Reliability of 1002 Software-based Systems in which one Channel is "Possibly Perfect"

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The set-up

- 1-out-of-2 design-diverse, 2-channel software based system
- We are interested in *probability of failure on demand (pfd)*
 - E.g. reactor protection system
 - But much of what we say here may also apply in wider contexts, e.g. continuously operating fault tolerant systems
- We know such fault tolerant approaches can be effective ways to *achieve* reliability
 - E.g. reliability in eventual operational use of the Airbus A320 and later fault tolerant flight control systems?
- BUT.....



....There's a problem:

Although such an approach may work "on average" (in some sense), it's hard to know whether it has worked in a particular instance - and *how reliable* the resulting system will be in operation

- Cannot assume independence of version/channel failures
 - In fact they will not fail independently
- $Pfd_{sys} > pfd_A.pfd_B$
 - Experiments tell us this
 - So does theory
- Need to know "how dependent" the failure processes of the different channels are
- Measuring this is as hard as measuring Pfd_{sys} by treating it as a black box
- So...an impasse?



A possible way out

Consider a 1002 system in which channel A is "highly functional", and therefore complex, *but channel* B *is simpler and thus possibly "perfect"*

- Perfect means it will never experience a failure
- Possibly perfect means there is some uncertainty about its perfection

– In particular there is a probability of imperfection

- For *A* our uncertainty concerns whether it will fail on a randomly selected demand: probability pfd_A
- for *B* our uncertainty concerns whether it is not perfect: probability pnp_A



Aleatory and Epistemic Uncertainty

- Aleatory uncertainty is "uncertainty in the world", or irreducible uncertainty
 - Uncertainty about *A* failing, about *B* not being perfect both involve aleatoric uncertainty
- Epistemic uncertainty is "uncertainty about the world", or reducible uncertainty
 - Sometimes called "model uncertainty"
 - E.g. uncertainty about the size of pfd_A and of pnp_B
- We now analyse our system in two stages: aleatoric, then epistemic

But now suppose for the moment we know $pfd_A = p_A$ and $pnp_B = p_B...$



Aleatoric uncertainty for 1002 system

 $P(\text{system fails on randomly selected demand} | pfd_A = p_A, pnp_B = p_B)$ $= P(\text{system fails} | A \text{ fails, } B \text{ not perfect}, pfd_A = p_A, pnp_B = p_B)$ $\times P(A \text{ fails, } B \text{ not perfect} | pfd_A = p_A, pnp_B = p_B)$ $+ P(\text{system fails} | A \text{ succeeds, } B \text{ not perfect}, pfd_A = p_A, pnp_B = p_B)$ $\times P(A \text{ succeeds, } B \text{ not perfect} | pfd_A = p_A, pnp_B = p_B)$ $+ P(\text{system fails} | A \text{ fails, } B \text{ perfect}, pfd_A = p_A, pnp_B = p_B)$ $\times P(A \text{ fails, } B \text{ perfect} | pfd_A = p_A, pnp_B = p_B)$ $+ P(\text{system fails} | A \text{ succeeds, } B \text{ perfect}, pfd_A = p_A, pnp_B = p_B)$ $\times P(A \text{ fails, } B \text{ perfect} | pfd_A = p_A, pnp_B = p_B)$ $+ P(\text{system fails} | A \text{ succeeds, } B \text{ perfect}, pfd_A = p_A, pnp_B = p_B)$ $\times P(A \text{ succeeds, } B \text{ perfect} | pfd_A = p_A, pnp_B = p_B)$

Assume, conservatively, that if *B* is imperfect it fails whenever *A* does

 $P(\text{system fails on randomly selected demand} | pfd_A = p_A, pnp_B = p_B)$

= $P(A \text{ fails}, B \text{ not perfect} | pfd_A = p_A, pnp_B = p_B)$

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Aleatory uncertainty (contd)

$$\begin{split} P(A \text{ fails}, B \text{ imperfect} \mid pfd_A &= p_A, pnp_B = p_B) \\ &= P(A \text{ fails} \mid B \text{ imperfect}, pfd_A = p_A, pnp_B = p_B) \\ &\times P(B \text{ imperfect} \mid pfd_A = p_A, pnp_B = p_B) \end{split}$$

