Exploring Response Patterns in Problem-Solving Items Using Process Data

Insights from Log Files of PSTRE in PIAAC

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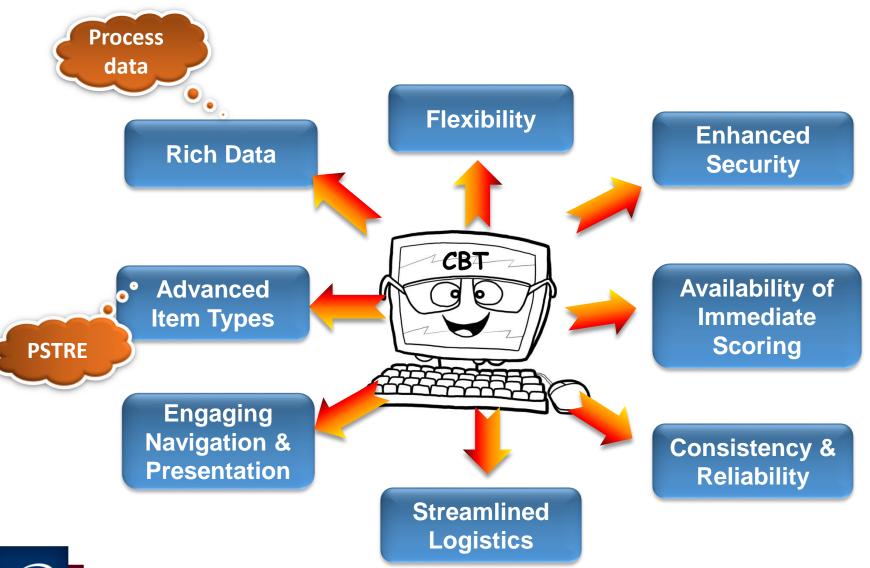


Outline

- Background
 - Computer-based testing and PSTRE items in PIAAC
 - Process data in log files
 - Motivation and research questions
- Exploring response patterns in PSTRE using process data
 - Sample and instrument
 - Methods: n-grams and sequential feature extraction model
 - Results and Discussions
- Conclusions and future work

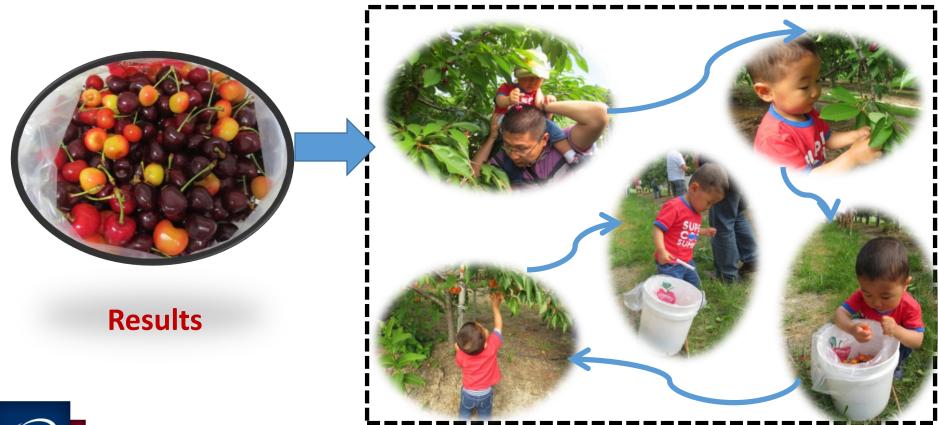


Benefits of Computer-Based Testing



Process Data from Log Files

 In CBTs, a variety of timing and process data accompanies test performance data. This means that much more than data is available besides correctness or incorrectness.



