

**Research Paper** 

# Studying Students' Knowledge of the Benefits, Challenges, and Applications of Big Data Analytics in Healthcare

Elham Nazari<sup>al</sup>, Maryam Edalati Khodabandeh<sup>a</sup>, Ali Dadashi<sup>a</sup>, Tahmineh Aldaghi<sup>b</sup>, Marjan Rasoulian<sup>a</sup>, Hamed Tabesh<sup>a</sup>

<sup>a</sup> Department of Medical Informatics, Faculty of Medicine, Mashhad University of Medical Sciences, Mashhad, Iran

<sup>b</sup> Department of Industrial Engineering, Tarbiat Modares University, Tehran, Iran

#### ARTICLE INFO ABSTRACT Received: 03 January 2022 The purpose of this study was to evaluate the students' familiarity from different universities of Mashhad with the benefits, applications and challenges of Big Data Reviewed: 21 January 2022 analysis. This is a cross-sectional study that was conducted on students of different fields, including Medical Engineering, Medical Informatics, Medical Revised: 07 February 2022 Records and Health Information Management in Mashhad-Iran. A questionnaire Accepted: 20 February 2022 was designed. The designed questionnaire evaluated the opinion of students regarding benefits, challenges and applications of Big Data analytics. 200 students **Keywords:** participated and participants' opinions were evaluated descriptively and analytically. Most students were between 20 and 30 years old. 43.5% had no work Big Data, Benefits, Challenges, experience. Current and previous field of study of most of the students were HIT, Analysis, Healthcare, Knowledge HIM, and Medical Records. Most of the participants in this study were undergraduates. 61.5% were economically active, 54.5% were exposed to Big Data. The mean scores of participants in benefits, applications, and challenges section were 3.71, 3.68, and 3.71, respectively, and process management was significant in different age groups (p=0.046), information, modelling, research, and health informatics across different fields of studies were significant (p=0.015, 0.033, 0.001, 0.024) Information and research were significantly different between groups (p=0.043 and 0.019), research in groups with / without economic activity was significant (p= 0.017) and information in exposed / non-exposed to Big Data groups was significant (p=0.02). Despite the importance and benefits of Big Data analytics, students' lack of familiarity with the necessity and importance is significant. The field of study and level of study does not appear to have an effect on the degree of knowledge of individuals regarding Big Data analysis. The design of technical training courses in this field may increase the level of knowledge of individuals regarding Big Data analysis.

<sup>1</sup> Corresponding Author Nazarie4001@mums.ac.ir

### **1. Introduction**

Today, with the advent of various technologies, a huge amount of data that is known as Big Data in being generated especially in healthcare. Big data analytics has become a hot topic and has been the focus of many academic communities and the subject of many students' research (Achariya & ahmed, 2016, Alharthi et al., 2017). This type of data has features such as high volume and diversity and due to these features, they cannot be managed and analysed using conventional hardware and software. Analytics for analysing Big Data are known as Big Data Analytics and have many benefits including useful data pattern discovery and important features extraction (Nahr et al., 2021, Nazari et al., 2021). This analysis has many applications in various medical and insurance industries (Archenaa & Anita, 2015). In addition to the many benefits of these analytics, there are challenges that if ignored, the results will change, such as a lack of expert staff, lack of familiarity with the tools and methods required, data type, security issues, budget and etc (Gharachorloo et al., 2021, Manogaran et al., 2017). Understanding the benefits, challenges, and applications of this area can be helpful in conducting useful and efficient research (Belle et al., 2015, Nozari et al., 2021). Due to the importance of Big Data analysis in various industries and the fact that students and their research are related to industry and applied research, this field in Iran is in the early stages of research and unfamiliar with the concepts is severely felt. The purpose of this study is to investigate students' familiarity with the different Benefits, applications, and challenges of Big Data.

### 2. Method

This cross-sectional study was designed for 200 students of Ferdowsi University and Mashhad University of Medical Sciences. Mashhad is the largest city in eastern Iran with a population of about three million, located on the border with Afghanistan and Turkmenistan on the Silk Road. Mashhad has two major universities, Ferdowsi and Medical Sciences, which students in engineering and basic sciences study at Ferdowsi University and students in medical sciences such as medical Records, Health Information Management and Medical Informatics study at Mashhad University of Medical Sciences.

A questionnaire was designed to assess the level of the knowledge of students in Mashhad universities about the benefits, applications and challenges of Big Data analysis. The questionnaire contains close-end questions with a five-point Likert scale. The basic items of the questionnaire were based on literature searches in Google Scholar, Science Direct and EMBASE databases and were designed and validated by the Delphi method with the participation of 10 experts from various fields (Medical Informatics, Biostatistics, HIT and Computer Science). The questionnaire was designed in the form of 3 general items of benefits, applications and challenges. Benefits included information with 5 questions, modelling with 3 questions, data with 5 questions, and process management with 6 questions. Application questions consisted of health service delivery with 17 questions, research with 4 questions, health, information with 16 questions, essential medicine with 15 questions, health financial with one question, leadership and governance with 6 questions and challenge included 9 questions. The questions are listed in Table 1:

