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Adaptive Mobile Learning Framework Based On IRT Theory

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ABSTRACT

Advances in remote systems and skilled online tools have made great enthusiasm for adaptable learning lately. Patterns show that diverse mobile learning is the next achievement of meaningful innovation. In any case, adaptable mobile learning progresses through its different perspectives and the links between them must be integrated. In this article, we will conduct a baseline survey of specific writing with mixed-use learning in the light of the fact that there is still a need for extensive research on disrupting classroom innovation, particularly on how the performance of various tasks affects the instructor function in the classroom as a media coach and learning facilitator. The easily accessible writing identified with the adaptable mobile adaptation unequivocally highlights that the robust adaptable learning system is missing and that few scientists have explained the existence of a permanent structure. To address this problem, a systematic framework is proposed in this paper to discover diversity that adapts to the individual characteristics of students as described in each student's learning preparation. This paper also presents the theory of material responsiveness and how it is generally applied in a test case for any online course for the time of flexible exams within the course [for the development of model information] of elements (evaluation questions), which are available in online vaults. Being a probability theory, the article depicts the variables to consider and how to detect probability that a student answers an appropriately defined element, known as the level of absorptive capacity, taking into account the prediction test and their previous answers.

Key words : Adaptive learning, content, context, mobile learning, item response theory

1. INTRODUCTION

The assessment process in one of the most noteworthy issues in the learning methodology and all the things considered is the technique that describes the game plan for direction because it measures the introduction of the introduction into the information system. In previous decades, the standardization of TICs in teaching learning methods has urged to address a variety of features of guardians and teachers. Improved learning methods have made it possible to change the different learning and teaching strategies that exist together in the learning framework at the time of making changes similar to the customer presentation process.

Flexible learning is the most recent model. This system seeks to modify each individual by displaying educational materials according to the characteristics of students with information [1]. Adaptive adaptation must be created by thinking that customary e-learning frameworks are not suitable for individual instruction; they provide a similar material for all experiences, [2]. The weaknesses of traditional frameworks become evident when these frameworks are used in educational situations where teaching aids fluctuate, and previous information is widely advanced, with different learning styles and different inclination. According to [3], during an e-learning session, we can adjust: communicate, transfer course, reveal content and meet together, and ultimately support joint efforts:

1. Flexible interaction suggests modifications to the interface structure and proposes to activate or support client connectivity to the system.

2. Flexible delivery of the course suggests modifications required for the design of a training course (or, occasional movement in courses) for an individual. The aim is to fix the "fit" between the course material and the characteristics / needs of the clients.

3. Content Discovery and Assembly involves the use of adaptive methods of exposure to convergence with learning materials / "content" from possibly sources / stores.

4. Supporting adaptive cooperation It is planned to capture adaptable assistance in forms of learning that include interpersonal correspondence (and, on this basis, social cooperation), and it is envisaged to make a concerted effort towards normal goals. Access to remote stations with high

transport capacity - for example, 3G media transfer frame, remote LAN, and the emergence of portable gadgets - has opened new open doors available for training. The true ability of e-learning as "whenever, anywhere" has finally begun to recognize the mobile learning approach (elearning).

2. METHODOLOGY

The target of this examination is to build up a versatile portable taking in structure from the accessible writing by utilizing Item Response Theory (IRT) to address the center viewpoints and attributes of versatile portable learning. This underlying arrangement is the beginning square for the further improvement of the versatile portable learning structure. The premise of proposing the system is writing, so different versatile portable learning structures, designs and models were looked, sieved and chose for building up a versatile portable learning system. The snowballing procedure was utilized for this reason. In this method, the information have been looked by following references in the lists of sources of different related articles [4]. Watchwords like versatile portable learning structure/engineering/model were utilized to discover differing and fascinating versatile portable learning systems.

