id		Phases/ categories	identifier	CSF's/challenges (literature)	Description	sources	Input (Requirements)	Output	Possible Al-tools
П			1.1		Access to contextual information such as process models, business rules, policy documents, legal and regulatory requirements that can				
		ne research question	1.2	Availability of contextual information  Planning (Process selection)	aid process mining [19], [20].  Mentifying questions or project goal(s), selecting business processes to be minied and composing the project team to execute process mining initiatives [19].  It is unclear what process properties are important [23].	Mans et al., Marrodu et al.  Marrodu et al., Cirisold et al.			
1	Define		1.3	Team configuration	The composition of barreland apper proportional of a process making project. The most configuration example, Self-Self-Self-Self-Self-Self-Self-Self-	Mamosto et al., Martin et al.	mocan's metodogi (e.g. Process marchid, etc.) [3], business nices, policy documents, legal and registrior reconsenents, possible projections members projections members	alletind business process [3], composed project flasm [3], project guard, defined research questions [3], [4]	manual late. But "section" Proces Analyses, the with the Proced we brights Engine (65), may result in (sees) research credition.
2	Di	Data contection	2.1	Event data availability	The extent to which historical event data is available for process mining analysis [19].  Combattering data assess baselines. Limited data access across desputements and organizational broaders writing [19].  Readleding data privacy applications. Compliations with calls privacy and except preparations and except preparations and except preparations with except data privacy applications. Compliations with calls privacy and except privacy and privacy and except privacy and except privacy and except privacy. The provides limited applied for expelled privacy and except privac	Mamoulu et al., Marin et al., Gilabit et al.	process secretifies (\$1, toules appears, Subblasses, databases decumentation, which tributed event data are available	socials to the distribution, data privacy regulations startled, staw data expected (E), (4) conceptual data model (4)	Dubbase crasher to find the belonging distabase. Safes and entities, Aponto OpenRP [81].  Veils Boraping Applications (82)
			2.2	Data extraction expertise	The required data analytics apertise for the extraction and integration of event data for process immig [91].  Teams such are responsible for data integration of event data for data integration of the data integration of the data integration of the data integration of the decision making [23]  Ostermining the data statistiction togos, exhibiting event data, and transfering process increfedge be-leven business experts and process analyses?	Mamodu et al., Grisold et al.			
$\vdash$			23		process analysts [19].	Mamudu et al.			
3	Detr	pre-processing	3.1	Duta preprocessing	Provisions for the extraction and preparation of event data from single or multiple sources for process min-risp based on leasens team? [19].  Complex data preparation: Subdanfial effort is required for data extraction and are-processing [22].  There are data fractions when process run on different systems [23].	Maemodu et al., Madin et al., Gifadd et al.	esponded raw dala [3]	filend eventing based on the research questions [1], [4]	Automated Exercising creation (63), PMASSAME (84)
			3.2	Event-log quality considerations	The data quality considerations and minimum requirements to be met by event logs for process mining [19], [20].  Source or event data are often in accurate, noisy, and/or incomplete [22].	Mans et al., Mamudu et al., Marlin et al.			
		General	4.1	Mining and Analysis	Applying process mining techniques to answer questions and gain neighb [20]. Insufficient technical adills: The lack of sufficient training in technical adills sequired to implement PM is determined to settling grant conducting PM [22]. Process managers miss information about how certain variables can inform decision-neigh [23].	Mans et al., Martin et al., Ofisiold et al.	EvenLag (I), Process model (I), research questions (4)	required insights with different Young (S), based on the reasonth questions also a optimized process model	Presedve Insights Engine, From Process Discovery to Process Intelligence (EI). Notices: on Advanced Predictive Process Manifolding Tooliki (EI)
			4.2	Tool capabilities: Integration capabilities	Integration of process mining capabilities with other data analytics capabilities [19].  Challenging (seal-time) system integration: fraufficient real-time system connectivity or integration into existing (1T infrastructure negatively impacts deriving insights through PM [22].	Mamudu et al., Marlin et al.			
4	g & analysis		4.3	Tool capabilities: Analytical Scalability	The foot's ability to analyse data for insights into sin-gle, multiple and EEE processes [19]  Fargmented soldiness. There is a lack of comprehensive PM solutions supporting a wide range of conceivable use cases [22].  Non-dandest visualization techniques used in PM may lead to  overcomplicated and hardly understandable business process	Mamudu et al., Marlin et al.			
	Minin			Incomprehensible outcomes	models [22].  PM lacks advanced features such as automation, simulation, and	Martin et al.			
		Discovery	4.6	Lack of advanced features  Tool capabilities: Process discovery	data anonyimization [22]. Automated process model discovery and process vis-ualisation from event data [19].  Officult analysis of papers papersions. PM tasks purpost for	Martin et al.  Mamodu et al., Martin et al.			
