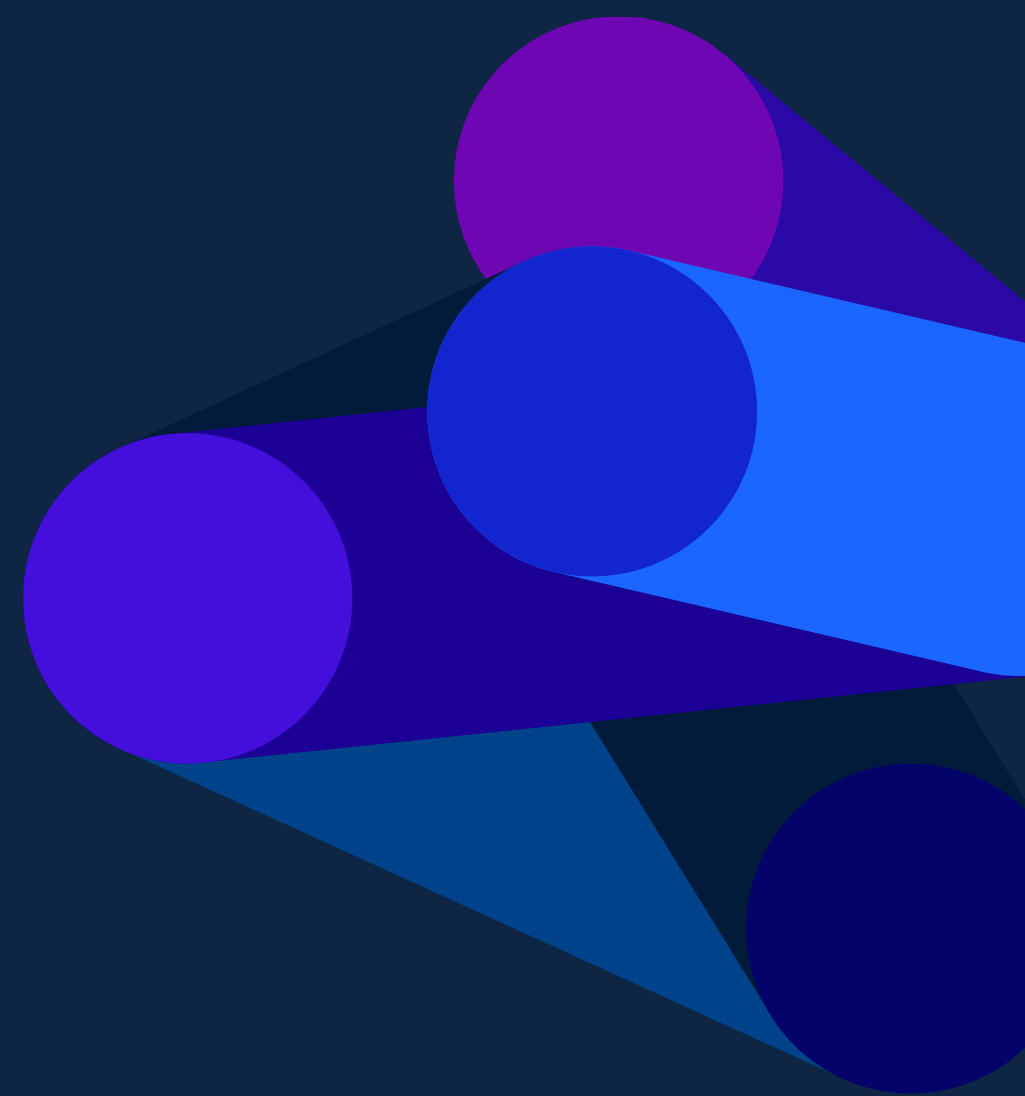


Process Mining Research in Korea

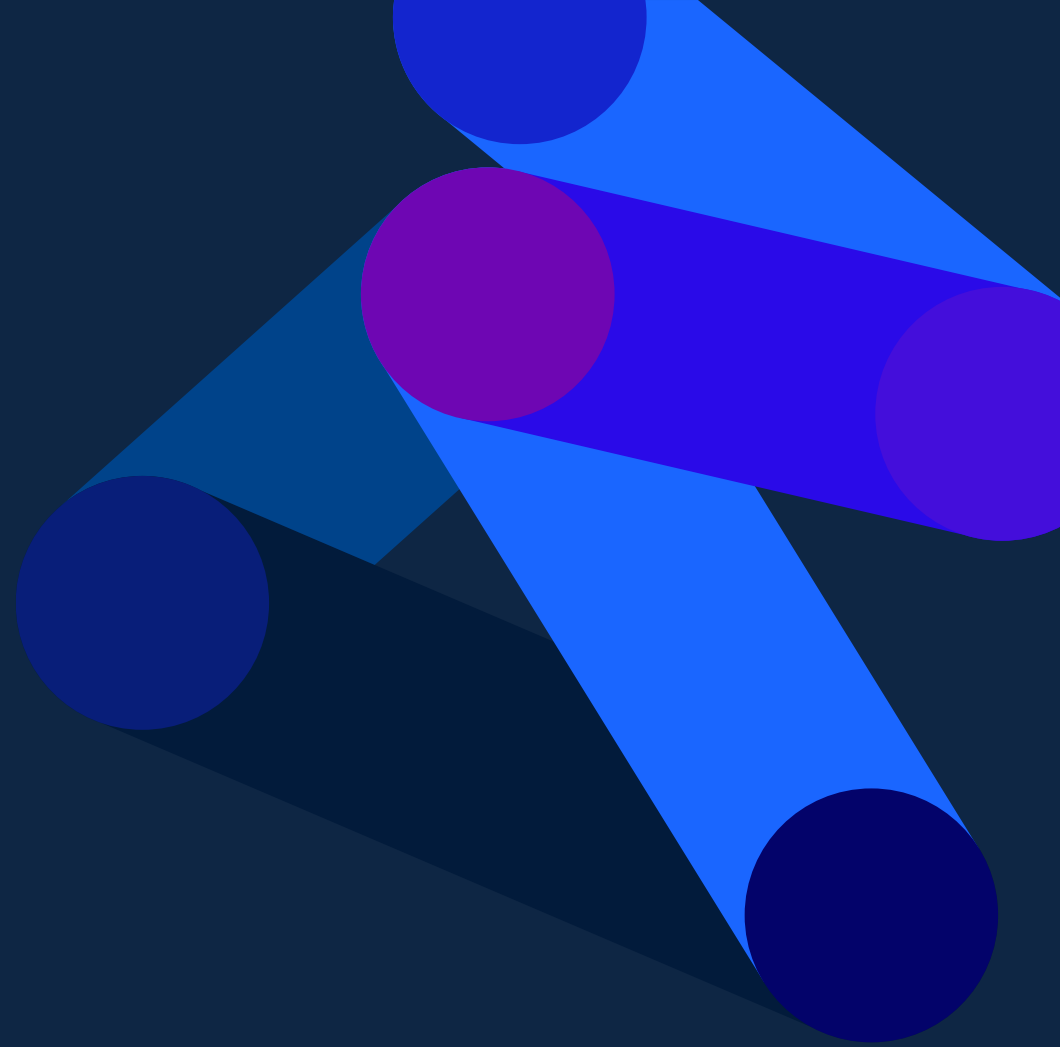
송민석

mssong@postech.ac.kr

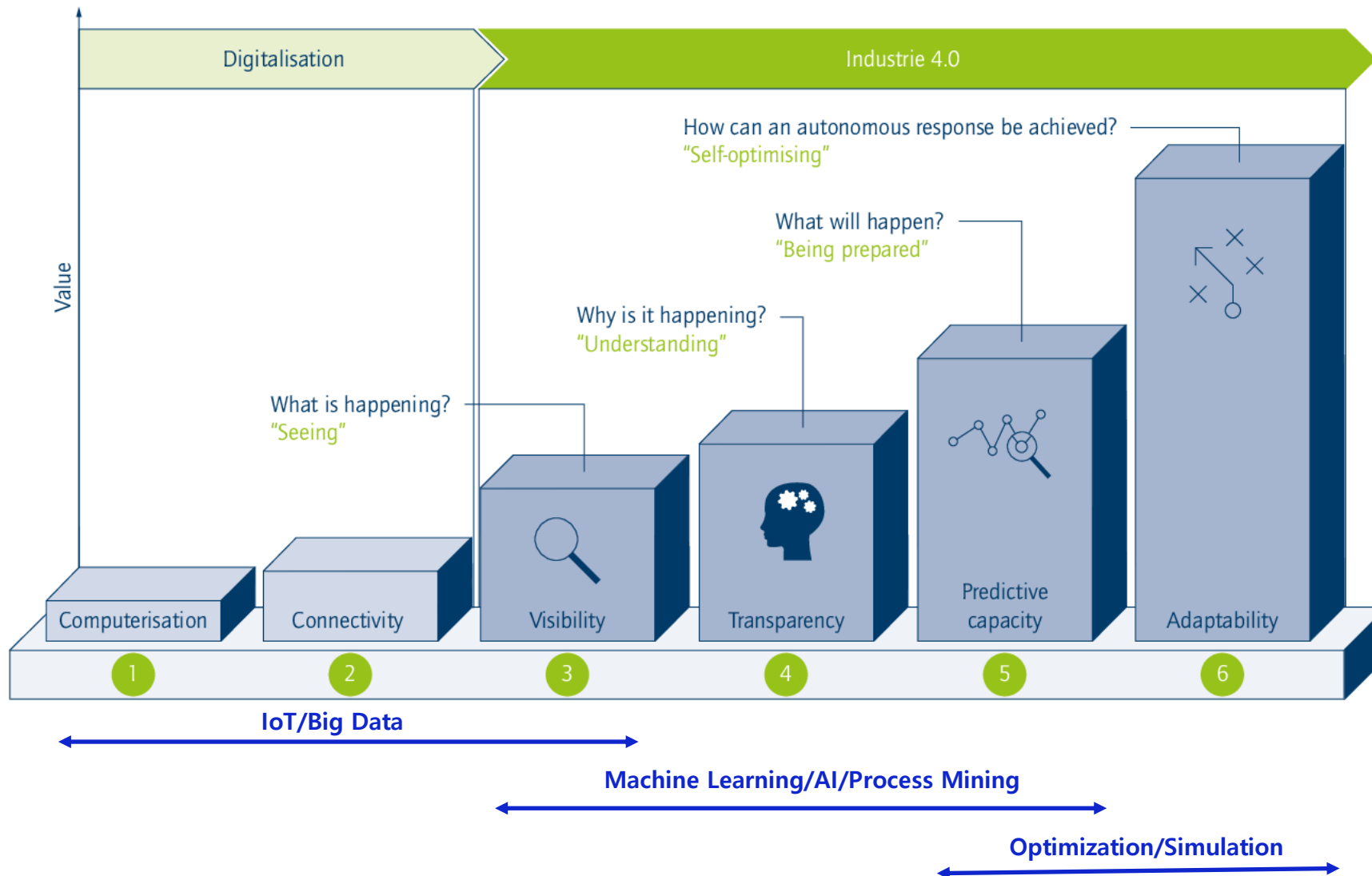


Contents

- Introduction
- Process Mining
- Cases
- Conclusion



Stages in the Industrie 4.0 development path



출처: Acatech, Industrie Maturity Index (2017)

RWTH: Wil Van der Aalst

Homepage Wil van der Aalst

HOME PUBLICATIONS RESEARCH COURSES PROCESS SCIENCE
DATA SCIENCE CURRICULUM VITAE PERSONAL NEWS PADS@RWTH

Wil van der Aalst is full professor at RWTH Aachen University, leading the Process and Data Science group, and chief scientist at Celonis.