Table 1. Questions			
Items	Questions	Category	Subcategory
			Generating new knowledge
	In your opinion, which		Sharing information
Advantages	advantages are related	Information	Displaying and summarizing information
0	with Big Data analysis?		Extracting information and delivery for better results
	0 ,		Using meaningful information
			Predicting disease epidemics
		Modeling	Increasing confidence
		Wodening	Discovering and exploring behavioral pattern or activities
			Decreasing ambiguity
		Dete	Increasing reliability
		Data	Reducing uncertainty
			Improving data quality
			Managing massive volumes of data
			Improving clinical trial quality
			Improving operational efficiencies
		Process	Interpreting easiness
		management	Improving entity detection
			Managing communications that are seemingly unrelated
			Improving the ability of intelligent systems
			Disease screening
			public health
			Disease earlier diagnosis
			Patient-centered services
			Therapeutic approaches improvement
			Surgery
			Rehabilitation
	In your opinion, Which applications are related to Big Data analysis?	1.Health	
Applications		Service	Clinical operations analysis
rippileadons		Delivery	Primary care
		Denvery	Readmissions management
			Health care delivery
			Disease management
			Cause of disease detection
			Decompensation management
			Blood transfusion management
			Triage management
			Health care data management
			Prediction
		2.Research	Disease pattern analysis
		2.1105001011	Side effects discovery
			Research& development& Innovation
		ľ	personalized medicine
			PHR(Personal Health Record) and HER
			EBM (Evidence Base Medicine)
			Patient monitoring
			Web and social media
			IOT(Internet Of Things)
			Semantic standards
		3.Health	Biometric
		Information	Patient profile analytics
			1 7
			CPOE (computerized physician order entry)
			Health informatics
			Coding management
			IT infrastructure management
			Quality measurement
			Bioinformatics and genetics
L			Comorbidity Discovery, Adverse events Discovery
			Diagnosis
	•	•	· -

#### Table 1. Questions

			Precision medicine
			CDSS(Clinical Decision Support System)
			Sensor processing
			RFID(Radio-Frequency identification)
			Signal processing
			Drug discovery & clinical Research
		4.(Essential)	Vision augment
		Medicines	GPS(Global Positioning System)
			Telemedicine, E-health, Remote healthcare system
			Mobile health
			Information Support
			Image processing
			BCI(Brain Computer Interface) and smart home
			Recommender systems
		5.Health Financing	Cost Reduction & Insurance service
			R & D in medications
		6.Leadership	Hospital quality monitoring
		and	Resource management
		Governance	Resource management
			Operational management
			Business and organizational and Strategic management
			Lack of knowledge about appropriate for the purpose
			Lack of IT infrastructure
	In your opinion, what		Lack of expertise about appropriate tools and algorithms
C1 11	challenges there are in		Variable and scalable data
Challenges	big data analysis		Lack of data quality
			Data uncertainty and missing data
			Unstructured data
			Security and privacy issue
			High cost

The validity and reliability of the questionnaire were confirmed by the presence of 10 validity experts and the reliability was confirmed by Alpha Cronbach's 92.1%. The questionnaires were then distributed to 200 students. Students of Medical Engineering, Medical Informatics, Medical Records and Health Information Management participated in the study. Data were collected to ensure that participants answered all the questions. 200 questionnaires were completed. Data entry and analysis were performed using EXCEL (v. 2007) and SPSS (v. 21).

### 3. Results

For this study, 200 students participated and the results are shown in Table 2.

Variables	Items	Frequency (percentage) of student (n=200)
	<20year	22(11%)
Age	20-30 year	113(56.5%)
0	30-40 year	46(23%)
	>40 year	19(9.5%)
	Male	126(63%)
Gender	Female	73(36.5%)
	Missing	1(0.5%)
	Medical Engineering	70(35%)
Field of study	MI	43(21.5%)
	HIT	82(97.5%)
	missing	5(2.5%)
	BA	77(38.5%)
Degree	MA	73(36.5%)
-	Professional doctorate	43(21.5%)
	HIT, HIM, Medical Record	55(27.5%)
Prior field	MI	12(6%)
	C-E-M*	33(16.5%)
	0 year	87(43.5%)
Work experience	1-5 year	62(31%)
	5-10	24(12%)
	>10	27(13.5%)
	Yes	123(61.5%)
Activity	No	70(35%)
	Missing	7(3.5%)
	Yes	81(40.5%)
Exposure	No	109(54.5%)
	missing	10(5%)

Table 2	Individual	characteristics	of the	narticinants
I able #	mairiauai	character istics	or the	paracipanto

Most students were between 20 and 30 years old. 63% of them were male and 43.5% had no work experience. Current and previous field of study of most of the students were HIT, HIM, and Medical Records. Most of the participants in this study were undergraduates. 61.5% were economically active. 54.5% were exposed to Big Data. The mean scores of participants in benefits, applications, and challenges section were 3.71, 3.68, and 3.71, respectively (SAS-challenge, SAS-advantage and SAS-application). Examination of SAS-challenge, SAS-advantage, and SAS-application by variables of age, gender, field of study, Prior field, work experience, with / without activity, exposure / non-exposure to Big Data can be seen on Table 3.