Log Files

```
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   <unitId><![CDATA[XYZ/Orientation-General]]></unitId>
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   <target><! [CDATA[MODULE]]></target>
   <event_name><! [CDATA[onItemBegin]]></event_name>
    <time><![CDATA[1395912781102]]></ti>
    <lang><! I Description | e of IR
 </event>
  <event>
   <moduleId><! [CDATA[platform]]></moduleId>
   <eventCounter><![CDATA[1]]></eventCounter>
   <unitId><![CDATA[XYZ/Orientation-General]]></unitId>
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   <itemId><![CDATA[GeneralOrientationQ01]]></itemId>
   <target><! [CDATA[MODULE]]></target>
    <event_name><! [CDATA[stimulusLoaded]]></event_name
    <time><![CDATA[1395912783070]]></time>
   <lang><! [CDATA[eng-IRL]]></lang>
  </event>
```

Our research draws on process data recorded in log files in the computer-based large-scale programs to address how sequences of actions recorded in problem-solving tasks are related to task performance.

studentId	eventCou	itemId	event_na	target	id	lang	time
840-51-01-003-00025	0	CS633Q00	onItemBeg	MODULE		eng-USA	1397193846084
840-51-01-003-00025	1	CS633Q00	stimulusLoaded	MODULE		eng-USA	1397193846545
840-51-01-003-00025	2	CS633Q00	QuestionLoaded	MODULE		eng-USA	1397193846570
840-51-01-003-00025	3	CS633Q00	StimulusAndQuestionLoa	MODULE		eng-USA	1397193846570
840-51-01-003-00025	4	CS633Q00	onItemEnd	MODULE		eng-USA	1397193850305
840-51-01-003-00025		CS633Q000	click	Ji .	next	eng-USA	1397193851030
840-51-01-003-00025	3	S 33 D 0			ess Da	tas	97193851104
840-51-01-003-00025	7	CS633Q000	QuestionLoaded	MODULE		eng-USA	1397193851283
840-51-01-003-00025	8	CS633Q000	stimulusLoaded	MODULE		eng-USA	1397193851427
840-51-01-003-00025	9	CS633Q000	StimulusAndQuestionLoa	MODULE		eng-USA	1397193851427
840-51-01-003-00025	10	CS633Q000	click	div	roof-color	eng-USA	1397193854737
840-51-01-003-00025	11	CS633Q000	click	span	stimulus_13	eng-USA	1397193855055
840-51-01-003-00025	12	CS633Q000	click	input	roofColorRadioRed	eng-USA	1397193855061
						1 172	



Motivations

- Obtain insights: how these action sequences are associated with different ways of cognitive processing and to identify key actions that lead to success or failure.
- Ongoing improvement: The results can be useful for test developers, psychometricians, and instructors to help them better understand what distinguishes successful from unsuccessful test takers and may eventually contribute to improved task and assessment design.



The Main Elements of PIAAC

Direct Assessment

- Numeracy
- Literacy
- Reading components
- Problem solving in technology-rich environments (PSTRE)

Module on Skills Use

Background Questionnaire

- Cognitive skills: rewriting, math and use of ICTs
- Interaction

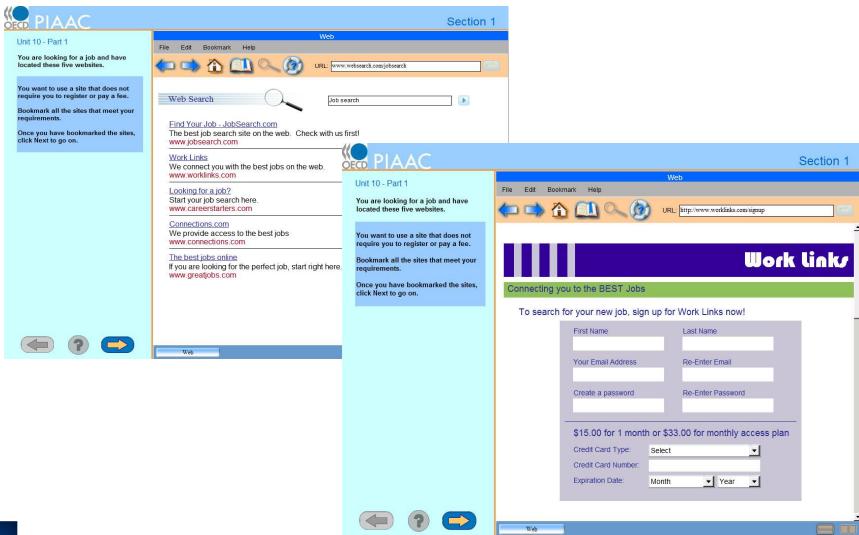
- **kills**: collaboration, planning,
- PSTRE items are used to assess the skills required to solve problems for personal, work, and civic purposes.
- Test takers have to set up appropriate goals and plans, and access and make use of information through computers and networks.
- More interactions are involved in the items.
- Available only in the computer-based path.
- Social and linguistic background

PIAAC PSTRE Sample Item (1)





PIAAC PSTRE Sample Item (2)





Research Objectives

Study Purposes:

- To extract and detect robust sequential action patterns that are associated with success or failure on one PSTRE item.
- To compare the extracted sequence patterns among selected countries.

