id	Ph	ases/categories	identifier	CSFs/challenges (literature)	Description Access to contextual information such as process models,	sources	Input (Requirements)	Output	Possible Al-tools
1			1.1		husiness rules, noticy documents, lengt and regulatory				
				Availability of contextual information	requirements that can aid process mining (Mans et al., Mamudu and Bandara).	Mans et al., Mamudu and Bandara			
					and Bandara).  teentrying questions or project goals), selecting business processes to be mined and composing the project seam to execute process mining initiatives (Mamudu ad Bandara).		1		
			1.2		execute process mining initiatives (Mamudu ad Bandara).				
					It is unclear what process properties are important (Grisold et				manual task. But "random" Proces Analyses, like with the Proactive Insights Engine (Volt et al.), may result in (new) research questions.
				Planning (Process selection)	al.)	Mamudu and Bendara, Grisold et al.			
	Defin	e research question			The composition of teams and expert groups involved in process		Process models, business rules, policy documents, legal and regulatory requirements, possible projectleam-members	selected business process, composed project team, project goals, defined questions	
					mining projects. Two main configurations namely: Established units: An internal team dedicated to ex-ecuting process mining initiatives. E.g., a Centre of Excellence (CoE). Ad-hoc units: A		regulatory requirements, possible projectioann-members	goss, osninia questionis	Engine (vert et al.), may result in (new) research quastons.
					initiatives. E.g., a Centre of Excellence (CoE). Ad-hoc units: A				
			1.3	Team confineration	group of experts assembled from dif-ferent departments within the organisation to execute process mining projects as and when required (Mamudu and Bandara).				
				*					
					Lack of interdisciplinary and cross-functional teams: PM suffers from a lack of interdisciplinary and cross-functional teams covering sponsors, IT, and data specialists as well as business users and project managers (Martin et al.)				
					from a lack of interdisciplinary and cross-functional teams covering sponsors. IT. and data specialists as well as business				
					users and project managers (Martin et al.)	Mamudu and Bendara, Martin et al.			
					The extent to which historical event data is available for process mining analysis (Mamudu and Bandara).		process description, tracked systems & distributes, distributes disconnectation, which helicitud event data are available	sees access to the distribution, data princy regulations defined, rate data experted, conceptual data model and a model.	Database or asker to find the belonging databases, takes and entires, Aquates Clamble F. York Sorrania Applications.
					ming analysis (warnood and bandara).				
					Constraining data access barriers: Limited data access across departmental and organizational boundaries restricts PM (Martin				
					et al.)				
					The availability of event data needed for PM is limited (Martin et				
					aL)				
			2.1	Evert data availability	Restricting data privacy regulations: Compliance with data privacy				
1				- Jacoby	and security regulations limits the detail of what can be discovered and analyzed through PM (Martin et al.)				
1					Difficult handling of unstructured data: PM provides limited support				
2		Data collection			for exploiting unstructured data that is not available in activity- based semantics or even format (Martin et al.).				
1									
1					There is an asymetry in terms of the permission to access and use of relevant data ( <b>Grisold et al.</b> ).				
1					Delays can occur due to data access, which is often tied to				
1					organizational barriers (Grisold et al.).				
					The required data analytics expertise for the extraction and	Mamudu and Bandara, Martin et al., Grisold et al.			
1					integration of event data for process mining (Mamudu and Bandara).				
1			2.2	Data extraction expertise	Teams who are recessful for data interesting after hour				
1					difficulties to obtain the data since they are not involved in the decision-making (Orisold et al.).  Determining the data extraction scope, extracting event data, and		-		
					decision-making (Grisold et al.).  Determining the data extraction scope, extracting event data, and	Mamudu and Bandara, Grisold et al.			
			2.3	Extraction	transferring process knowledge be-tween business experts and process analysts (Mamudu and Bandara).	Mamudu and Bandara			
					Provisions for the extraction and preparation of event data from single or multiple sources for process min-ing based on lessons				
					karnt (Mamudu and Bandara).		exported raw dutia		
			3.1	Data preprocessing	Complex data preparation: Substantial effort is required for data				
					extraction and pre-processing (Martin et al.)			filtered event-log based on the research questions	Automated Eusschop crusion (fl.gest de Marillas et al.). PM44/NRHE (Mourant et al.)
3	Dat	ta pre-processing			There are data fractions when process run on different systems (Grisold et al.)				
						Mamudu and Bandara, Martin et al.			
			3.2	Evert-log quality considerations	The data quality considerations and minimum requirements to be met by event logs for process mining (Mans et al., Mamudu and				
					Bandara).				
					Source or event data are often in accurate, noisy, and/or				
_					incomplete (Martin et al.)	Mans et al., Mamudu and Bandara, Martin et al.			i
					Applying process mining techniques to answer ques-tions and gair insights (Mans et al.).				
1									Presides Insight Engine: From Process Discovery to Process Intelligence (Wall et al.)  Molitade an Advanced Prediction Process Monitoring Todds (Ritzel et al.)