		groups	
Questions	Age	n	Mean $\pm$ SD(n)
		Ν	
	<20 year	22	.6986±.11620
	20-30 year	113	.7522±.12519
Advantages	30-40 year	46	.7574±.12829.
0	>40 year	19	.7252±.12159
	Total	200	.7449±.12508
	<20 year	22	.6989±.12051
	20-30 year	113	.7413±.13019
Applications	30-40 year	46	.7528±.11257.
	>40 year	19	.7147±.12070
	Total	200	.7368±.12462
	<20 year	22	.6869±.15257
	20-30 year	113	.7392±.16566
Challenges	30-40 year	46	.7744±.15188
<u> </u>	>40 year	19	.6982±.18948
	Total	200	.7377±.16466

## Table 3. Comparison of mean of SAS-challenge, SAS-advantage and SAS-application across different age groups

One-way ANOVA test was used to compare the mean of SAS-challenge, SAS-advantage and SAS-application in different age groups with no significant difference in different age groups in these factors. P-Value was 0.228, 0.317, and 0.139 respectively.

## Table 4. Comparison of the mean of SAS-challenge, SAS-advantage and SAS-application in different gender groups

		81	
	Gender	n	Mean $\pm$ SD(n)
	Gender	Ν	Mean ± SD(n)
Advantages	Male	126	.7454±.11719
Ū.	Female	73	.7471±.13709
Applications	Male	126	.7329±.13281
	Female	73	.7446±.11009
Challenges	Male	126	.7383±.17504
0	Female	73	.7370±.14741

According to Table 4, the Independent t-test was used to compare the mean of SAS-challenge, SAS-advantage and SAS-application in different gender groups with no significant difference in different age groups in these factors.

## Table 5. Comparison of the average of SAS-challenge, SAS-advantage, and SAS-application across different fields of study

indus of sources			
	field	n	$Mean \pm SD(n)$
	neid	Ν	Mean ± 3D(II)
	Medical engineering	70	.7302±.13611
advantages	MI	43	.7760±.12040
Ũ	HIT	82	.7488±.11194
	Total	195	.7481±.12348
	Medical engineering	70	.7236±.11333
applications	MI	43	.7778±.13359
	HIT	82	.7337±.12527
	Total	195	.7398±.12416
	Medical engineering	70	.7140±.15265
challenges	MI	43	.8114±.16246
Ŭ	HIT	82	.7293±.16021
	Total	195	.7419±.16167

In Table 5, the results of the One-way ANOVA test were showed which compare the mean of SAS-challenge, SAS-advantage and SAS-application in different fields, but the mean of SAS-application and SAS-advantage were not significant .The mean of SAS-challenge was significant in different disciplines. The mean of SAS-challenge in medical informatics was higher than other majors (Fig. 1).

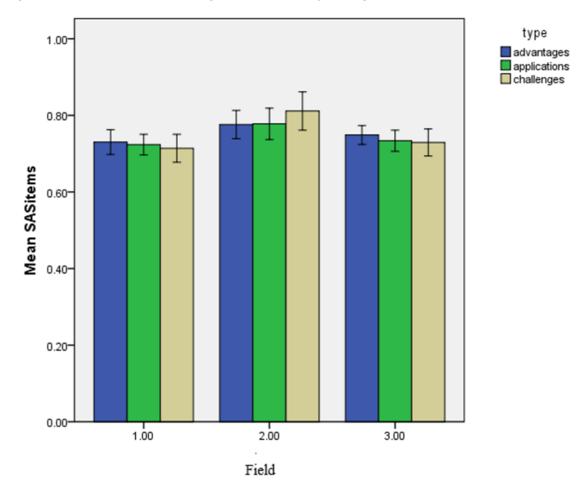


Fig. 1. Mean of Benefits, applications and challenges in terms of the different fields of study

levels of study				
	Degree	n	$ Mean \pm SD(n)$	
	Degree	Ν	Mean ± 3D(n)	
	BSC	77	.7270±.12249	
Advantages	MSC	73	.7521±.13956	
	PHD	43	.7718±08582	
	Total	193	.7465±.12313	
	BSC	77	.7249±.13235	
Applications	MSC	73	.7415±.12568	
11	PHD	43	.7602±.09684	
	Total	193	.7390±.12285	
	BSC	77	.6987±.16116	
Challenges	MSC	73	.7461±.17661	
Ŭ	PHD	43	.7953±.12388	
	Total	193	.7382±.16345	

 Table 6. Comparison of the mean of SAS-challenge, SAS-advantage, and SAS-application between different

 lavels of study

One-way ANOVA test was used to compare the mean of SAS-challenge, SAS-advantage, and SAS-application at different levels of study that the mean of SAS-application, SAS-advantage, and SAS-challenge were not, according to Table 6. Significant P-Value were 0.142, 0.313, and 0.006 respectively.

