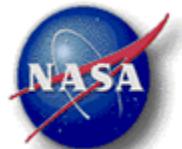


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The Effect of Job Performance Aids on Quality Assurance

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NASA Quality Leadership Forum
March 9-10, 2016





Motivation

- Quality Engineer for Sandia National Laboratories since 2005
- Purdue MSIE, May 2014, Human Factors Engineering
- Wanted to bridge the disciplines of Human Factors and Quality Assurance (QA)
- Previously created a ***job performance aid*** (JPA) for novice QA co-workers for concurrent dual verification tasks
- A checklist is one type of JPA (others are procedures, manuals, training videos, etc.)
- First-ever research on JPAs in a QA context



Quality Assurance Context



- *DOE Guide to Good Practices for Independent Verification (1993):*
Concurrent Dual Verification – A method of checking an operation, an act of positioning, or a calculation in which the verifier independently observes and/or confirms the activity
- NASA-STD 8709.22 (2010) definitions:
Process Witnessing – Physical observation of a contractor test or work process to ensure that the process is being correctly performed in accordance with prescribed procedures and contract requirements.



History

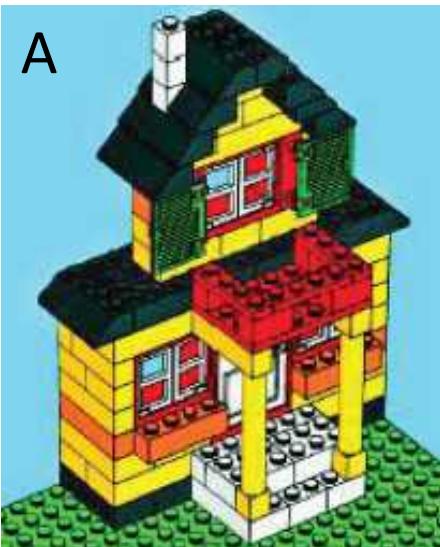
- Boeing 299 (later B-17) crash in 1935 led to pilot's checklist
- USAF behavioral research on training aids (e.g. Miller, 1953) led to the “Task Analysis” methodology
- JPA research continued through the 1970s; findings included:
 - Reduced errors in complex tasks that were infrequently performed
 - Shortened the training time for novice users
 - Different formats (pictures or text) conveyed information differently
- JPA interest resurfaced after Three Mile Island incident (1979)
- JPAs now adopted by various “high consequence” industries: aviation, nuclear power, medicine, aerospace
- Popular interest: *The Checklist Manifesto* (2010)



Experimental Task Selection

- Guidelines:
 - Not too simple, not too complex
 - Consistent with high consequence environment
- Solution: Lego™ assembly task
 - Participant expertise not a covariant: all users are novices
 - Reasonable similarity to manufacturing environment
 - Easy to inject faults and measure performance
- Within subjects design, 2 different Lego™ patterns
 - One assembled with JPA present, one assembled without
 - 24 participants, counterbalanced for learning effect

Lego™ Patterns



Pattern A: 104 pieces

Pattern B: 150 pieces

7 faults injected into each pattern (14 total)

Fault Types:

1. Markings



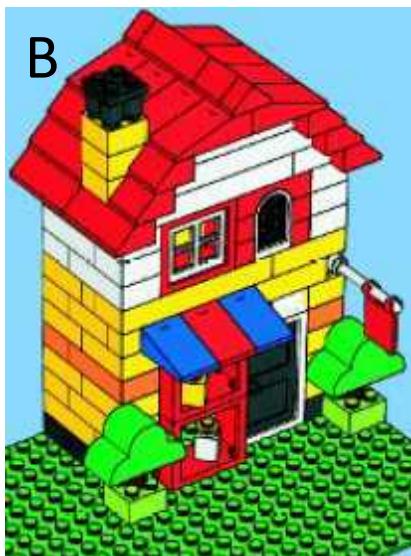
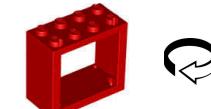
2. Incorrect piece(s)



3. Wrong order



4. Wrong orientation



Assumption: Constant probability of detection for all fault types



JPA Design

- Common themes in the literature*:
 - The focus should be on the user
 - Fully understand the job function
 - Fully understand the behaviors used
 - Information must be task oriented
 - Brief, concise, explicit instructions; be directive and action-specific
 - Use simplified and standard language
 - Final important step: validation with expert users
- JPA for this experiment:
 - Short, concise, and simple checklist
 - Elicits behavioral cues to enhance the detection of faults

* Best references are Shriver et al. (1982), Smillie (1985), and Gawande (2010)



Checklist

- Your role as an observer is an essential part of this important task. Complex assemblies require a second set of eyes in order to catch any errors.
- Pay attention for the following types of error:
 - An incorrect piece is installed, meaning that it is either the wrong size, wrong color, or wrong markings
 - The correct piece is installed, but in the wrong orientation
 - The correct piece is installed, but in the wrong location
- Feel free to ask questions about the task at any time. If necessary, ask the assembler to stop until you are comfortable with proceeding.
- The assembler should not turn to the next page of the instructions without your approval.
- For each page of the instructions, the order of assembly does not matter.
- The box contains 512 total parts. Some parts will be used and some will not.