3. LITERATURE REVIEW

To be particular created by means of [5]. Regardless, in this one we centered our evaluation on the overview of metaindicative examinations about m-learning dispersed due to the fact that 2010 in order to show a set of the momentum inquire about in this field. The words used in glancing thru articles in the SCOPUS database were m-learning and metaanaly* and we restrained the request to articles from 2010 to 2019. A sorted out assessment was once coordinated, establishing with a fundamental overview of the quantity of changed works, trailed with the aid of a via and through study of picked articles as showed by means of the noteworthiness of the journals whereby they were appropriated. This assessment suggests that the m-learning is a rising subject of research, exhibiting a tireless addition to the extent quantity of disseminations considering the commencing of the multi decade. Some meta-efficient papers in the in the past years [6]; [7]; [8]; [9]; [10] show this development and divulge the consideration on concentrates related to the feasibility of m-learning sought after via mlearning gadget structure.

As tested via these makers, "e-learning investigation is at the early lion's offer stage and foci have moved from differentiating the reasonability of e-learning with making models for e-learning conditions and for educating and getting to know frameworks interior quite a number elearning circumstances.

On the off risk that m-learning articles pursue a comparative way, we may also assume extra lookup examines on methodologies and system (...) soon." [8].

Nonetheless, one of the most important issues that basically every examination in this subject tries to set up is a constant which means of m-learning. Being a relative new area of find out about and seeing some innovative leaps ahead in its initial presence, a few definitions have been recommended since the mid 2000s. For example, a few creators identified m-learning as a attribute outcome of the e-learning improvement [12] but later definitions position m-learning as a strategy that converges transportable figuring and elearning , that receives the utilization of versatile innovation to accomplish whenever, anyplace, omnipresent mastering and that underlines students' portability and custom-made learning [13].

As a ways as primary bibliometric data,[10] surveyed learns about m-learning disbursed in six full-size research diaries identified with innovation extended gaining from 2001 to 2010 and designated that from 2006 to 2010 the extent of articles recognized with MUL (Mobile and Ubiquitous Learning) nearly quadrupled in connection to the 2001–2005 period. These figures can be more suitable by the ones created by using [8] and [11] that directed meta-expository surveys of m-taking in patterns from 2003 to 2008 and from 2010 to 2015. In spite of the truth that these creators utilized specific diary databases, the outcomes moreover current a parallel development diagram on account of the [8] find out about and a gradually unassuming yet at the equal time evolutional sketch as a ways as variety of distributions in the timeframe published in the [11] writing survey.

As a ways as test gatherings, both [10] and [9] report that dispensed papers reveal a excessive predominance of concentrates with Higher Education understudies, trailed through grade school understudies and K-12 understudies. Strangely (or perhaps not, as we will contend further) just a couple of concentrates in both meta-investigations were linked with the utilization of m-gaining from the educators or instructors' point of view.

As to instructive settings of m-learning contemplates, [11] document that when those settings are uncovered, casual learning settings are transcendent, trailed via formal settings and a combination of both. This consequence is predictable with outcomes announced with the aid of exclusive creators, in specific with the aid of [13].

Together with the incidence of informal instructive settings in m-learning allotted research, [10] likewise record that most investigations do not listen on a precise gaining knowledge of space but rather current results recognized with inspiration, recognitions and frames of thinking of understudies towards m-learning. By and by, the recognitions and frames of thought of instructors are only right here and there found. These effects are lined up with the ones likewise introduced by using [11] in a later examination. As far as results, [9] document that 86% of the examinations on m-learning current nice results. This type of result is likewise found in [11] article, which document that "the greater phase of the 144 M-Learning concentrates current effective results. (...) Neutral result located straightaway and terrible end result located the least." These results are pretty illustrative of a frequent inspirational frame of idea in the direction of m-discovering that crosses a extensive part of the writing in this field.