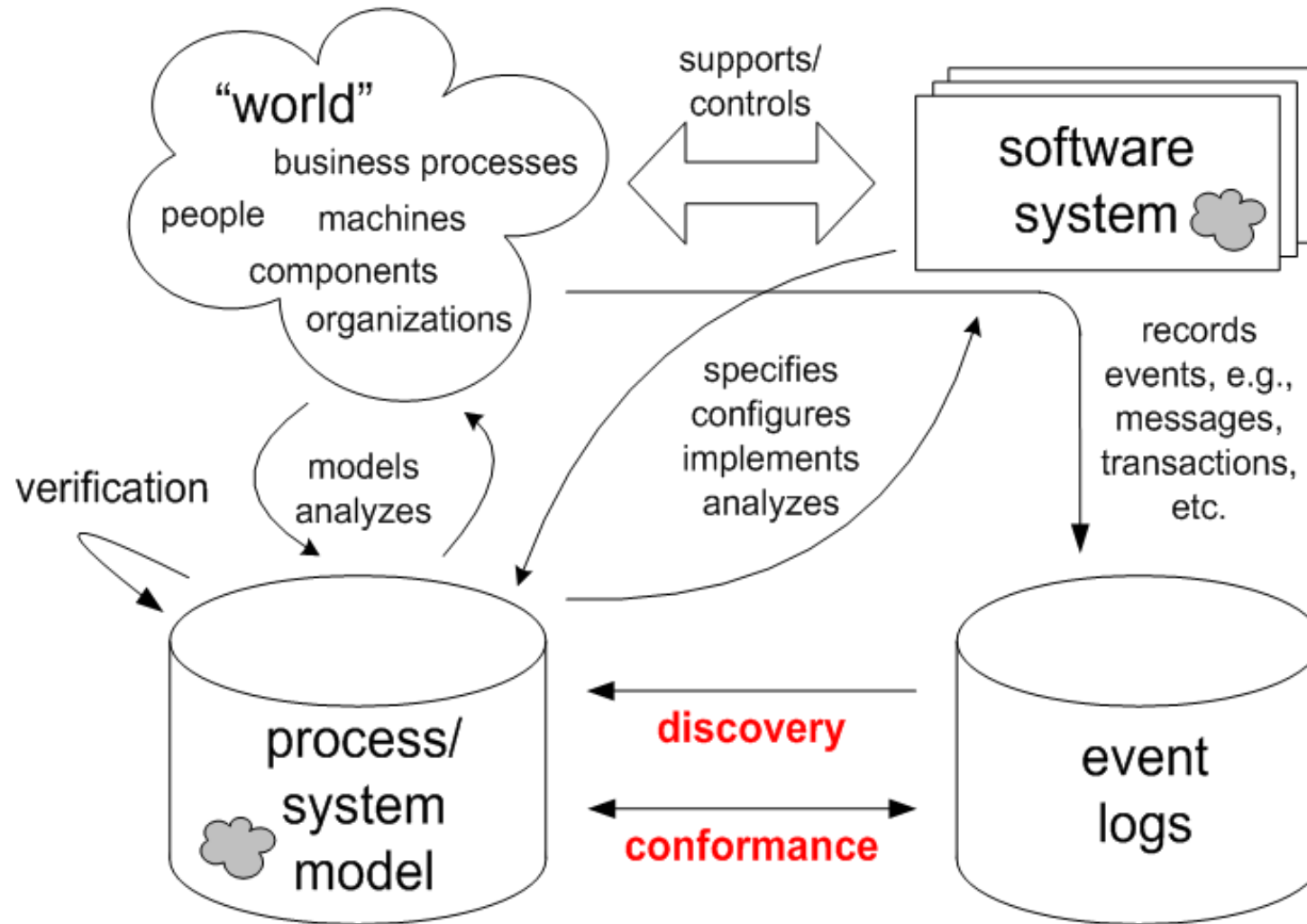
Welcome to my virtual home!

- Check out my publications!
- More about WvdA
- More on process mining
- Check out the latest news!
- 2021 Publications are available
- Also visit the PADS site for news

Follow @wvdaalst
Tweet to @wvdaalst

H-index:163, Citations:122,000
By Google Scholar

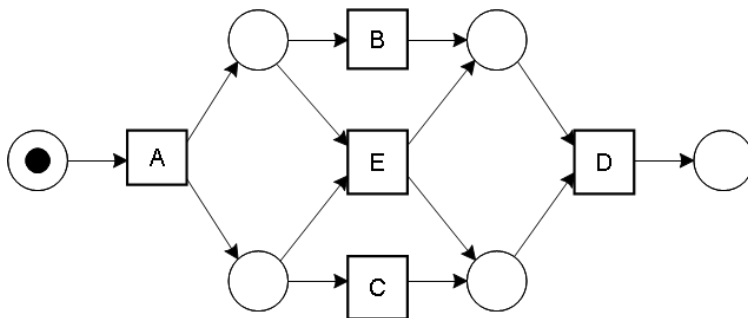
Process mining: Linking events to models



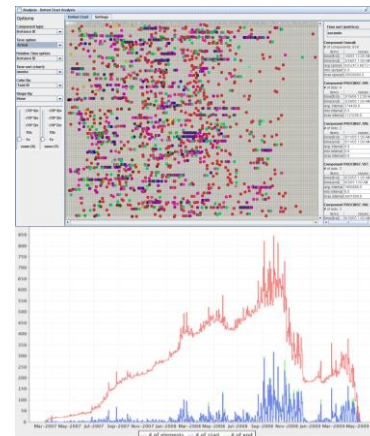
Discovery

Cases	Logs
1	(A, John, 10:22:21), (B, Sue, 10:25:20), (C, John, 10:40:21), (D, Pete, 10:45:25)
2	(A, John, 10:24:27), (C, Carol, 10:32:12), (B, Clare, 10:44:23), (D, Pete, 10:50:45)
3	(A, Carol, 11:52:35), (E, Mike, 13:23:56), (D, Sue, 14:42:13)
4	(A, Pete, 13:21:25), (C, Carol, 14:23:21), (B, Clare, 15:34:13) (D, Pete, 16:21:13)
5	(A, Sue, 14:22:34), (E, Pete, 15:32:42), (D, Clare, 16:13:15)