	neius	or study	
	Prior field	n	Mean $\pm$ SD(n)
		Ν	
	HIT	55	.7678±.11516
	MI	12	.7675±.06710
Advantages	Engineering, electronics, math	33	.7652±.16349
	Total	100	.7669±.12796
	HIT	55	.7503±.11978
	MI	12	.7893±.08649
Applications	Engineering, electronics, math	33	.7548±.12074
	Total	100	.7564±.11628
	HIT	55	.7693±.16136
	MI	12	.7889±.13283
Challenges	Engineering, electronics, math	33	.7946±.16150
	Total	100	.7800±.15728

 Table 7. Comparison of the mean of SAS-challenge, SAS-advantage, and SAS-application between previous fields of study

The one-way ANOVA test was used to compare the mean of SAS-challenge, SAS-advantage and SAS-application between the previous fields of study, but according to Table 7, the mean of SAS-application, SAS-advantage and SAS-challenge were not significant.

work experiences				
	Work experience	n	Mean $\pm$ SD(n)	
	0 year	87	.7459±.11910	
	1-5 year	62	.7620±.13001	
Advantages	5-10	24	.7154±14901	
0	>10	27	.7290±.10848	
	Total	200	.7449±.12508	
	0 year	87	.7426±.12193	
	1-5 year	62	.7441±.13534	
Applications	5-10	24	.7185±.10836	
* *	>10	27	.7176±.12415	
	Total	200	.7368±.12462	
	0 year	87	.7367±.16638	
	1-5 year	62	.7559±.15951	
Challenges	5-10	24	.7167±.16671	
Ŭ	>10	27	.7177±.17332	
	Total	200	.7377±.16466	

Table 8. Comparison of the mean of SAS-challenge, SAS-advantage, and SAS-application between different
work experiences

On Table 8, One-way ANOVA test was used to compare the mean of SAS-challenge, SAS-advantage and SAS-application between different work experiences that the mean of SAS-application, SAS-advantage and SAS-challenge were not significant. P-Value were 0.404, 0.673, and 0.673 respectively.

	Activity	Ν	Mean $\pm$ SD(n)
Advantages	Yes	123	.7521±.12231.
_	No	70	.7403±.13160.
Applications	Yes	123	.7454±.12092.
	No	70	.7185±.13357.
Challenges	Yes	123	.7478±17636
	No	70	.7251±.14521.

 Table 9. Comparison of the mean of SAS-challenge, SAS-advantage, and SAS-application in groups with /

 without economic activity

On Table 9, the Independent t-test was used to compare the mean of SAS-challenge, SAS-advantage and SAS-application in the groups with / without economic activity in these factors. P-Value were 0.532, 0.155, and 0.361 respectively.

 Table 10. Comparison of the mean of SAS-challenge, SAS-advantage, and SAS-application in groups with /

 without exposure to Big Data

	Exposure	n N	Mean $\pm$ SD(n)
	Exposure		Mean ± 5D(II)
Advantages	Yes	81	.7619±.11752
_	No	109	.7359±.13009
Applications	Yes	81	.7561±.11112
	No	109	.7239±.13370
Challenges	Yes	81	.7627±.15108
	No	109	.7252±.16977

According to Table 10, the Independent t-test was used to compare the mean of SAS-advantage, SASchallenge and SAS-application in the groups with / without exposure to Big Data that there is no significant difference between the groups with / without exposure to Big Data in these factors. P-Value were 0.157, 0.08, and 0.116 respectively. In order to examine the SAS-advantage, SAS-challenge and SAS-application subdomains, the previous analysis of each sub-domain is repeated in terms of variables such as age, gender, field of study, degree, and so on.

	A = -	n	$M_{com} + SD(n)$
	Age	N	
	<20 year	22	.7491±.14458
	20-30 year	113	.7692±.17013
Information	30-40 year	46	.7843±.15966
	>40 year	19	.7789±.12534
	Total	200	.7714±.16057
	<20 year	22	.7364±.15324
	20-30 year	113	.7611±.18425
Modeling	30-40 year	46	.7754±.18979
	>40 year	19	.7719±.16226
	Total	200	.7627±.17954
	<20 year	22	.6545±.14790
	20-30 year	113	.7384±.15538
Data	30-40 year	46	.7252±.14910
	>40 year	19	.7137±.17802
	Total	200	.7238±.15637
	<20 year	22	.6742±.17516
	20-30 year	113	.7451±.14399
Process_Managment	30-40 year	46	.7529±.14633
	>40 year	19	.6667±.20458
	Total	200	.7317±.15655

Table 11. Comparison of the mean of SAS-advantage, SAS-challenge and SAS-application domains by age

	<20 year	22	.6786±.13032
	20-30 year	113	.7358±.14164
Health_Sevice_Delivery	30-40 year	46	.7453±.11181
	>40 year	19	.7009±.11773
	Total	200	.7284±.13269
	<20 year	22	.6909±19557
	20-30 year	113	.7996±19451
Research	30-40 year	46	.8087±.20718
	>40 year	19	.7421±.20430
	Total	200	.7843±.20054
	<20 year	22	.7182±.13040
	20-30 year	113	.7357±.15005
Health_Information	30-40 year	46	.7405±.13875
	>40 year	19	.7092±.16398
	Total	200	.7324±.14611
	<20 year	22	.6982±.11377
	20-30 year	113	.7371±.15626
Essential_Medicines	30-40 year	46	.7591±.13224
	>40 year	19	.7039±.13988
	Total	200	.7347±.14565
	<20 year	22	.7091±.24477
	20-30 year	113	.7469±.21384
Health_Financing	30-40 year	46	.7478±.20842
	>40 year	19	.7789±.22992
	Total	200	.7460±.21661
	<20 year	22	.7106±.13584
	20-30 year	113	.7428±.16770
Leadership_Governance	30-40 year	46	.7551±.16810
	>40 year	19	.7667±.14741
	Total	200	.7443±.16227