			4.7	Data processing	deriving inrights from process exceptions [22].  Using process mining tools to create views, aggregate events, enrich or filter logs to generate the required insights from event logs [19].	Mamudu et al., Martin et al.			
		Conformance	4.8	Tool capabilities: Conformance checking/Compliance	Detection of deviations from process norms using event data [19].  Using event data for comparison of process behav-lours and process performance [19].	Mamodu et al.			
		Social network analysis	4.10	Tool capabilities: Process Benchmarking /	Insufficient prescriptive capabilities: PM tools are limited regarding their prescriptive capabilities [22]. No challenges found	Mamudu et al., Martin et al.	1		
		Comparitive analysis Stakeholder evaluation	5.1	7	No challenges found  Relating analysis results to improvement ideas to achieve project goals [19].  One of the challenges in process mining projects is often that the anness analysis are not downs in aments for the orners they are	<i>y</i>	Insights with different views, prepared in an understandable way for	Enthusiadic stakeholders who will continue to support PM in the	Artificial Intelligence Enabled Project Management (67).
5	Results	Implementation	5.2	Evaluation Missing involvement from process experts	One of the challenges in process mining projects is often that the process analysis are not domain experts for the process they are analysing 10(1). Siny, which means that they may have difficulties determining the causes of unexpected analysis results. Using gained insights to modify the actual process execution [19].	Mamudu et al., Bozkaya et al., Suriadi et al.	tesights with different views, prepared in an understandable way for the stakeholders (presentation, etc.) [29], direct suggestions for improvement which should be made	Enthussatic statementers who will constitue to support PM in the future, Documentation	PM4KNME [84]
ш				Process improvement and support	1	Mamudu et al.	I	l .	

				Supports	$\hat{\mathbf{U}}$	Supp	poorts	)
	Stakeholder Support and Snothwester.	6.1	Management support	Top-Level Management/Senior Executives support [19], [20]. talidating, funding, and conducting PM initiatives requires a strong management commitment [22]. Process managers need guidance to convince decision-makers [23].	Mans et al., Marrudo et al., Marlin et al., Grisold et al.		telomised and educated stateholdern with an understanding of the importance of PM to the business and the PM project being designated.	manual ta A. PASCHME (64)
		6.2	External stalleholder support	Engagement with external collaborators or industry partners (auch as suppliers) who influence an organisation's business process and how they are executed [19].  Transparency may lead to distrust and perceived surveillance [23].	Mamodu and Bandara, Grisold et al.			
		6.3	Subject matter experts (SMEs)	SMEs of a particular business domain who contribute to process mining efforts [19].	Mamudu and Bandara			
		6.4	User groups	The contribution of ultimate users (such as first-line personnel) to process mining outcomes [19].	Mamudu and Bandara			
6		6.5	Process mining expertise	The required know-how needed to execute process mining initiatives and interpret outcomes [19].	Mans et al., Mamudu and Bandara	Time, money, persuasion and training (workshops).		
		6.6	r notes mining superior	The required expertise for designing, streamlining, and re- engineering business processes [19], [20].	Many et al., Manusch and Bandara, Martin et al.			
				Insufficient analytical skills: The lack of fundamental analytical skills, including business process modelling and optimization, impedes deriving value from PM [22].				
			Process analyst expertise	Insufficient domain expertise: The lack of comprehensive domain and business expertise inhibits the ability to customize PM as well as to adequately interpret the results [22].				
		6.7	Training	The education and sensitisation of stakeholders on the appropriate execution of process mining initiatives for the intended results [19].				
				Insufficient technical skills: The lack of sufficient training in technical skills required to implement PM is detrimental to setting up and conducting PM [22].				
F				I	I	I		
7	Organizational and divolegic alignment	7.1	Change Management	The series of activities that ensure that the needed change emanating from process mining results is im-plemented in the organisation [19].				
				Unclear organizational anchoring: It is unclear how PM expertise should be anchored within the organization [22].				
				It is important to cope with the increased transparency created through process mining [23].	Mamudu et al., Martin et al., Grisold et al.			
		7.2		The management of activities and resources, such as time and cost throughout all phases of the process mining project to obtain the defined nonzero renormed [50]	These are general challenges and concerns belonging PM. Thats	These are general challenges and concerns belonging PM. Thats		
		7.3	Project Management	It is unknown which organizational setups and properties ensure an efficient and effective use of PM [22].	Mans et al., Mamudu et al.	not a concrete phase in the PM-process. For this reason there are no direct inputs or Outputs.	not a concrete phase in the PM-process. For this reason there are no direct Inputs or Outputs.	Artificial Intelligence Enabled Project Management (67)
		7.4	Unclear success factors  The bosiness value of PM is difficult to determine with egged to the alignment of strategic and operational goals as well as the quantification of costs and benefits [27].	Martin et al.				
			Elusive business value	Process managers do not know how to calculate the business value of PM activities [23].	Martin et al., Grisold et al.		l	
1				There is a lack of comprehensive guidance on the implementation				