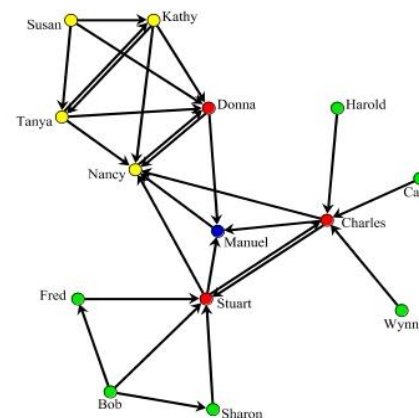
Process Model



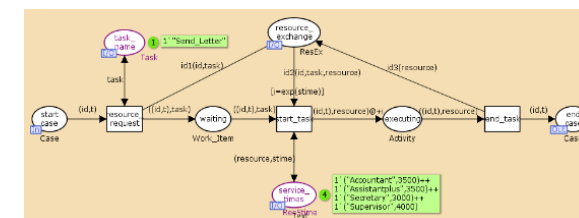
Performance Metrics

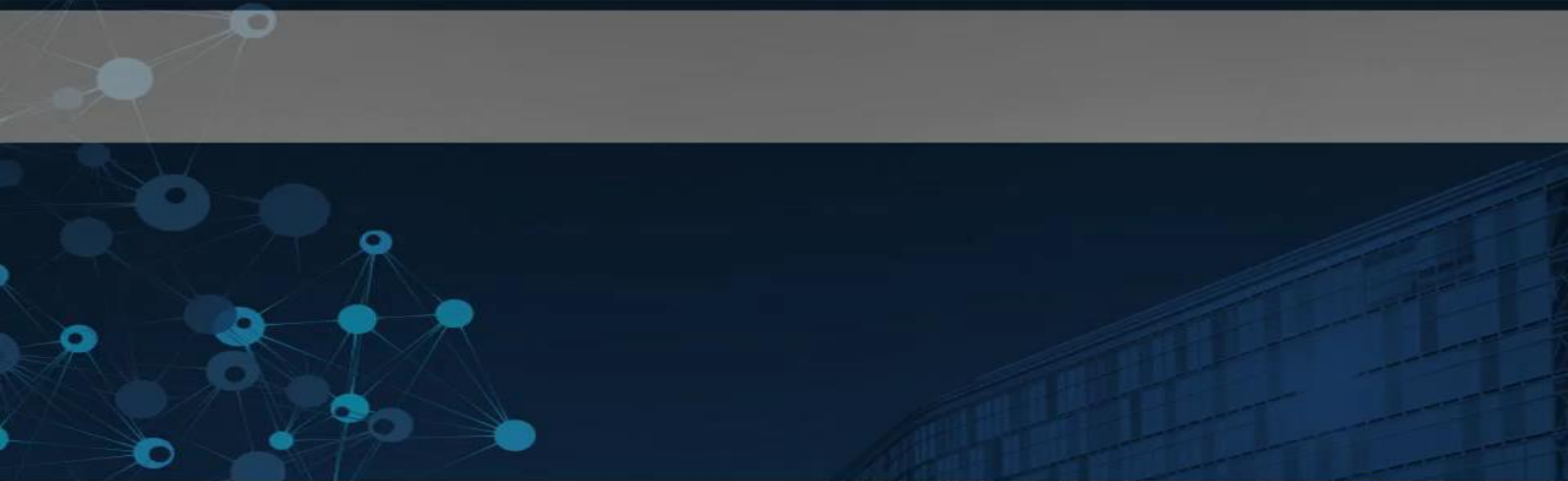


Social Network

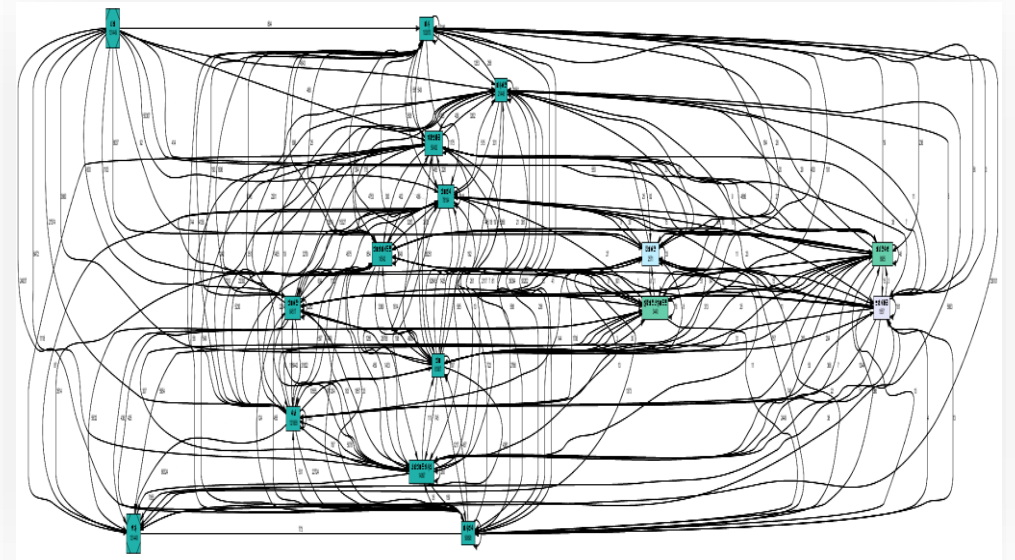


Simulation Models

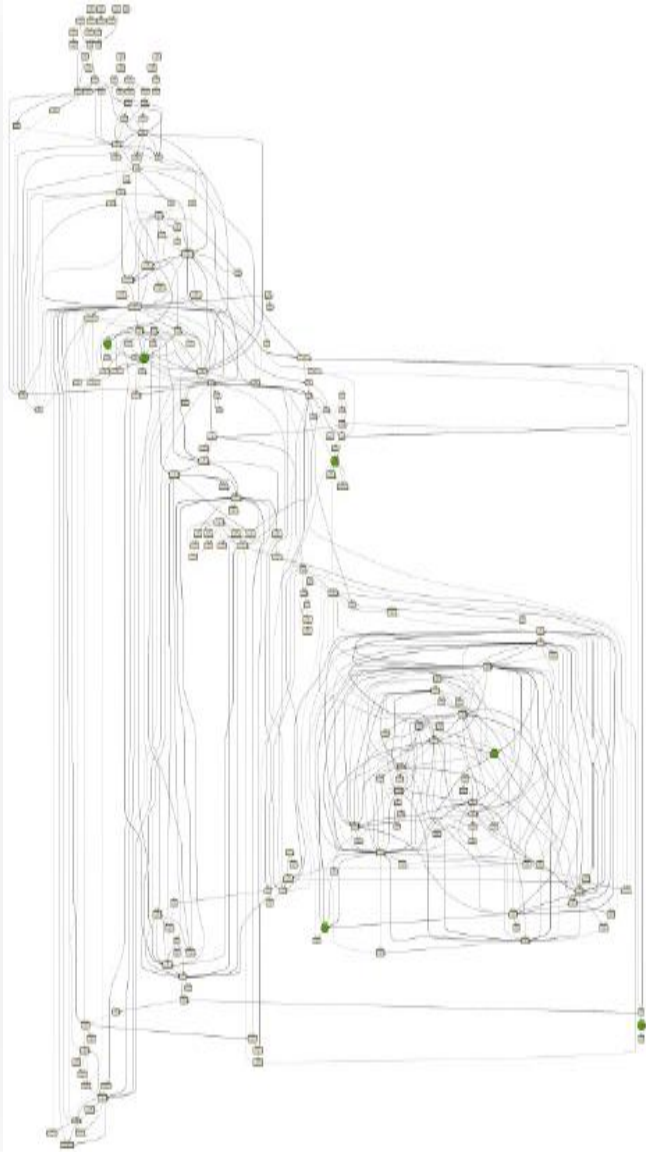




Outpatient processes in SNUBH



A manufacturing process in Samsung Electronics



A shipbuilding process in HHI



Trace clustering - examples

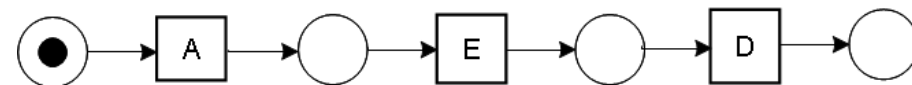
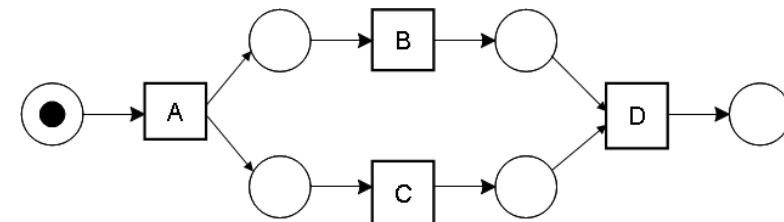
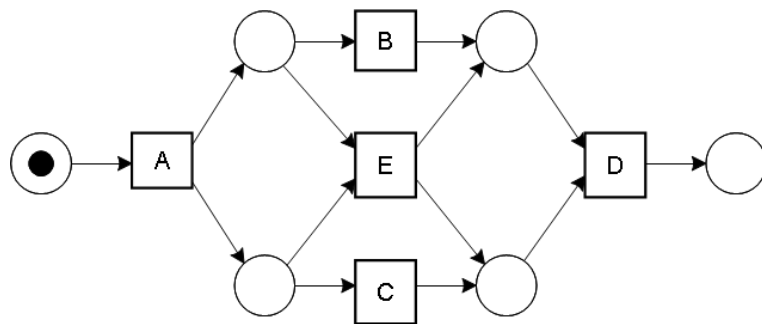
Cases	Logs
1	(A,John),(B,Sue),(C,John),(D,Pete)
2	(A,John),(C,Mike),(B,John),(D,Sue)
3	(A,Carol),(E,Mike),(D,Sue)
4	(A,Pete),(C,Carol),(B,Clare) (D,Pete)
5	(A,Sue),(E,Pete),(D,Clare)