Research Questions:

- How sequences of actions recorded in problemsolving tasks are related to task performance?
- Can the key actions / action patterns that lead to success or failure be identified?



Sample

Characteristics	Total	US	NL	JP
N	3926	1340	1508	1078
Correct (%)	2754 (70.1)	882(65.8)	1104 (73.2)	768 (71.2)
Incorrect (%)	1172 (29.9)	458 (34.2)	404 (26.8)	310 (28.8)
Gender				
Female	2025	629	711	526
Male	1901	711	629	552
Age (years)				
Mean (S.D.)	39.60	39.21	40.84	38.35
iviedii (3.D.)	(14.01)	(14.00)	(14.29)	(13.49)
Educational level				
Less than high school	615	124	401	90
High school	1493	534	590	369
Above high school	1812	680	513	619
Missing	6	2	4	0

Note. US, NL and JP represent the sample from the United States, the Netherlands and Japan.



Instrument: A PSTRE Item

- The task is to identify the ID number of a specified club member and send this number to a correspondent by email.
- Two environments are involved:
 - A spreadsheet environment that contains a database as the stimulus material that displays the information required to solve task.
 - An email environment to provide the response.
- The interim score is evaluated based only on the email responses.



Methods

```
Start, SS, SS_So, SS_So_1B, SS_So_OK, E, Next, FINALENDING

Start, SS, E, SS, SS_Se, SS_Type_FN, E, Next, Next_C, Next, FINALENDING

Start, Next, FINALENDING
```

- Similar structure between action sequences and languages.
- Motivated by the methodologies of natural language processing and text mining.
- Utilized feature selection models in analyzing the process data at a variety of aggregate levels.
- Evaluated the different methodologies in terms of predictive power of the evidence extracted from process data.



N-grams Model

I am happy to give a talk today.

unigrams

bigrams

trigrams

```
Unigrams (8) "START", "SS", "SS_Type_FN", "E", "E_S", "Next", "Next_OK Recode Next_OK, END into "FINALENDING"

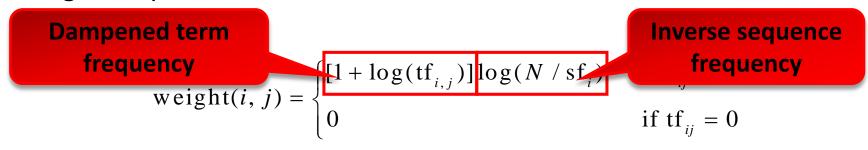
Bigrams (7) "START, SS", "SS, SS_Type_FN", "SS_Type_FN, E", "E, E_S", "L_S, Next", "Next_OK", "Next_OK, END"

Trigram (6) "START, SS, SS_Type_FN", "SS, SS_Type_FN, E", "SS_Type_FN, E, E_S", "E, E_S, Next", "E_S, Next, Next_OK", "Next_OK", "Next, Next_OK, END"
```



Term Weights

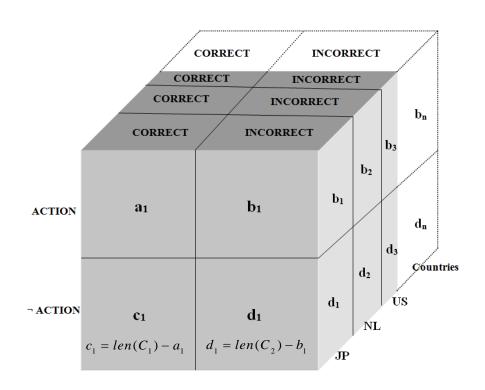
- An inverse sequence frequency was applied for attenuating the effect of actions that occurred too often in the collection to be meaningful.
- A dampened term frequency was also used to adjust the importance of an action with multiple occurrences in a single sequence.