One-way ANOVA test was used to compare the mean of SAS-advantage, SAS-challenge and SAS-application domains by age groups that process management, according to Table 11, became significant. P-Value were 0.855, 0.861, 0.145, 0.046, 0.172, 0.072, 0.831, 0.315, 0.784, and 0.680, respectively.

	Gender	n N	Mean $\pm$ SD(n)
	Gender		Mean ± SD(n)
Information	Male	126	.7679±.15380
	Female	73	.7759±.17317
Modeling	Male	126	.7566±.17391
C	Female	73	.7735±.19076
Data	Male	126	.7168±.15714
	Female	73	.7370±.15605
Process management	Male	126	.7447±.14168
U U	Female	73	.7183±.16207
Health service delivery	Male	126	.7252±.13799
	Female	73	.7357±.12353
Research	Male	126	.7794±.21441
	Female	73	.7932±.17664
Health information	Male	126	.7268±.15116
	Female	73	.7426±.13836
Essential medicines	Male	126	.7328±.15952
	Female	73	.7394±.15952
Health financing	Male	126	.7317±.22650
0	Female	73	.7699±.19908
Leadership governance	Male	126	.7405±.16874
* ~	Female	73	.7516±.15245

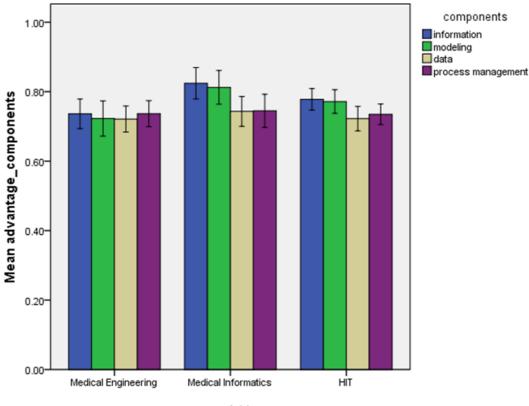
Table 12. Mean comparison of SAS-advantage, SAS-challenge and SAS-application domains by gender

According to Table 12, the Independent t-test was used to compare the mean of SAS-advantage, SAS-challenge and SAS-application in gender groups with no significant difference in gender in these factors. P-Value were 0.738, 0.525, 0.383, 0.230, 0.592, 0.642, 0.463, 0.761, and 0.234, respectively.

		of study	
	Field of study	n	
	5	Ν	
	Medical engineering	70	.7354±.18137
Information	MI	43	.8242±.14688
	HIT	82	.7780±.14113
	Total	195	.7729±.16058
	Medical engineering	70	.7238±.20968
Modeling	MI	43	.8124±.15753
	HIT	82	.7715±.15539
	Total	195	.7634±.17949
	Medical engineering	70	.7211±.15692
Data	MI	43	.7433±.13972
	HIT	82	.7224±.16066
	Total	195	.7266±.15441
	Medical engineering	70	.7367±.15739
Process Management	MI	43	.7450±.15514
0	HIT	82	.7350±.13465
	Total	195	.7378±.14699
	Medical engineering	70	.7227±.12854
Health Service Delivery	MI	43	.7502±.13100
,	HIT	82	.7261±.13873
	Total	195	.7302±.13320
	Medical engineering	70	.7321±.17796
Research	MI	43	.8802±.16873
	HIT	82	.7854±.21907
	Total	195	.7872±.20119
	Medical engineering	70	.7212±.13062
Health Information	MI	43	.7887±.14366
	HIT	82	.7216±.14812
	Total	195	.7363±.14310
	Medical engineering	70	.7181±.13169
Essential Medicines	MI	43	.7758±.14452
	HIT	82	.7379±.15266
	Total	195	.7391±.14449
	Medical engineering	70	.7457±19537
Health Financing	MI	43	.7349±.22560
manuli i munching	HIT	82	.7512±.22566
	Total	195	.7456±.21423
	Medical engineering	70	.7367±.14625
Leadership Governance	MI	43	.7713±.19032
e	HIT	82	.7394±.16188
	Total	195	.7455±.16304

Table 13. Mean comparison of SAS-advantage, SAS-challenge and SAS-application domains by different fields of study

On Table 13. One-way ANOVA test was used to compare the mean of SAS-advantage, SAS-challenge, and SAS-application domains by field of study, that the mean of SAS-advantage, SAS-challenge, and SAS-challenge in information, modelling, research, and health informatics were significant. P-Value were 0.015, 0.033, 0.726, 0.935, 0.532, 0.001, 0.024, 0.119, 0.922 and 0.500 respectively (Fig. 2 and Fig. 3).



