependable and Secure Computing (since 2004) IEEE TDSC — https://www.computer.org/csdl/journal/tq



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### We Have Built a Framework

... and skimmed over a number of applicable techniques

... You should now be ready to pursue some of the details

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