## A Filter-Map-Reduce Framework for Extracting Features from Log Files

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Work Product Final product that the learner produces. - Answer to question - Essay - Simulator state

Work Process Steps learner takes to get to (candidate) solution - Action sequence - Pauses - Event log

• Often have advice about problem solving strategy

- Science and Engineering Practices
- Writing and Revision



## Case Study: Physics Playground

#### Use knowledge of physics to get ball to balloon.



Adjust parameters (mass of ball, gravity, air resistance).

Draw objects (e.g., lever) on screen.

Players earn gold, silver or no trophy.

#### Learning support videos.



0.10 /25

#### Four-Processes Architecture



Figure 1: Four-Process Architecture for Assessments

- Context Determination (aka Activity Selection)
   Rules for what activity (task, problem, game
  - level, item) to do next
  - Rules for when to stop
- Evidence Capture (aka Presentation)
  - Converts physical events (e.g., mouse click) into logical events (e.g., move slider)
  - Logs logical events
  - Determines scoring context
- Evidence Identification
  - Summarizes events within a scoring context
  - Produces a vector of observeable outcomes for each scoring context
- Evidence Accumulation
  - Accumulates evidence across scoring contexts
  - Produces the scores



To: From: Topic: Date:

Context:

Body (Contains topic dependent data)

Signature & Watermarks



#### Scoring Contexts and Windows



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- Output of El is a row in a data table, with key (Learner ID, Task ID)
- Columns are observable outcomes

Goal is to have educators design these without needing programmers to implement them.

- 1. "Which trophy (if any) was earned?"
- 2. Did the player try to draw a lever?
- 3. How many times did the player adjust the gravity slider?
- 4. How much time did the player spend viewing support videos?



Most observables are meaningful in more than one scoring context.

Many observables are only meaningful in some scoring contexts.

Associate observables with sets of contexts:

• All Game Levels

- 1. Trophy
- 2. Time spent on learning supports
- Sketching Level
  - 2. Lever drawn
- Manipulation Level
  - 3. Gravity slider manipulation count



Algorithm for most observables can be partitioned into three phases:

Pre-filter Only observables related to the context are computed. Filter Remove events which are not relevant (e.g., "Ball moved") Map Extract relevant datum from event message. (e.g., 1 if gravity slider, 0 if not) Reduce Reduce the mapped values to a single outcome (e.g., sum(), count(), any())

Filter and mapping operations are inherently parallel and can exploit cluster computing. Reduction can be partitioned if the operator is commutative and associative.



#### A Graphical View of the algorithm



Three kinds of filtering:

- Certain observables only relevant in certain context sets (implicit filter).
- Only certain message topics are interesting
- Value of data field must be certain value or in a certain range.

Separating out the conditions allows optimizing computations.



Takes an event and returns value and timestamp.

Value could be:

- Constant
- Field from the data
- Transformed field from the data (common numerical operators)
- Conditional Rule:
  - If condition 1 then value expression 1
  - If condition 2 then value expression 2
  - Otherwise default value expression



Takes vector of values and generate final observable values.

- Logical summaries: any() or all()
- Numerical summaries: count(), sum(), mean()
- Ordinal summaries: min(), max()
- Position order: first(), last()

#### Example 1: What trophy did the player earn?

Filter: topic in {"Level Start", "Trophy Awarded"}
Map:

If topic == "Level Start" then none
If topic == "Trophy Awarded" then data.trophy

Reduction: max()



Note that the EC process does the classification of drawings as engines because it requires access to the physics engine.

Filter: topic in {"Agent Identified"}

Map:

- If data.agent == "Lever" then TRUE
- Otherwise FALSE

Reduction: any()



# Example 3: How many times did the player adjust the gravity slider

#### Filter:

```
• topic in {"Adjusted Slider"}
```

• data.control == "Gravity Slider"

Map: - Always 1

```
Reduction: count()
```



# Example 4: For how long did the player use learning supports

System logs start and stop times. Convert to number of seconds, and make start times negative. Then take sum.

Filter:

- topic in {"Learning Support Start", "Learning Support Stop"}
- o data.support in {<Physics Videos>}

Map:

If topic %contains% "Start" then -((int) timestamp)
If topic %contains% "Stop" then (int) timestamp)

Reduction: sum()



- **Goal**: Get algorithm directly from the specification by content author.
- Algorithm can be naturally partitioned to take advantage of parallel computing
  - Kafka Queuing algorithm
  - Apache Spark Distributed Computing
  - jq and MongoDB pipelines
- Watermarks allow system to reason about completeness of data.
- Specification, not yet implementation.
- May need labeling preprocessing for some applications (e.g., text analysis).



- Big issue is maintaining consistencey between EC, EI and EA processes.
- Need to better capture possible data fields output from EC.
- Maybe possible to work with analyses of audio/video recordings (typically NVivo).



#### Thanks



https://ralmond.net/EvidenceStream



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## Message Format (JSON)

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}

```
messageType: "ESEvent",
app: <Application GUID>,
uid: <Unique learner ID within app>,
window: <Scoring Context (Task) ID>,
timestamp: <ISO Time>,
sender: <ProccessName>,
topic: <Message Topic>| <verb>/<object>,
```

```
data: { <JSON OBJECT>},
```

```
processed: <logical>,
pError: [<error message>],
watermarks: {<mark>:<timestamp>,...}
signatures: [<encrypted checksums>]
```