(Im)perfection of B tells us nothing about the failure of A on this demand; hence,

$$= P(A \text{ fails} | pfd_A = p_A, pnp_B = p_B)$$
$$\times P(B \text{ imperfect} | pfd_A = p_A, pnp_B = p_B)$$
$$= p_A \times p_B$$

Compare with two (un)reliable channels, where failure of B on this demand does increase likelihood A will fail on same demand

$$P(A \text{ fails} | \boldsymbol{B} \text{ fails}, pfd_A = p_A, pfd_B = p_B)$$

$$\geq P(A \text{ fails} | pfd_A = p_A, pfd_B = p_B)$$

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Epistemic uncertainty for 1002 system

• We have shown that the events "*A* fails" and "*B* is imperfect" are conditionally independent at the aleatoric level

– Probability of system failure is (conditionally) $p_A \times p_B$

- Remaining uncertainty centres *only* on p_A and p_B
- We represent this *epistemic uncertainty by*

 $F(p_A, p_B) = P(pfd_A < p_A, pnp_B < p_B)$

- E.g. could think of this as his Bayesian posterior distribution when an assessor has collected evidence from testing, verification, other kinds of analysis, etc, etc
- The unconditional (subjective) probability of system failure is

 $\int_{\substack{0 \le p_A \le 1\\ 0 \le p_B \le 1}} p_A p_B dF(p_A, p_B)$



Epistemic uncertainty (contd)

- The *only* source of dependence in the model comes in via *F*
- If this were to factorise, i.e assessor's beliefs about the parameters were independent,

P(system fails on randomly selected demand) = P(A fails, B not perfect)

$$= \int_{\substack{0 \le p_A \le 1\\ 0 \le p_B \le 1}} p_A p_B dF(p_A, p_B)$$

$$= \int_{0 \le p_A \le 1} p_A dF(p_A) \times \int_{0 \le p_B \le 1} p_B dF(p_B)$$

And the assessor's task is reduced to estimating just the two posterior (marginal) means

• But this will never be true!



Reliability estimation of 1002 system

- Most assessors would find it hard to tell us what their *F* is
- So what can be done?
- Well....where does the "dependency of beliefs" about the parameters come from?
- A source of dependency is the possibility of common faults at a high level, e.g. misunderstanding of system requirements
- One way forward is to place probability mass, say C, at the point (1,1) in the (p_A, p_B) -plane to represent the assessor's (subjective) probability that there *are* such faults
- The effect of this is *conservative*: if there *are* such faults he believes *A* fails with certainty, and *B* is not perfect with certainty

- there is a chance C that $p_A p_B = 1$, i.e. that the system is certain to fail



Reliability estimation (contd)

$$P(\text{system fails on randomly selected demand}) = \int_{\substack{0 \le p_A \le 1\\0 \le p_B \le 1}} p_A p_B dF(p_A, p_B)$$

$$= C \times \int p_A p_B dF(p_A, p_B | p_A = p_B = 1) + (1 - C) \times \int p_A p_B dF(p_A, p_B | p_A, p_B \neq 1)$$

= $C + (1 - C) \times \int p_A p_B dF(p_A, p_B | p_A, p_B \neq 1)$
But the last integrand factorises, so

P(system fails on randomly selected demand)

$$= C + (1 - C) \times \int p_A dF(p_A \mid p_A \neq 1) \times \int p_B dF(p_B \mid p_B \neq 1)$$
$$= C + (1 - C) \times P_A^* \times P_B^*$$



Discussion

Of course this is not a silver bullet. But...

- The handling of aleatory uncertainty is greatly simplified compared with the case of two *certainly fallible* channels
- The architecture *is* a special one, but it is very plausible for certain applications
 - E.g. as a means of *achieving* reliability for, say, a protection system; or for functional channel plus monitor; or highly functional channel plus getyou-home channel
- The conservative bottom-line result involves only *three parameters* and it may be possible to estimate these for real systems