Behavior cues
Error avoidance

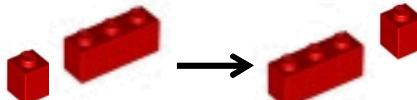


Results (1)

- Participant scores ranged from 43% - 100% detection of faults
 - Majority of participants scored in the 50-60% range
 - Traditional inspection results yield ~80% success rate
- Poor performance overall
- Suggests limitations to concurrent dual verification

| Subject | Pattern A Trials | Pattern A Detections | Pattern B Trials | Pattern B Detections | Percent Detected |
|---------|------------------|----------------------|------------------|----------------------|------------------|
| 1 | 7 | 7 | 7 | 7 | 100% |
| 2 | 7 | 5 | 6 | 5 | 77% |
| 3 | 7 | 4 | 7 | 3 | 50% |
| 5 | 7 | 7 | 6 | 3 | 77% |
| 6 | 7 | 6 | 6 | 4 | 77% |
| 7 | 7 | 5 | 7 | 4 | 64% |
| 8 | 7 | 4 | 7 | 5 | 64% |
| 9 | 7 | 6 | 7 | 5 | 79% |
| 10 | 7 | 5 | 7 | 7 | 86% |
| 11 | 7 | 3 | 7 | 4 | 50% |
| 12 | 7 | 3 | 7 | 4 | 50% |
| 13 | 7 | 3 | 7 | 4 | 50% |
| 14 | 7 | 3 | 7 | 6 | 64% |
| 15 | 7 | 4 | 7 | 4 | 57% |
| 16 | 7 | 4 | 7 | 5 | 64% |
| 17 | 7 | 6 | 7 | 3 | 64% |
| 18 | 7 | 3 | 7 | 4 | 50% |
| 19 | 7 | 4 | 7 | 2 | 43% |
| 20 | 7 | 4 | 7 | 3 | 50% |
| 21 | 7 | 4 | 7 | 5 | 64% |
| 22 | 7 | 4 | 7 | 5 | 64% |
| 23 | 7 | 4 | 7 | 5 | 64% |
| 24 | 7 | 3 | 7 | 3 | 43% |
| 25 | 7 | 4 | 7 | 3 | 50% |

Results (2)

- Performance by fault number (and fault type) yielded more intriguing results
- Faults 2, 4, and 11 were always detected (type 3, wrong order)

- Fault type 1 (markings) frequently missed


| Pattern | Fault Number | Fault Type | Number of Trials | Number of Detects | Percent Detected |
|---------|--------------|------------|------------------|-------------------|------------------|
| A | 1 | 1 | 24 | 5 | 21% |
| A | 2 | 3 | 24 | 24 | 100% |
| A | 3 | 3 | 24 | 23 | 96% |
| A | 4 | 3 | 24 | 24 | 100% |
| A | 5 | 4 | 24 | 17 | 71% |
| A | 6 | 1 | 24 | 6 | 25% |
| A | 7 | 1 | 24 | 6 | 25% |
| B | 8 | 2 | 22 | 15 | 68% |
| B | 9 | 4 | 24 | 21 | 88% |
| B | 10 | 1 | 24 | 5 | 21% |
| B | 11 | 3 | 23 | 23 | 100% |
| B | 12 | 1 | 24 | 20 | 83% |
| B | 13 | 2 | 24 | 17 | 71% |
| B | 14 | 1 | 24 | 2 | 8% |

Marking errors (fault type 1) are more difficult to detect

Analysis (1)

- Binary logistic regression (Agresti, 2013) used to model the probability of detecting a fault

$$\log \left(\frac{\pi(\text{Err}(i), \text{Seq}(j))}{1 - \pi(\text{Err}(i), \text{Seq}(j))} \right) = \alpha_0 + \beta_i + \gamma_j$$

- Estimates for Pattern A
 - γ terms are all statistically **non-zero** and **positive**
 - Faults detected *less frequently* in the standard sequence:
 - A{JB}, or...
 - Pattern A first, then Pattern B with checklist

| Parameter | Estimate | Standard Error Estimate | Z-ratio | P-value |
|--------------------|----------|-------------------------|---------|---------|
| α_0 | -2.845 | 0.810 | -3.51 | 0.000 |
| $\gamma_{B\{JA\}}$ | 1.792 | 0.776 | 2.31 | 0.021 |
| $\gamma_{\{JA\}B}$ | 1.999 | 0.778 | 2.57 | 0.010 |
| $\gamma_{\{JB\}A}$ | 1.578 | 0.775 | 2.04 | 0.042 |
| β_3 | 4.967 | 1.218 | 4.08 | 0.000 |
| β_5 | 2.494 | 0.731 | 3.41 | 0.001 |
| β_6 | 0.251 | 0.710 | 0.35 | 0.724 |
| β_7 | 0.251 | 0.710 | 0.35 | 0.724 |

3-way interaction between sequence, checklist presence, and Pattern A



Analysis (2)