4. ALGORITHMIC ADAPTIVELY

This integrative writing audit pursues a portion of our past work around there, to be specific created by [5 regardless, in this one we focused our assessment on the review of metademonstrative examinations about m-learning circulated since 2010 to show a set of the momentum inquire about in this field. The watchwords used in glancing through articles in the SCOPUS database were m-learning and meta-analy* and we confined the request to articles from 2010 to 2017. A composed review was coordinated, beginning with a hidden study of the extensive number of adjusted works, trailed by a through and through study of picked articles as showed by the importance of the journals wherein they were dispersed. This overview shows that the m-learning is a rising field of research, exhibiting a constant augmentation to the extent number of disseminations since the beginning of the multi decade. Some meta-methodical papers in the earlier years [6]; [7]; [8]; [9]; [10] show this development and reveal the consideration on concentrates related to the reasonability of m-learning sought after by m-learning system structure [11]. These examples and foci are somewhat expected, in light of the fact that they reveal a foreseen route since the colleague of an advancement with its appointment and blend. As showed by these makers, "e-learning investigation is at the early lion's offer stage and foci have moved from differentiating the practicality of e-learning with making models for e-learning conditions and for teaching and learning frameworks inside various e-learning circumstances. In case m-learning articles seek after a near way, we may expect more research looks at on strategies and framework (...) soon." [8]. In any case, one of the principle gives that fundamentally every assessment in this field endeavors to set up is a relentless importance of m-learning. Being a relative new field of study and seeing some inventive jumps forward in its underlying nearness, a couple of definitions have been prescribed since the mid 2000s. For instance, a couple of makers perceived m-learning as a trademark result of the e-learning improvement [12] yet later definitions position m-learning as a methodology that combines compact figuring and e-learning, that gets the use of flexible development to achieve at whatever point, wherever, inescapable learning and that underlines understudies' compactness and modified learning [13].

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2010 and detailed that from 2006 to 2010 the quantity of articles identified with MUL (Mobile and Ubiquitous Learning) nearly quadrupled in connection to the 2001–2005 period. These figures can be enhanced by the ones created by [8] and [11] that directed meta-expository surveys of m-taking in patterns from 2003 to 2008 and from 2010 to 2015. In spite of the fact that these creators utilized distinctive diary databases, the outcomes additionally present a parallel advancement design on account of the [8] study and a progressively unassuming yet at the same time evolutional design as far as number of distributions in the timeframe revealed in the [11] writing survey.

As far as test gatherings, both [10] and [9] report that distributed papers demonstrate a high predominance of concentrates with Higher Education understudies, trailed by grade school understudies and K-12 understudies. Strangely (or maybe not, as we will contend further) just a couple of concentrates in both meta-investigations were connected with the utilization of m-gaining from the educators or instructors' point of view.

As to instructive settings of m-learning contemplates, [11] report that when those settings are uncovered, casual learning settings are transcendent, trailed by formal settings and a blend of both. This outcome is predictable with results announced by different creators, in particular by [13].

Together with the prevalence of casual instructive settings in m-learning distributed research, [10] likewise report that most investigations don't concentrate on a specific learning space but instead present outcomes identified with inspiration, recognitions and frames of mind of understudies towards m-learning. By and by, the recognitions and frames of mind of instructors are only here and there found. These outcomes are lined up with the ones likewise announced by [11] in a later examination.

At long last, as far as results, [9] report that 86% of the examinations on m-learning present positive results. This sort of result is likewise found in [11] article, which report that "the greater part of the 144 M-Learning concentrates present positive results. (...) Neutral result positioned straightaway and negative result positioned the least." These outcomes are exceptionally illustrative of a general inspirational frame of mind towards m-discovering that crosses a significant part of the writing in this field.

p(L1)ku=p(L0)k,

 $p(Lt+1|obs = correct)ku = p_{t}L_{t})^{k}u'(1 - p(S)^{k}) / - p(S)k) + (1 - p(Lt)ku) \bullet p(G)k \quad (1b)$

 $p(Lt+1|obs = wrong)ku= p(Lt)ku\bullet p(S)k / (S)k + (1 - p(Lt)ku) \bullet (1 - p(G)k)$ (1c)

$$p(L_t+1)k_u = p(L_t+1)bs^k u + (1 - p(L_t+1)bs^k u) p(T)^k$$
, (1d)

Ammar Wisam Altaher et al., International Journal of Advanced Trends in Computer Science and Engineering 8(5) September - October 2010 2647-2652

 $p(Ct+1)ku=p(Lt)ku \bullet (1 - p(S)k) + (1 - p(Lt)ku) \bullet p(G)k(1e)$

In the standard BKT model, we use one copy of each of the above four

parameters h p(L0), p(T), p(S), p(G) i per skill. BKT models are usually fit using

the expectation maximization method $\left(EM\right)$, Conjugate Gradient Search ,

or discretized brute-force search [16].