Field

Fig 2. Average of the components of Benefits by field of study

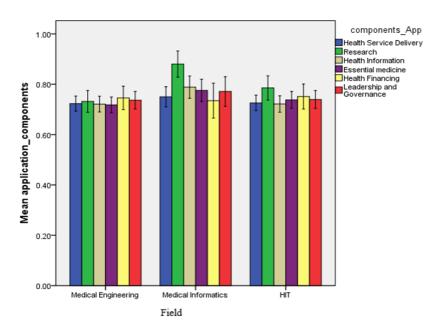


Fig 3. Average of the components of Application by field of study

	Degree	N	
	BSC	77	.7356±.17902
Information	MSC	73	.7863±.15581
	PHD	43	.8047±.12443
	Total	193	.7702±.16131
	BSC	77	.7359±.18589
Modeling	MSC	73	.7553±.19470
0	PHD	43	.8155±13161
	Total	193	.7610±18059
	BSC	77	.7122±.16066
Data	MSC	73	.7332±.16547
	PHD	43	.7386±.11787
	Total	193	.7260±.15380
	BSC	77	.7277±.13050
Process Management	MSC	73	.7379±.18116
0	PHD	43	.7504±.10193
	Total	193	.7366±.14627
	BSC	77	.7178±.13130
Health Service Delivery	MSC	73	.7289±.14290
,	PHD	43	.7505±.10839
	Total	193	.7293±.13117
	BSC	77	.7487±.20822
Research	MSC	73	.7849±.21192
	PHD	43	.8558±.14809
	Total	193	.7863±.20112
	BSC	77	.7237±.15430
Health Information	MSC	73	.7408±14471
	PHD	43	.7462±.12148
	Total	193	.7352±.14352
	BSC	77	.7186±.15838
Essential Medicines	MSC	73	.7450±.14472
	PHD	43	.7606±.11002
	Total	193	.7380±.14393
	BSC	77	.7377±.22771
Health Financing	MSC	73	.7479±.22367
0	PHD	43	.7581±.17759
	Total	193	.7461±.21505
	BSC	77	.7455±.15284
Leadership Governance	MSC	73	.7406±.17080
-	PHD	43	.7605±.15106
	Total	193	.7470±.15886

# Table 14. Mean comparison of SAS-advantage, SAS-challenge and SAS-application domains by different levels of study

On Table 14, One-way ANOVA test was used to compare the mean of SAS-advantage, SAS-challenge and SAS-application domains by different levels of study that the mean of SAS-advantage, SAS-challenge and SAS-application in information and research were significant that was more significant at PhD level. P-Value were 0.043, 0.064, 0.589, 0.717, 0.427, 0.019, 0.654, 0.269, 0.880, and 0.807, respectively.

	Degree	n	Mean $\pm$ SD(n)
	Degree	Ν	Mean ± SD(n)
	HIT	55	.8196±.12290
Information	MI	12	.8033±.13694
	C-E-M	33	.7600±.19183
	Total	100	.7980±.15153
	HIT	55	.7782±.16079
Modeling	MI	12	.8444±.12818
0	C-E-M	33	.7980±.19825
	Total	100	.7927±.17053
	HIT	55	.7484±.13612
Data	MI	12	.7133±.12630
	C-E-M	33	.7442±.18599
	Total	100	.7428±.15226
	HIT	55	.7358±.15567
Process Management	MI	12	.7444±.10856
0	C-E-M	33	.7707±.18004
	Total	100	.7483±.15894
	HIT	55	.7435±.12499
Health Service Delivery	MI	12	.7578±.11098
,	C-E-M	33	.7390±.15264
	Total	100	.7438±.13211
	HIT	55	.8091±.21860
Research	MI	12	.9333±.07177
	C-E-M	33	.8167±.16802
	Total	100	.8265±.19325
	HIT	55	.7389±.14249
Health Information	MI	12	.7885±.10091
	C-E-M	33	.7583±.14340
	Total	100	.7512±.13829
	HIT	55	.7556±.13039
Essential Medicines	MI	12	.7978±.10511
	C-E-M	33	.7503±.13676
	Total	100	.7589±.12946
	HIT	55	.7964±.19048
Health Financing	MI	12	.7333±.19695
incatin i marcing	C-E-M	33	.7455±.23061
	Total	100	.7720±.20503
	HIT	55	.7394±.16840
Leadership Governance	MI	12	.7722±.11962
r	C-E-M	33	.7616±.18373
	Total	100	.7507±.16774

# Table 15. Mean comparison of SAS-advantage, SAS-challenge and SAS-application domains by different previous fields of study

According to Table 15, One-way ANOVA test was used to compare the mean of SAS-advantage, SAS-challenge and SAS-application domains by different previous fields of study that the mean of SAS-advantage, SAS-challenge and SAS-application was not significant. P-Value were 0.202, 0.469, 0.772, 0.610, 0.916, 0.122, 0.501, 0.537, 0.420 and 0.749 respectively.