Cases	Logs
1	(A,John),(B,Sue),(C,John),(D,Pete)
2	(A,John),(C,Mike),(B,John),(D,Sue)
4	(A,Pete),(C,Carol),(B,Clare) (D,Pete)

Cases	Logs
3	(A,Carol),(E,Mike),(D,Sue)
5	(A,Sue),(E,Pete),(D,Clare)

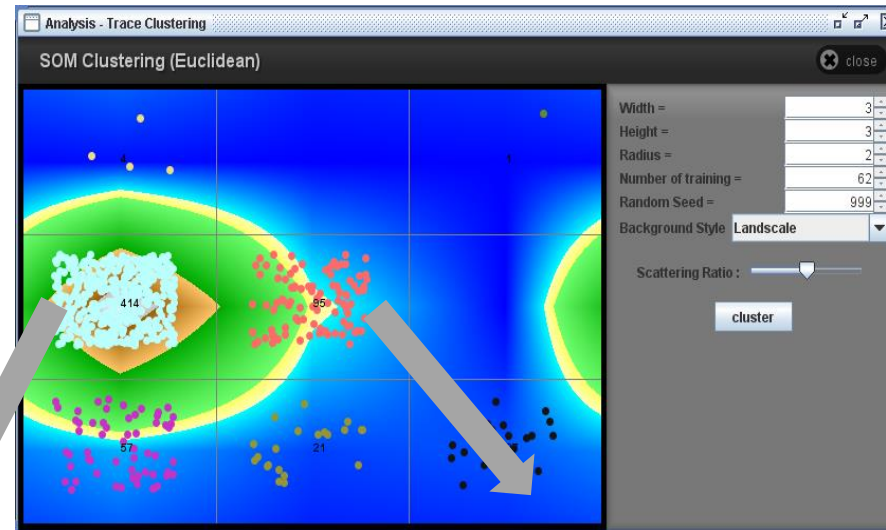
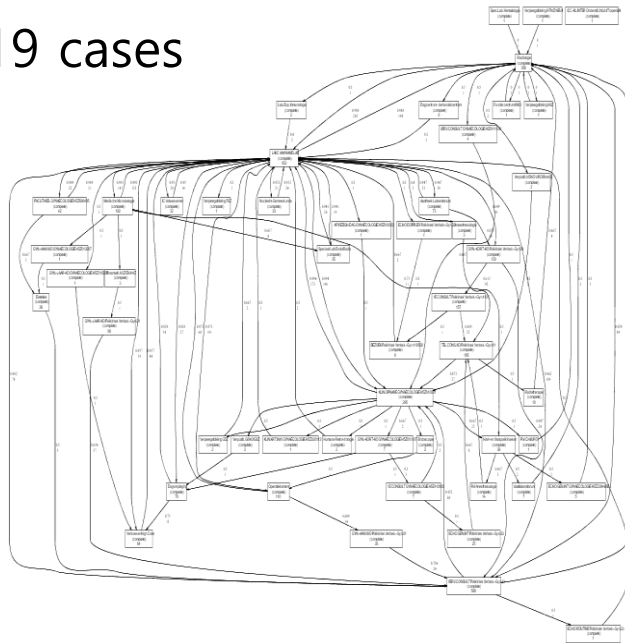
Process Log

Mining result (process model)



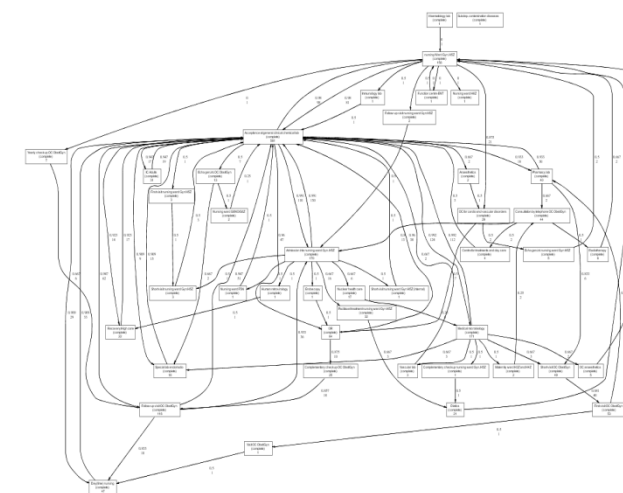
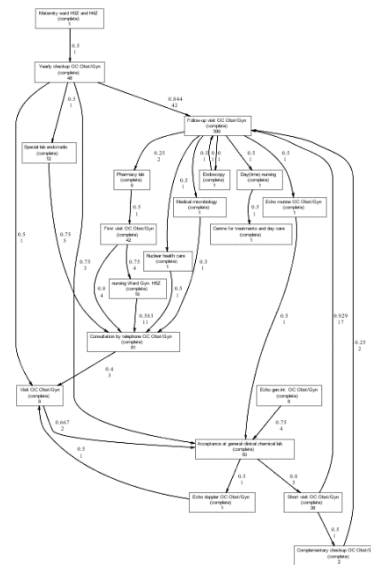
Trace clustering