			4.1	i	Insufficient technical skills: The lack of sufficient training in technical skills required to implement PM is detrimental to setting up and conducting PM (Martin et al.)			required insights with different relevas, based on the receasersh quantities who is opinitized process model	
				Afficiance of Academic	Process managers miss information about how certain variables	Mans et al. Martin et al., Grisoid et al.			
1				Mining and Analysis	can inform decision-making ( <b>Grisoti et al.</b> ).	manus es as, swiftin et al., Orisido et al.			
1					Integration of process mining capabilities with other data analytics capabilities (Mamudu and Bandara).	1			
1		General .	42	α.	Challenging (real-time) system integration: Insufficient real-time				
1					Challenging (real-time) system integration: Insufficient real-time system connectivity or integration into existing IT infrastructure negatively impacts deriving insights through PM (Martin et al.)				
1				Tool capabilities: Integration capabilities		Mamudu and Bandara, Martin et al.			
					The tool's ability to analyse data for insights into sin-gle, multiple and e2e processes (Mamudu and Bandara).				
1			4.3		Fragmented solutions: There is a lack of comprehensive PM				
1	as/tar				solutions supporting a wide range of conceivable use cases	L			
4	38.00			Tool capabilities: Analytical Scalability	(Martin et al.). Non-standard visualization techniques used in PM may lead to	Mamudu and Bandara, Martin et al.	Event-Log, Process model, research questions		
1	Minga		4.4	Incomprehensible outcomes		Martin et al.			
	-		4.5	Lack of advanced features	models (Martin et al.). PM lacks advanced features such as automation, simulation, and rists approximization (Martin et al.).	Martin et al.			
1				- Accessed management	data anonymization (Martin et al.).  Automated process model discovery and process vis-ualisation from evert data (Marnudu and Bandara).		1		
			4.6						
		Discovery	,	Tool capabilities: Process discovery	Difficult analysis of process exceptions: PM lacks support for deriving insights from process exceptions (Martin et al.).	Mamudu and Bendara, Martin et al.			
1			4.7		Using process mining tools to create views, aggregate events, enrich or filter loss to generate the required insights from event				
				Data processing	Unificult analysis of process exceptions: PM lacks support for deriving insights from process exceptions (Martin et al.). Using process mining tools to create views, aggregate events, enrich or filter logs to generate the required insights from event logs (Mamudu and Bandara).	Mamudu and Bendara	-		
1		Conformance	4.8	Tool capabilities: Conformance checking/Compliance	Datastica of desistings from process power union count data				
					(Marrudu and Bandara). Using event data for comparison of process behav-lours and process performance (Marrudu and Bandara).	Memudu and Bandara	-		
		Durfamore	4.9	L	Insufficient prescriptive capabilities: PM tools are limited regarding their prescriptive capabilities (Martin et al.).				
		Performance		Tool capabilities: Process Benchmarking	their prescriptive capabilities (Martin et al.). No challenges found No challenges found	Mamudu and Bandara, Martin et al.	4		
			4.10			I .	]		
		Performance  Social network analysis  Comparitive analysis	4.10 4.11	I.			1		
			4.10 4.11	I I					
			s 4.10 4.11	I I	Relating analysis results to improvement ideas to achieve project goals (Mamudu and Bandara).				
			4.10 4.11	I I	Relating analysis results to improvement ideas to achieve project goals (Mamudu and Bandara).		Insichts with different views, prenared in an understandolds		
5	Results		4.10 4.11	J.	Relating analysis results to improvement ideas to achieve project goals (Mamudu and Bandara).  One of the challenges in process mining projects is often that the process analysts are not domain experts for the process they are analyzing (Bookaya et al., Surfadie et al.), which means that they		Insights with different views, prepared in an understandable way for the stakeholders (presentation, etc.), direct suggestions for	Enthusiastic stakeholders who will continue to support PM in the future.	Artificial Inteligence Enabled Project Management (Taboeda et al.), PMASSAME (Rourani et al.)
5			4.10 4.11	/ / Evaluation			Insights with different views, prepared in an understandable way for the stakeholders (presentation, etc.), direct suggestions for improvement which should be made	Enthusiastic stakeholders who will continue to support PM in the feature.	Artificial Intelligence Enabled Project Management (Taboada et al.), PMAROMNE (Mourant et al.)
s		Social network analysis Comparitive analysis Stakeholder evaluation		Evaluation Mission involvement from process excepts	Rabifor analysis results to improvement ideas to actives project point (Marmoda and Banderas).  One of the childenges in process mining projects is often that the process analysis are not downs experts for the process that employing (Bodzeys et al., Swindt et al.), which means that they enalysis results are determining the creams of unaspected analysis results.	Mamudu and Bandara. Bothava et al., Suriad et al.	Insights with different views, proposed in an understandable way for the stakeholders (presentation, etc.), direct suggestions for improvement which should be made	Enthusiastic stakeholders who will continue to support PM in the future.	Adficial intelligence Enabled Project Management (Taboada et al.), PMAYNME (Mourant et al.)
s			s 4.10 4.11 5.1		Relating analysis results to improvement ideas to achieve project goals (Mamoutu and Bandara).  One of the childregs in process mining projects is often that the process analysts are not domain experts for the process they are analysing (Booksaya et al., Surladi et al.), which means that they may have difficulties determining the course of unswelling.		Insights with different views, proposed in an understandable way for the stakeholders (presentation, etc.), direct suggestions for improvement which alread to made	Enthusiasds stakeholders who will continue to support PM in the future.	Artificial Intelligence Enabled Project Management (Taboada et al.), PMANNALE (Kourani et al.)