		experience	
	Work experience	n / N	Mean $\pm$ SD(n)
	0year	87	.7674±.15341
	1-5 year	62	.7923±.17650
Information	5-10 year	24	.7283±.15999
	>10 year	27	.7748±.14471
	Total	200	.7714±.16057
	Oyear	87	.7678±.16824
	1-5 year	62	.7785±.18098
Modeling	5-10 year	24	.6944±.23003
-	>10 year	27	.7704±.15616
	Total	200	.7627±.17954
	0year	87	.7223±.15424
	1-5 year	62	.7497±.14771
Data	5-10 year	24	.6917±.18062
	>10 year	27	.6978±.15858
	Total	200	.7238±.15637
	Oyear	87	.7368±.15668
	1-5 year	62	.7387±.14897
Process Management	5-10 year	24	.7347±.14955
0	>10 year	27	.6963±.18171
	Total	200	.7317±.15655
	Oyear	87	.7348±.14072
	1-5 year	62	.7454±.13775
Health service delivery	5-10 year	24	.7049±.11289
ricatifi service delivery	>10 year	27	.6893±.10273
	Total	200	.7284±.13269
	Oyear	87	.7810±.19681
	1-5 year	62	.8081±.19210
Research	5-10 year	24	.7437±.22904
Research	>10 year	27	.7759±.20911
	Total	200	.7843±.20054
	Oyear	87	.7385±.14954
	1-5 year	62	.7403±.14303
Health information	5-10 year	24	.7156±.13054
	>10 year	27	.7093±.15893
	Total	200	.7324±.14611
		87	.7352±.14611
	Oyear 1-5 year	62	.7452±.14810
Essential medicines		24	
Essential medicines	5-10 year	24	.7206±.13025
	>10 year		.7220±.14984
	Total	200	.7347±.14565
	Oyear	87	.7770±.19630
TT 1.1 C '	1-5 year	62	.6968±.22830
Health financing	5-10 year	24	.7583±.26361
	>10 year	27	.7481±.19684
	Total	200	.7460±.21661
	Oyear	87	.7625±.13826
- 1 1·	1-5 year	62	.7129±.18782
Leadership governance	5-10 year	24	.7361±.16794
	>10 year	27	.7654±.16316
	Total	200	.7443±.16227

# Table 16. Mean comparison of SAS-advantage, SAS-challenge and SAS-application domains by experience

On Table 16, an One-way ANOVA test was used to compare the mean of SAS-advantage, SAS-challenge and SAS-application domains by experience that the mean of SAS-application, SAS-advantage, and SAS-challenge were not significant. P-Value were 0.419, 0.255, 0.327, 0.661, 0.231, 0.592, 0.725, 0.863, 0.167, and 0.270 respectively.

economic activity			
	Activity	Ν	Mean $\pm$ SD(n)
Information	yes	123	.7776±.16034
miomation	no	70	.7657±.16447
Modeling	yes	123	.7691±.17850
Modeling	no	70	.7533±.18970
Data	yes	123	.7389±.15444
Data	no	70	.7091±.15941
Process Management	yes	123	.7333±.15611
i iocess management	no	70	.7386±.14319
Health service delivery	yes	123	.7296±.13578
rieatin service derivery	no	70	.7217±.13088
Research	yes	123	.8089±.19110
Research	no	70	.7364±.21752
Health information	yes	123	.7419±.14312
	no	70	.7112±.15432
Essential medicines	yes	123	.7457±.13364
Essential medicines	no	70	.7168±.16645
Health financing	yes	123	.7463±.22914
	no	70	.7314±.19821
Leadership governance	yes	123	.7561±.15770
Leadership governance	no	70	.7195±.17450

# Table 17. Mean comparison of SAS-advantage, SAS-challenge and SAS-application domains by economic activity

The Independent t-test was used to compare the mean of SAS-advantage, SAS-challenge and SAS-application by economic activity that, according to Table 17, there was a significant difference in different groups in research. P-Value were 0.625, 0.565, 0.205, 0.693, 0.818, 0.017, 0.167, 0.761, 0.188, 0.649 and 0.133, respectively.

 Table 18. Mean comparison of SAS-advantage, SAS-challenge and SAS-application domains by exposure / non-exposure to Big Data

	Exposure	Ν	Mean $\pm$ SD(n)
Information	yes	81	.7970±.14571
Information	no	109	.7545±.16905
Modeling	yes	81	.7835±.17177
Modeling	no	109	.7468±.18851
Data	yes	81	.7358±.15135
Data	no	109	.7196±.16271
Process Management	yes	81	.7436±.15755
Frocess Management	no	109	.7284±.14613
Health service delivery	yes	81	.7413±.13150
Treatur service delivery	no	109	.7182±.13423
Research	yes	81	.8142±.19577
Research	no	109	.7624±.20940
Health information	yes	81	.7475±.13899
Treatur information	no	109	.7218±.15380
Essential medicines	yes	81	.7567±.12935
Essential medicines	no	109	.7231±.15477
Health financing	yes	81	.7704±.20028
i icaitii iiiiaitciiig	no	109	.7248±.23060
Leadership governance	yes	81	.7778±.15330
readership governance	no	109	.7217±.16957

The Independent t-test was used to compare the mean SAS-advantage, SAS-challenge and SAS-application by exposure / non-exposure to the Big Data that, according to Table 18, there was a significant difference between groups of information. P-Value were 0.071, 0.169, 0.486, 0.085, 0.494, 0.236, 0.114, 0.156, 0.020,

0.761, 0.188, 0.649, and 0.133 respectively. The mean of SAS-information was higher among those exposed to the Big Data than those not exposed to the Big Data.