- Estimates for Pattern A
- Fault #3 (incorrect order) detected *more frequently* than the standard fault #1 (markings)
- Same effect for β_5 , which is a wrong orientation fault

| Parameter | Estimate | Standard Error Estimate | Z-ratio | P-value |
|--------------------|----------|-------------------------|---------|---------|
| α_0 | -2.845 | 0.810 | -3.51 | 0.000 |
| $\gamma_{B\{JA\}}$ | 1.792 | 0.776 | 2.31 | 0.021 |
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| β_6 | 0.251 | 0.710 | 0.35 | 0.724 |
| β_7 | 0.251 | 0.710 | 0.35 | 0.724 |

This suggests that Pattern A appears in the 3-way interaction because it has more marking errors



Fitted Model Validation

- No evidence for lack-of-fit in the model
- Formal tests (where $p > 0.05$ is significant):
 - Pearson: $p=0.171$
 - Deviance: $p=0.194$
 - Hosmer-Lemeshow: $p=0.725$
- Reasonable similarity between Estimated Probability of Detection and Observed Fraction of Detection
- However...

| Fault # | Sequence | Estimated Probability of Detection | Observed Fraction Detected |
|---------|----------|------------------------------------|----------------------------|
| 1 | A {JB} | 0.055 | 0.000 |
| 1 | B {JA} | 0.259 | 0.500 |
| 1 | {JA} B | 0.300 | 0.167 |
| 1 | {JB} A | 0.220 | 0.167 |
| 3 | A {JB} | 0.893 | 1.000 |
| 3 | B {JA} | 0.980 | 1.000 |
| 3 | {JA} B | 0.984 | 1.000 |
| 3 | {JB} A | 0.976 | 0.833 |
| 5 | A {JB} | 0.413 | 0.500 |
| 5 | B {JA} | 0.809 | 0.500 |
| 5 | {JA} B | 0.839 | 0.833 |
| 5 | {JB} A | 0.773 | 1.000 |
| 6 | A {JB} | 0.069 | 0.000 |
| 6 | B {JA} | 0.310 | 0.333 |
| 6 | {JA} B | 0.355 | 0.500 |
| 6 | {JB} A | 0.266 | 0.167 |
| 7 | A {JB} | 0.069 | 0.000 |
| 7 | B {JA} | 0.310 | 0.333 |
| 7 | {JA} B | 0.355 | 0.333 |
| 7 | {JB} A | 0.266 | 0.333 |

The probability of detection for each fault is **not** equal.



Finding (1)

- Created a testing methodology sensitive enough to detect differences in the effects on performance between:
 - Pattern sequence
 - Checklist presence
 - Pattern A
- If the *main effect* of a checklist on performance (of a concurrent dual verification task) were easily identifiable, then it would have been detected long ago



Finding (2)

- The assumption of average probability of detection between different types of error was ***empirically verified*** to be wrong
- Fault (Error) Types:
 1. Markings
 2. Incorrect piece(s)
 3. Wrong order
 4. Wrong orientation



Finding (3)

- Concurrent dual verification is not necessarily an effective control against defects, both **with** and **without** a checklist
- Verification techniques presented in the literature may be *conditional*, especially for specific types of errors (ie: markings)
- No JPA format is best for all circumstances
- Quality assurance tools must be well designed and well understood by **both** the designer and the user, in order to effectively control risk



Conclusions

- This is the first known research study to have examined:
 - The effect of a checklist on performance in a quality assurance setting
 - Subtle and complex interactions between JPA design, error types, and base error probability of detection
 - Probability of detection of different error types in the following context:
 - Quality Assurance (concurrent dual verification)
 - Use of a JPA, specifically a checklist
 - Simple assembly task



References

- Agresti, A. (2013). *Categorical Data Analysis* (3rd ed.). New York: John Wiley.
- DOE. (1993). *Guide to Good Practices for Independent Verification*. DOE-STD-1036-93. Washington, DC: United States Department of Energy Technical Standards Program.
- Fosshage, E. (2014). *The Effect of Job Performance Aids on Quality Assurance* (No. SAND2014-4762). Albuquerque, NM: Sandia National Laboratories.
- Gawande, A. (2010). *The Checklist Manifesto: How to Get Things Right*. New York: Picador.
- Kemeny, J.G. (1979). *The Need for Change, the Legacy of TMI: Report of the President's Commission on the Accident at Three Mile Island*. Washington, DC: U.S. Government Printing Office.
- Miller, R.B. (1953). *A Method for Man-Machine Task Analysis*. Technical Report 53-137. Wright-Patterson AFB, OH: Wright Air Development Center.
- NASA. (2010). *Safety and Mission Assurance Acronyms, Abbreviations, and Definitions*. NASA-STD 8709.22. Washington, DC: National Aeronautics and Space Administration.
- Shriver, E.L., Zach, S.E., and Foley Jr, J.P. (1982). *Test of Job Performance Aids for Power Plants*. No. EPRI-NP-2676. Alexandria, VA: Kinton, Inc.
- Smillie, R.J. (1985) Design Strategies for Job Performance Aids. In T.M. Duffy & R.W. Waller (Eds.) *Designing Usable Texts*, 213-241. Academic Press, Inc.