- i, j action i in sequence j
- $tf_{i,j}$ frequency of action i in sequence j
- sf_i frequency of sequence that contains action i
- N number of sequences (test takers)



Feature Selection Models (1) Chi-square Feature Selection Model



$$\chi^{2} = \frac{M (ad - bc)^{2}}{(a+b)(a+c)(b+d)(c+d)}$$

$$c = len(C_{1}) - a$$

$$d = len(C_{2}) - b$$

$$M = a+b+c+d$$

The actions with **higher chi-square scores** are **more discriminative** in classification. Therefore, we ranked the chisquare score of each action in a **descending order**. The actions
ranked to the top were defined as the robust classifiers.



Feature Selection Models (2) Weighted Log Likelihood Ratio (WLLR)

 The product of probability of each action sequence and the logarithm of the ratio between conditional probability of the sequence in different performance groups.

$$WLLR(t, C_i) = P(t \mid C_i) \log \frac{P(t \mid C_i)}{P(t \mid \neg C_i)}$$
$$= P(t \mid C_i) \log \frac{P(t \mid C_i)}{Q(t \mid C_i)}$$

 $P(t \mid C_i)$ the conditional probability of action t in the class C_i $Q(t \mid C_i)$ the conditional probability of action t not in the class C_i

The higher the WLLR, the more likely the action belongs to class C_i .

Conversely, the lower the WLLR, the more likely the action belongs to class $\neg C_i$.



Results (1) Features of Actions by Performance Groups

		. ,				- C. C. P.
Robust Feat	tures of Actions and	Action Sec	quences Distinguishing Co	errect and	Correct group:	using tools such
	Unigram		Bigrams			ngine and sorting
	Actions	χ²	Actions	χ^2	with a clear su	b-goal
Correct	SS	70.72	E, SS	229.99	E, \$	272.49
	SS_Type_SN	68.04	SS, E	191.18	ATART, E, SS	226.42
	SS_So_OK	64.58	SS_So_OK, E	153.90	SS, E, E_S	211.37
	SS_So_1B	59.66	SS_So_1B, SS_So_OK	122.49	SS_So_OK, E, SS	150.25
Incorrect g	roup: hesita	ative	Type_SN, E	120.56	SS_So_1B, SS_So_OK	E 137.53
_	· · · · · · · · · · · · · · · · · · ·		Se, SS_Type_SN	98.21	SS, E, SS	133.85
benaviors (using "canco		So, SS_So_1B	84.43	SS_Se, SS_Type_SN, E	108.55
	SS_So_2A	1	START, SS_Se	70.03	SS_Type_SN, E, SS	108.20
Incorrect	Next_C	892.80	TART, Next	2416.20	START, Next, FINALE	NDING 2420.26
	SS_Save	98.90	Next, Next_C	521.74	Next, Next_0 *xt	478.16
	SS_Type_PGN	33.19	Next_C, Next	504.22	START, E, N	399.02
	SS_H	15.75	E_S, E_S	492.26	Next Nonresh	onse pattern:
	SS_So_3D	14.56	E_S, E	364.66	1 E S 1	
	SS_So_C		S SS	299.74	L, L_(ext, FINALENDING
	E_S In	correc	t group: using "	'Help"	(NONRES	SPONSE)
	CC T DC		a lot and aimle		S, E	338.26

function a lot and aimless save

the results in the server



Results (2) Country Level vs. Aggregate Level

Consistency Rate of Extracted Classifiers by Performance Groups Compared Between Country Level and Aggregate Level

М	ea	n	0.	.79

M	ea	n	=	0	7	1

20	US	Netherlands	Japan
Correct			
Unigrams	0.88	0.88	0.63
Bigrams	0.75	0.88	0.75
Trigrams	0.75	0.88	0.75
Incorrect			
Unigrams	0.63	0.63	0.63
Bigrams	0.63	0.88	0.88
Trigrams	0.75	0.63	0.75