Exhibits Modeling understudy learning as an idle variable is a prominent methodology. The dormant variable is refreshed dependent on the rightness of the watched understudy chances to apply the ability being referred to. When all is said in one case, this demonstrating approach is known as a Hidden Markov Model. An uncommon instance of the methodology is known as Bayesian Knowledge Tracing (BKT) [17]. BKT accept that understudy information is spoken to as a lot of paired factors - one for every aptitude (the ability is either aced by the understudy or not). Perceptions in BKT are additionally parallel: an understudy gets an issue [step] either right or wrong. BKT has a long history of being effectively utilized in Intelligent Tutoring Systems (ITS) with regards to authority learning and issue sequencing. In its standard execution that is still in dominating use today, BKT just has aptitude explicit parameters. Beginning with the first distribution on BKT [17] and including later works (for example [18]), there exist solid markers that BKT models (regularly called individualized BKT models) that by one way or another record for understudy fluctuation are better than the standard BKT model.

Another known framework is Item Response Theory (IRT), developed in the field of psychometrics to model the interaction of a learner with discrete items.

Item Response Theory is an evaluation method used to calculate mental attributes in the development and analysis of instructive and behavioral evaluations (performance measures, rating scales, inventories, different or instruments).that measure mental qualities. This is increasingly becoming popular among instructive assessment experts to evaluate psychological estimation, yet little is thought of as the principle to break down non-cognitive estimation. (For example, personality, frame of mind and psychopathology) a paper by [19] provides data on models, presumptions and use of IRT in psychological and nonsubjective assessments. The [19] infer that, for sociological, mental and psychopathological evaluations, IRT is a superior structure that can be abused by scientists to examine intellectual information for assessment and evaluation, and non-subjective information. All the factual assumptions must be met, however, and the test information must be in line with the IRT model for legitimate, reliable and tenaciousness



Figure 1: Adaptive mobile framework with IRT model test data

A number of parameters may be used when estimating the ability of a person using IRT:

• The 1 parameter logistic model (1PL) also known as the Rasch model, only uses item difficulty as a parameter for calculating a person's ability.

• The 2 parameter logistic model (2PL) uses both item difficulty and item discrimination (the extent which the item is measuring the underlying psychological construct) as parameters.

• The 3 parameter calculated model (3PL) utilizes item trouble, item separation and the degree which competitors can figure the right answer, as parameters.

Not with standing the model utilized, IRT based tests have various focal points over old style test theory based tests. IRT permits item banking, which implies that competitors would all be able to be given a totally extraordinary arrangement of items, yet at the same time give a similarly exact gauge of capacity. Likewise, IRT permits the utilization of versatile testing, tests which tailor the trouble of the test to every individual competitor. Progressively, psychometricians are going to IRT based models to configuration explore and distribute psychometric tests, with more research on IRT distributed each year. Additionally, the greater part of the significant test distributers, for example, SHL and Saville counseling are utilizing IRT based techniques in their test improvement forms.

4.1. Model Setup

• Consider a test with p binary (correct/incorrect) responses Each item is assumed to 'reflect' one underlying (latent) dimensions of 'achievement' or 'ability' Start with an assumed 1-dimensional test, say of swim attitude mathematics with 10 items. Now we will try to get a value (score) on the mathematics scale from a set of 40 (1/0) responses from each individual?

First some basic notation...