5		Social network analysis Comparitive analysis Stakeholder evaluation		Evaluation Missian insistement from oncoses ascerts Process improvement and support	Rabifor analysis results to improvement ideas to actives project point (Marmoda and Banderas).  One of the childenges in process mining projects is often that the process analysis are not downs experts for the process that employing (Bodzeys et al., Swindt et al.), which means that they enalysis results are determining the creams of unaspected analysis results.	Memody and Bandara, Bobbara et al., Soried et al.  Memody and Bandara	Insights with different views, properted in an understandable way for the stakeholders (presentation, etc.), direct appearions for improvement which should be made	Erthusiadic stakeholders who will confinue to support PM in the follow.	Adficial Intelligence Southed Project Management (Tabosda et al.), PM40NME (Gourant et al.)

_	Ĺ	_

Supports



Supports



6	Stakeholder Support and Innoherment	6.1	Management support	Top-Level Management/Senior Executives support (Mans et al., Marmotu and Bandara).  Indiating, funding, and conducting PM initiatives requires a strong management commitment (Martin et al.)  Process management need galdance to convince decision-makers (Martin et al.)	Mans et al., Mannudu and Bandara, Martin et al., Griscid et al.	Time, money, perseasion and training (senhalops). Informed and education reportance of PM		Practice (Source et al.)
		6.2	External stakeholider support	retrainment at al.  Engagement with endernal collaborators or industry partners (such as suppliers) who influence an organization's business process and how they are executed (Marmadu and Bandara).  Transparency leads to distrust and perceived surveillance (Orisoid et al.).	Mamudu and Bandara, Grisold et al.			
		6.3	Subject matter experts (SMEs)	SMEs of a particular business domain who contribute to process mining efforts (Mamudu and Bandara).	Mamudu and Bandara			
		6.4	User groups	The contribution of ultimate users (such as first-line personnel) to process mining outcomes (Mamudu and Bandara).	Mamudu and Bandara			
		6.5	Process mining expertise	The required know-how needed to execute process mining initiatives and interpret outcomes (Mamudu and Bandara).	Mans et al. Marrudu and Bandara		Informed and educated stakeholders with an understanding of the importance of PM to the business and the PM project being completed.	
		6.6	Process unin/est expertise	The required expertise for designing, streamfring, and re- organizery bourness processes (Mess et al., Mammobs and Desidera).  Instifficient samples of white. The lack of fundamental analysis of substitution analysis of white. The lack of fundamental analysis of substitution analysis of white set as a large of contraction of the substitution of substitution of the substitution of the substitution of and to barriess expertise sinkless the white for contraction PM as well as to substitution of the substitution of the substitution of the substitution of the substitution of the substitution of the substitutio	Mares et al. Marrodu pro Bacdara. Martin et al.			
		6.7	Training	The encustors are sensistance or statements on the appropriate execution of process mining initiatives for the intended results (Mamudu and Bandlara). Insufficient sechnical skills: The lack of sufficient training in technical skills required to implement PM is detrimental to setting up and conducting PM (Martin et al.)				
7	Organizational and strategic adigment	7.1	Change Management	The series of activities that ensure that the needed change emanating from process mining results is in-plemented in the organisation (Manasa and Bandera). Unclear organization (Manasa of Bandera). Unclear organizational archoring it is unclear how PM expertise should be enrolled within the organization (Martin et al.). It is important to cope with the horsessed transparency created through process mining (Orleaded et al.).	Memole and Bandara, Martin et al., Grinold et al.  Mores et al., Memole and Bandara.	not a concrete phase in the PM-process. For this reason there	These are general challenges and concern belonging PM. These red a concern place in the PR-Sproces. For this reason these	Arthial Halligarus Eroldel Prijest Management (Falkeda et al.)
		7.2	Project Management	The management of activities and resources, such as time and cost throughout all phases of the process mining project to obtain the defined project outcomes (Mans et al., Mamudu and Bandara).				
		7.3	Unclear success factors	It is unknown which organizational setups and properties ensure an efficient and effective use of PM (Martin et al.).		are no direct Inputs or Outputs.	are no direct Inputs or Outputs.	
		7.4		The business value of PM is difficult to determine with regard to the alignment of strategic and operational goals as well as the quartification of costs and benefits (Martin et al.).  Process managers do not know how to calculate the business	Martin et al.			
		7.5	Elusive business value  Missing implementation guidance	value of PMI activities ( <b>Crisold et al.</b> ) There is a lack of comprehensive guidance on the implementation of PMI for efferent organizations, domains, contexts, and strategic gools ( <b>Martin et al.</b> ).	Martin et al., Grisold et al.  Martin et al.			