Results (3) Features of Actions by Countries

Rol	bust Features of A	ctions and	d Action Sequences Across Cou	U	US: Double clicks or			
	Unigram	S	Bigrams			mail page		
	Actions	χ²	Actions	χ²		Than page		
US	Next_C	20.40	E, E	261.08	E, E, E		309.01	
	SS_Type_FN	15.64	START, Next	39.82	E, E, Next		278.87	
	E	13.25	Next, E	39.28	SS, E, E		132.21	
	SS_Type_PGN	10.14	START, E	38.97	START, E, E		85.14	
	SS_Save	6.22	SS_So_C, SS_Type_FN	37.63	SS_Type_FN, E, E		54.23	
NL	SS_Type_FN	315.30	SS_Se, SS_Type_FN	252.93	START, SS_Se_SS_	Time GN	226.67	
	SS_Type_GN	232.93	SS_Type_FN, SS_Type_FN	249.97	STAR NL:	More likely	use full	
	SS_Se	60.88	SS_Type_FN, E	203.30	SS_Type_FI nan	ne and given	names	
	SS_So_3B	31.59	SS_Se, SS_Type_GN	202.10	SS_Type_FI whe	en doing sea	rching	
	SS_So_2A	16.15	START, SS_Se	117.42	SS_Se, SS_Type_TTV		101.00	
JP	SS_Type_SM	383.58	SS_Type_SM, SS_Type_SM	308.58	SS_Type_SM, SS_Ty	ype_SM, SS_Type_S	SM 248.84	
	SS_Type_null	123.49	SS_Type C S_So	166.12	E_S, Next, Next_C		149.25	
	SS_Type_UM	70.75	F	137.22	SS_Type_SM, SS_So	o, SS_So_1B	149.21	
	P· Snelling r	nistak	es (optimal	116.73	SS_Type_SM, SS_Ty	ype_SM, SS_Sq	140.96	
	. •		st name and	115.33	SS_Type_SM, SS_Ty	ype_SM, E	116.15	
7	Juce Detvic							

ETS

last name)

JP: strategy changed

Results (4) Correlation between CHI and WLLR

Correlation between CHI and WLLR in Different Performance Groups by N-grams

	Correct	Incorrect
Unigrams	0.74	0.60
Bigrams	0.87	0.98
Trigrams	0.88	0.94

- The CHI and WLLR scores were moderately correlated in the unigrams and highly correlated in the bigrams and trigrams in both the correct and incorrect groups.
- It also proves that the mini-sequences (**bigrams and trigrams**) are more informative in process data analysis compared with single actions (unigrams).

Conclusions and Implications (1)

- With increasing use of computer-based assessments, process data play an increasingly important role in tracking test takers' thinking and action sequences, which is specially helpful in analyzing problem-solving items.
- The pilot study presented what we think is a promising method to analyze process data and extract robust sequence features that are informative for differentiating between performance groups.



Conclusions and Implications (2)

- The studies also demonstrate that process data can be useful in detecting nonresponse due to low engagement of test takers and checking item designs, especially in the field test.
- We explored these topics as part of the ongoing improvement of the software platform used in the PIAAC assessment for test delivery.



Future Studies

- The future study will focus on adapting existing methods for sequence data mining and develop a generalized toolkit for process data analysis.
- We recommend including background characteristics and timing data in the analysis of process data to further explore their interaction effects on performance.
- Explorations of how process data may inform adaptive testing appear to be a potential valuable research direction.



Acknowledgements

 We are thankful for all the supports from the Center for Global Assessment and Data Analysis Group in ETS for the helpful information and data preparations.



Related Publications

- He, Q., & von Davier, M. (2016). Analyzing Process Data from Problem-Solving Items with N-Grams: Insights from a Computer-Based Large-Scale Assessment. In Y. Rosen, S. Ferrara, & M. Mosharraf (Eds.) Handbook of Research on Technology Tools for Real-World Skill Development (pp. 749-776). Hershey, PA: Information Science Reference. Doi:10.4018/978-1-4666-9441-5.ch029.
- He, Q., & von Davier, M. (2015). Identifying Feature Sequences from Process Data in Problem-Solving Items with N-grams. In A. van der Ark, D. Bolt, S. Chow, J. Douglas & W. Wang (Eds.), Quantitative Psychology Research: Proceedings of the 79th Annual Meeting of the Psychometric Society (pp.173-190). New York: Springer. Doi: 10.1007/978-3-319-19977-1_13



Thank you very much!

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