• fj is the latent (factor) score for individual j. $\pi i j$ is the probability that individual j responds correctly to item i. Then a simple item response model is: $\pi i j = ai + bi fj$ Just like a simple regression but with an unobserved predictor, Classical item analysis

• Can be viewed as an item response model (IRM) – $\pi i j = ai + bi f j$

The maximum likelihood estimate of (red used for a random variable) is given by the 'raw test score'
, ai is the item difficulty and bi is the item discrimination
Instead of using a linear linkage between the latent variable and the observed total score, IRT uses a logit transformation log(πij /1-πij) = logit(πij) = ai + bi fj

Parameter Interpretation

• Difficulty

Point on the theta continuum (x-axis) that corresponds to a 50% probability of endorsing the item
A more difficult item is located further to the right

than an easier item
Values are interpreted almost the reverse of CTT

Values are interpreted annost the reverse
Difficulty is in a z-score metric

Difficulty is in a
Discrimination

the slope of the IRF

- The steeper the slope, the greater the ability of the item to differentiate between people

Assessed at the difficulty of the item



Figure 2: Application of IRT in Measurement in 21st Century

Figure 2 shows the application of IRT in measurement in 21^{st} century. Utilization of IRT is vital in estimation network; it is material in quantities of zones as it has defeated various unrealistic mountains which CTT can't survive.

Proof from writing uncovers that it is pertinent not just in instructive testing which is increasingly recognizable by scientists; it is material in breaking down non-psychological information. [20] asserts That Item Response Theory (IRT) is utilized in a variety of controls together with human science, political theory, intelligence science, human improvement, business, and correspondences, simply as in preparation where it commenced as strategy for the investigation of instructive tests. [21] Uncovers that assessments and polls count on a pressing job in intellectual evaluation. Both subjective measures (e.g., insight tests) and non-intellectual measures (e.g., disposition polls, character surveys) have a location with the practitioner"s toolbox in a number of fields of Genius research. For instance, in work pressure dedication methodology other than insight testing, frequently persona surveys are utilized to consider whether an up-and-comer is fabulous for a specific activity.

In instructive trying out IRT has been being used for a long time with reliable outcomes. As a foundation in estimation it very properly may also be utilized in test development, grid inspecting, it serves to improve the nature of the exams and scales produce, take a look at comparing and organization, it builds up a comprehension about differential item working, automated versatile testing, wellbeing numeracy and test scoring and perception amongst others. All in all, IRT is a most beneficial machine that can be misused through analysts in breaking down subjective information in appraisal, evaluation appear into and non-intellectual information, in sociological, mental, psychopathology reviews However, all the measurable suspicions must be met, and the test facts ought to suit the IRT model for legitimate, reliable and tenable outcomes.

5. CONCLUSION

IRT takes into account dynamic testing in which a test is more or less difficult, depending on the competitor's presentation, to match the test to its potential. IRT offers the test manufacturer, the consumer and the up-and-comer bunch of advantages, and IRT-based models are slowly becoming prevalent within psychometric. IRT attempts to demonstrate an analyst's ability of noting a test item effectively dependent on the example of responses to the items that comprise a test. IRT can evaluate the parameters of an item that are autonomous of the qualities of both the test takers to which it is discovered and the various items that make up the test. For making forecasts, four unmistakable conditions called 1PL, 2PL, 3PL and 4PL (strategic parameter) models are used directly. These models are the foundation of IRT; they are the turns on which the theory relies and they reveal information about the dormant conduct of the items and the examinee which make it simple for estimation network to make right expectations.

REFERENCES

[1] Surjono H.D, 2011, The Design of Adaptive E-Learning System based on Student'sLearning Styles International Journal of Computer Science and, Vol. 2 (5), 2011, 2350-2353, ISSN InformationTechnologies:0975-9646

 [2] Brusilovsky, P. 2001. Adaptive hypermedia. User Modeling and User AdaptedInteraction, 11, 87-110.
https://doi.org/10.1022/A:1011142116206

https://doi.org/10.1023/A:1011143116306

[3] Paramythis, A., & Loidl-Reisinger, S. 2004.Adaptive learning environments andeLearning standards.http://www.fim.uni-

linz.ac.at/staff/paramythis/papers/ecel2003.pdf.

- [4] Jalali, S., & Wohlin, C. 2012. Systematic Literature Studies: Database Searchers vs. Backward Snowballing. In Proceedings of ACM-IEEE International Symposium on Empirical Software Engineering and Measurement (ESEM) 2012, 29–38.
- [5] Aresta, M., Pedro, L., & Santos, C. (2015). Mobile
- Learning and Higher Education: A Theoretical Overview.
- Journal of Mobile Multimedia, 1(1/2), 147–156.