619 cases



95 cases

414 cases

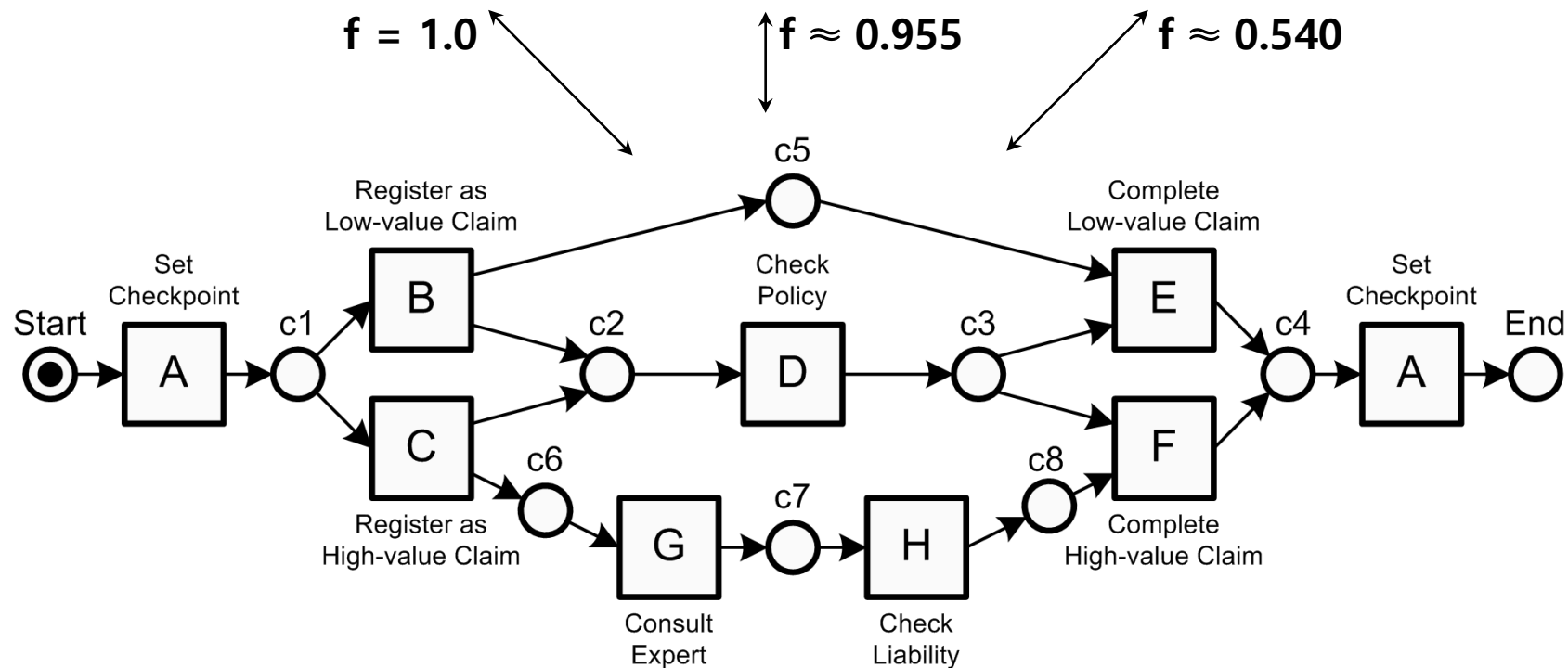


Conformance Checking: measuring fitness

No. of Instances	Log Traces
4070	ABDEA
245	ACDGHFA
56	ACGDHFA

No. of Instances	Log Traces
1207	ABDEA
145	ACDGHFA
56	ACGDHFA
23	AChDFA
28	ACDHFA

No. of Instances	Log Traces
24	BDE
7	AABHF
15	CHF
6	ADBE
1	ACBGDFAA
8	ABEDA



The Two Most Important Quotes In Business

Your #1 business mistake is that you're running your business blind!