### 4. Conclusion

Today, with the advent of technologies and the production of huge amounts of data, Big Data analytics have received much attention especially in healthcare. Understanding this field and recognizing its benefits, applications and challenges provide useful background for conducting efficient research. Therefore, the purpose of this study was to evaluate the students' familiarity from different universities of Mashhad with the benefits, applications and challenges of Big Data analysis. Most students were between 20 and 30 years old. Most of them were male and had no work experience. Current and previous field of study of most of the students were HIT, HIM, and Medical Records. Most of the participants in this study were undergraduates. Most of them were economically active and were exposed to Big Data. The mean scores of participants in benefits, applications, and challenges section were 3.71, 3.68, and 3.71, respectively. Considering that the participants in this study are students from the top universities in the country and have done some Big Data research, it is assumed that Mashhad students have a better level of knowledge in the field of Big Data analysis. Yet there should be more opportunities for students, even organizations' staff to get to know the field more. Training in this field is essential for many disciplines, also conferences could be effective in introducing this field. Students can also provide more familiarity and usage of functional analytics by conducting new researches in this field. In the section of challenges, benefits and application analytics, process management was significantly in different age groups, research, modelling and information and health informatics across different fields of studies were significant. Information and research were significantly different between different levels of studies. Research in groups with / without economic activity was significant and information in exposure / non exposure to Big Data groups was significant. Despite the importance and benefits of Big Data analytics, students' lack of familiarity with the necessity and importance of these analytics in industries and research is significant. The field of study and level of study does not appear to have an effect on the degree of knowledge of individuals regarding Big Data analysis. In future studies, it is suggested that students, practitioners, and other disciplines in different cities and countries evaluate the specific benefits and applications of Big Data analytics and compare the results. Because it will be possible to study in different places and different perspectives. In other businesses, checking their familiarity with Big Data analytics can be helpful in applying management and advertising policies. Big data analytics can play a constructive role in all industries, and today it is widespread in most industries and businesses. Because of the growing trend of data generation, Big Data analytics will become a necessity for all industries and areas in coming years.

### Availability of data and materials

These data are available.

### Funding

The author(s) received *no* financial support for the research, authorship, and/or publication of this article.

#### Acknowledgments

The present study is the result of research project approved by the vice chancellery for research of Mashhad University of Medical Sciences (grant number 961731).

### Ethics approval and consent to participate

Ethics approval is under grant number 961731 Publication is permitted by Mashhad University of Medical Science

#### **Consent for publication**

Publication is permitted by Mashhad University of Medical Science

### References

- Acharjya, D. P., & Ahmed, K. (2016). A survey on big data analytics: challenges, open research issues and tools. International Journal of Advanced Computer Science and Applications, 7(2), 511-518.
- Alharthi, A., Krotov, V., & Bowman, M. (2017). Addressing barriers to big data. Business Horizons, 60(3), 285-292.
- Archenaa, J., & Anita, E. M. (2015). A survey of big data analytics in healthcare and government. Procedia Computer Science, 50, 408-413.
- Belle, A., Thiagarajan, R., Soroushmehr, S. M., Navidi, F., Beard, D. A., & Najarian, K. (2015). Big data analytics in healthcare. BioMed research international, 2015.
- Gharachorloo, N., Nahr, J. G., & Nozari, H. (2021). SWOT analysis in the General Organization of Labor, Cooperation and Social Welfare of East Azerbaijan Province with a scientific and technological approach. International Journal of Innovation in Engineering, 1(4), 47-61.
- Manogaran, G., Lopez, D., Thota, C., Abbas, K. M., Pyne, S., & Sundarasekar, R. (2017). Big data analytics in healthcare Internet of Things. In Innovative healthcare systems for the 21st century (pp. 263-284). Springer, Cham.
- Nahr, J. G., Nozari, H., & Sadeghi, M. E. (2021). Green supply chain based on artificial intelligence of things (AIoT). International Journal of Innovation in Management, Economics and Social Sciences, 1(2), 56-63.
- Nazari, E., Norouzi, S., Aldaghi, T., Rasoulian, M., Shahriari, M. H., Kheirdoust, A., & Tabesh, H. (2021). A Survey of Students' Attitudes to Big Data Analysis in Iranian Universities. International Journal of Innovation in Engineering, 1(4), 62-71.
- Nozari, H., Fallah, M., Kazemipoor, H., & Najafi, S. E. (2021). Big data analysis of IoT-based supply chain management considering FMCG industries. Бизнес-информатика, 15(1 (eng)).
- Srinivasan, U., & Arunasalam, B. (2013). Leveraging big data analytics to reduce healthcare costs. IT professional, 15(6), 21-28.



This work is licensed under a Creative Commons Attribution 4.0 International License.