[6] Chee, K. N., Yahaya, N., Ibrahim, N. H., & Noor Hassan, M. (2017). Review of mobile learning trends 2010-2015: A meta-analysis. Educational Technology & Society, 20(2), 113–126.

[7] Al-Zahrani, H., & Laxman, K. (2016). A critical metaanalysis of mobile learning research in higher education. The Journal of Technology Studies, 42(1).

https://doi.org/10.21061/jots.v41i2.a.1

[8] Hung, J. L., & Zhang, K. (2012). Examining mobile learning trends 2003–2008: A categorical meta-trend analysis using text mining techniques. Journal of Computing in Higher Education, 24, 1–17.

[9] Wu, W., Wu, Y., Chen, C., Kao, H., Lin, C., & Huang, S. (2012). Review of trends from mobile learning studies: A meta-analysis. Computers & Education, 59, 817–827.

[10] Hwang, G., & Tsai, C. (2011). Research trends in mobile and ubiquitous learning: A review of publications in selected journals from 2001 to 2010. British Journal of Educational Technology, 42(4), 65–70.

[11] Chee, K. N., Yahaya, N., Ibrahim, N. H., & Noor Hassan, M. (2017). Review of mobile learning trends 2010-2015: A meta-analysis. Educational Technology & Society, 20(2), 113–126.

[12] Georgiev, T., Georgieva, E., & Smrikarov, A. (2004). M-learning -- a new stage of E-learning. Paper presented at the International Conference on Computer Systems and Technologies CompSysTech'2004. IV.28-1-5.

https://doi.org/10.1145/1050330.1050437

[13] Vázquez-Cano, E. (2014). Mobile distance learning with Smartphones and apps in higher education. Educational Sciences: Theory & Practice, 14(4), 1505–1520.

https://doi.org/10.12738/estp.2014.4.2012

[14] Corbett, A. T.; Anderson, J. R. (1995). "Knowledge tracing: Modeling the acquisition of procedural knowledge". User Modeling and User-Adapted Interaction. 4 (4): 253–278.

[15][^] Jump up to:a b Yudelson, M.V.; Koedinger, K.R.; Gordon, G.J. (2013). "Individualized bayesian knowledge tracing models". Artificial Intelligence in Education.

[16] Baker, R. S. J., Corbett, A. T., and Aleven, V.: More Accurate Student Modeling through Contextual Estimation of Slip and Guess Probabilities in Bayesian Knowledge Tracing. In: Woolf, B.P., A⁻imeur, E., Nkambou, R., Lajoie, S.P. (eds.)Proceedings of the 9th International Conference On Intelligent Tutoring Systems (ITS 2008), LNCS vol. 5091, pp. 406-415. Springer (2008)

[17] Corbett, A. T. and Anderson, J. R.: Knowledge tracing: Modeling the acquisition of procedural knowledge. User Modeling and User-Adapted Interaction, 4(4), 253-278. (1995)

https://doi.org/10.1007/BF01099821

[18] Pardos, Z. A. and Heffernan, N. T. (2010). Modeling Individualization in a Bayesian Networks Implementation of Knowledge Tracing. In: Paul De Bra, Alfred Kobsa, David N. Chin (eds.) Proceedings of the 18th International Conference on User Modeling, Adaptation, and Personalization (UMAP 2010), LNCS vol. 6075 pp. 255-266. Springer (2010)

https://doi.org/10.1007/978-3-642-13470-8_24

[19] Ogunsakin, I. B., & Shogbesan, Y. O. (2018). Item Response Theory (IRT): A Modern Statistical Theory for Solving Measurement Problem in 21st Century. International Journal of Scientific Research in Education, 11(3B), 627-635. Retrieved [DATE] from http://www.ijsre.com.

[20] Templin, J. (2012, July 9-13). Item Response Theory. Colorado, Georgia, USA Van der Linden, W. J., & Hambleton, R. K. (Eds.). (1997). Handbook of modern item response

theory. New York: Springer.

[21] Egberink, I. L. (2010). Applications of item response theory to non-cognitive data Groningen: s.n.