"If you can't **measure** it, you can't **improve** it" - Peter Drucker -



Recent works

Visibility

- Model Discovery

Transparency

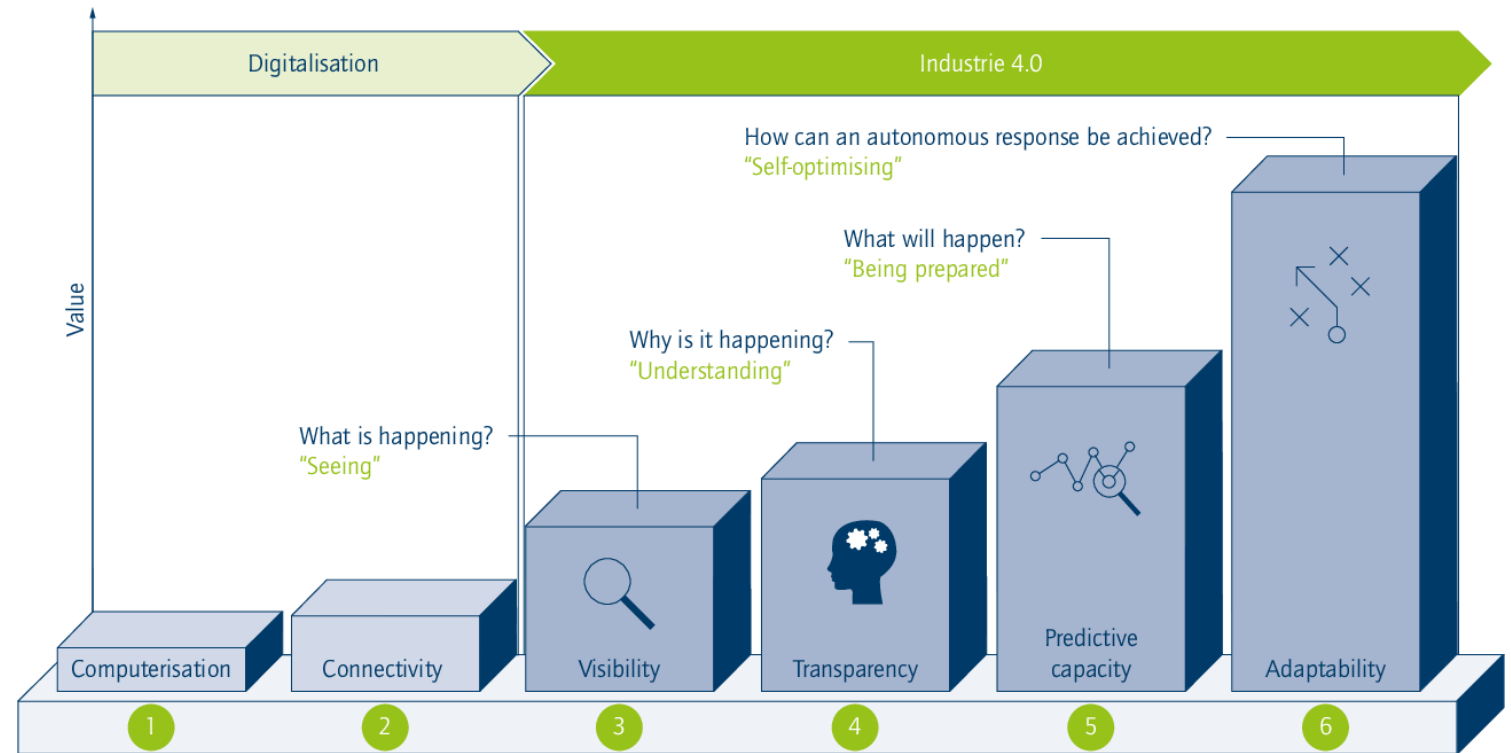
- Process Mining + Machine Learning

Prediction

- Process Mining + AI

Adaptability

- Process Mining + Optimization + Simulation + RPA

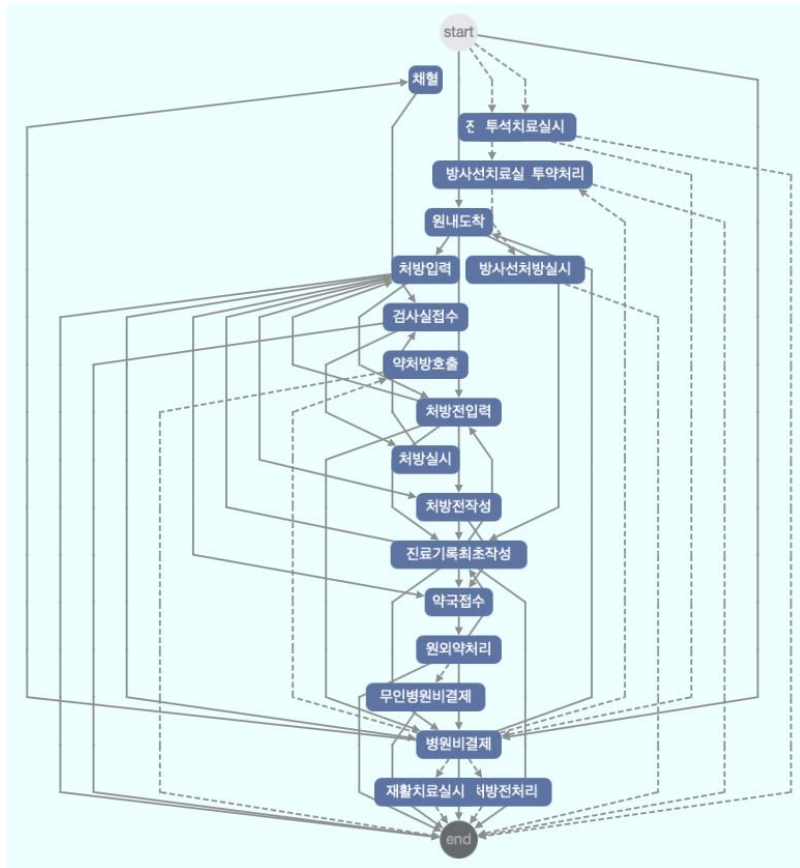


Visualization

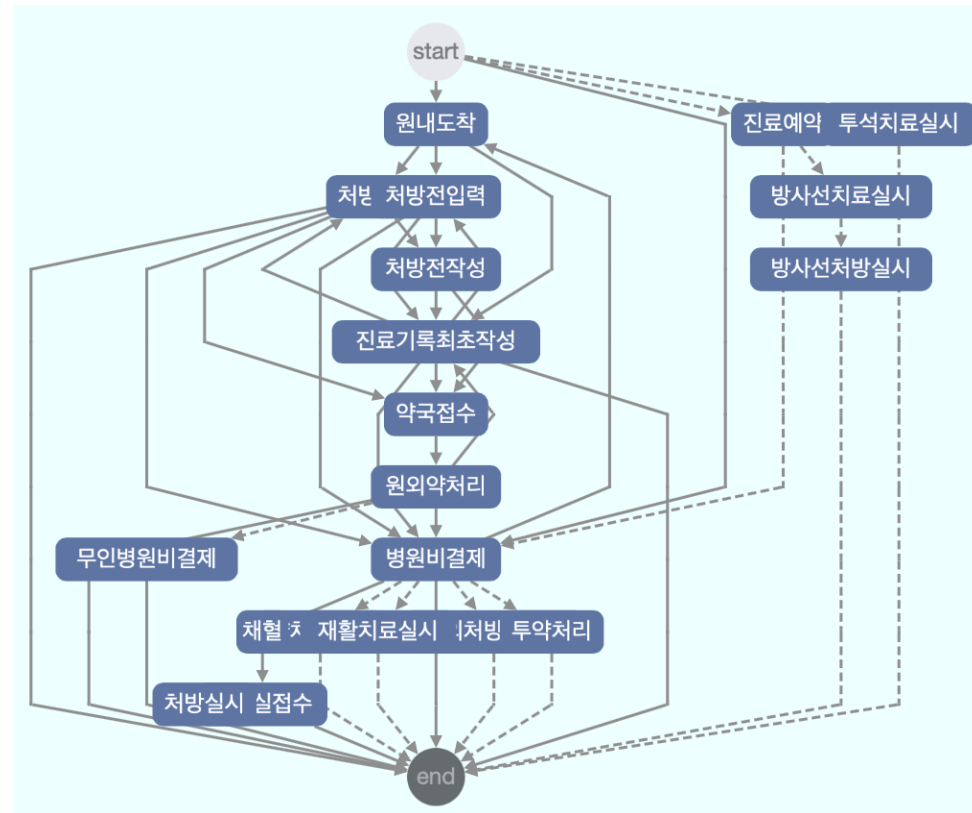
Understandable Process Model Discovery



기존 stable graph layout

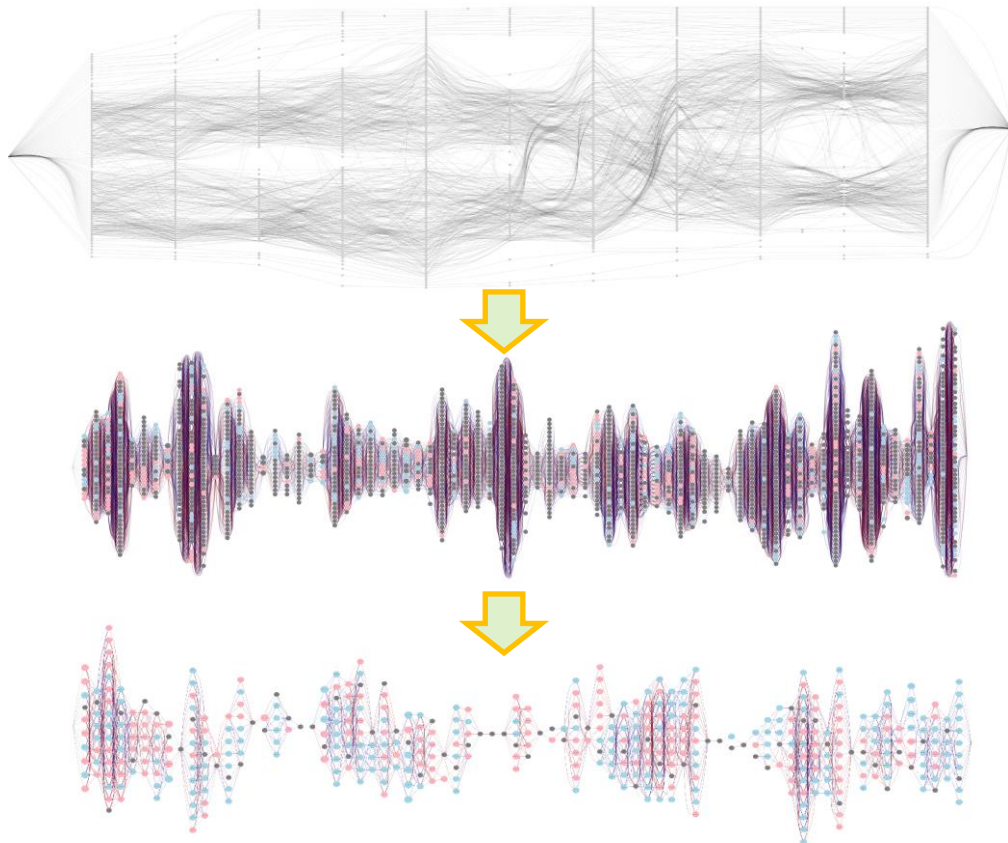


제시 알고리즘



Visualization

Understandable Process Model Discovery

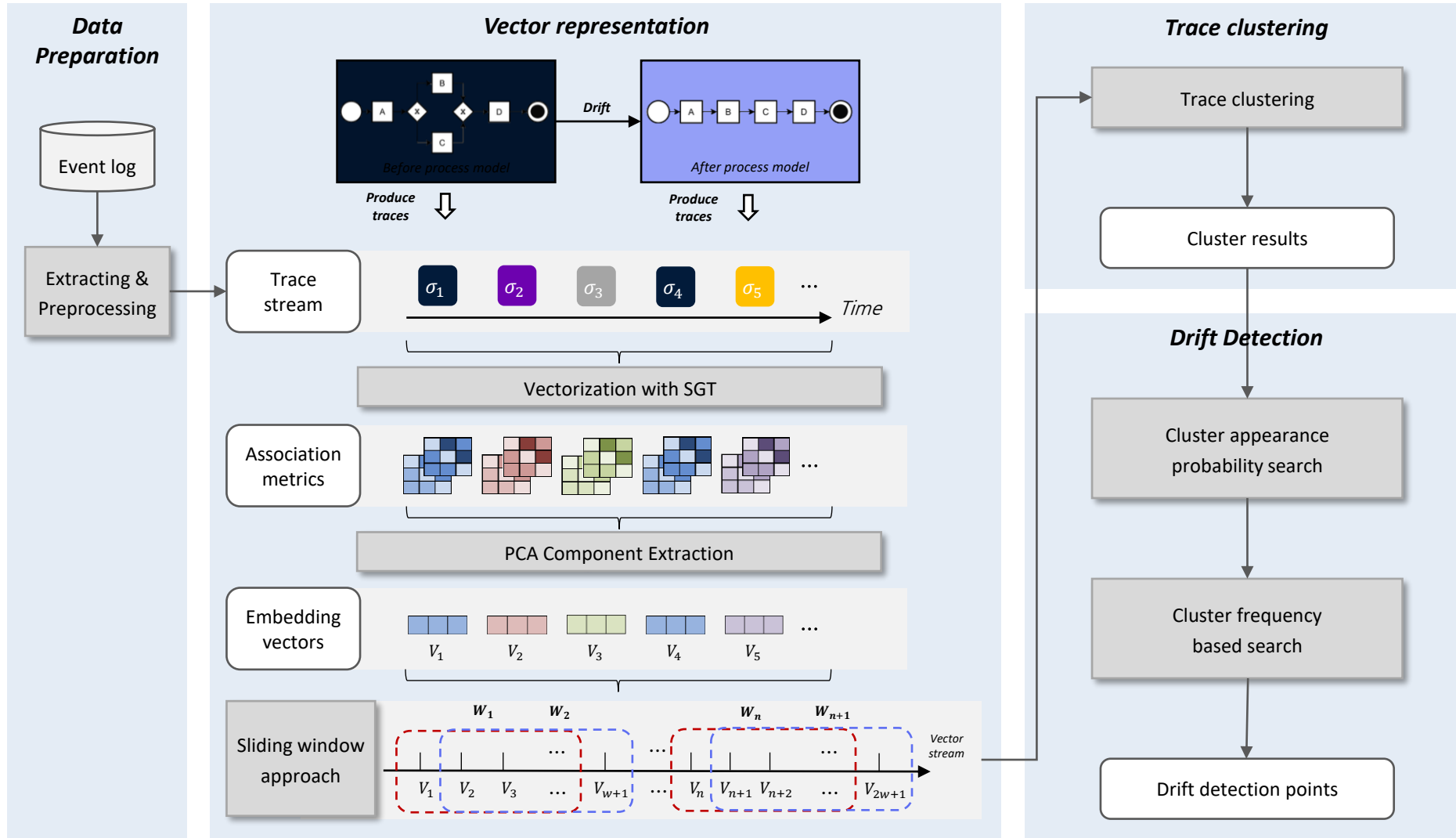


	단순화 전	단순화 후	비율
Node 수	1675	476	71.6% ↓
Arc 수	27462	2347	91.5% ↓

Cho, M., Park, G., Song, M., Lee, J., Lee, B., Kim, E., "Discovery of Resource-oriented Transition Systems for Yield Enhancement in Semiconductor Manufacturing." IEEE Transactions on Semiconductor Manufacturing, Vol. 34, No. 1, pp. 17-24, 2021.

Transparency

Finding Concept Drift

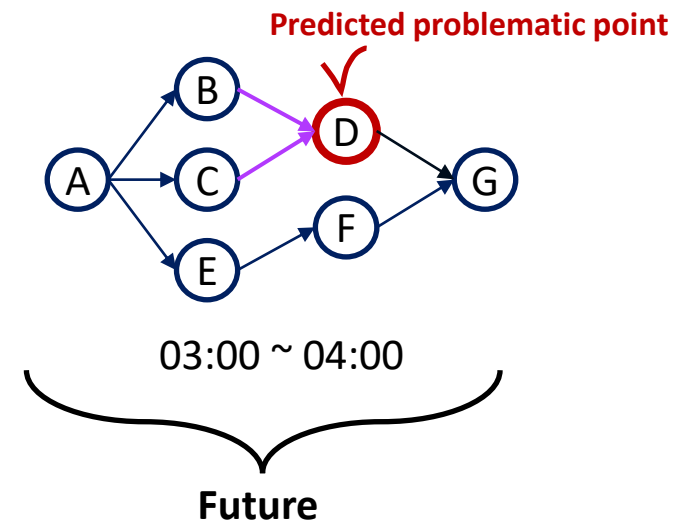
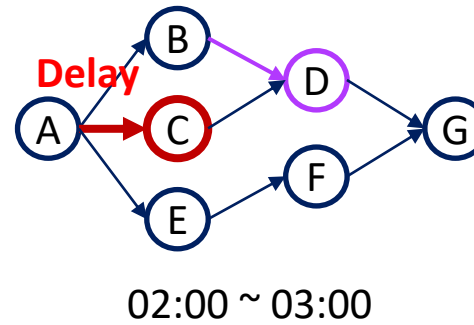
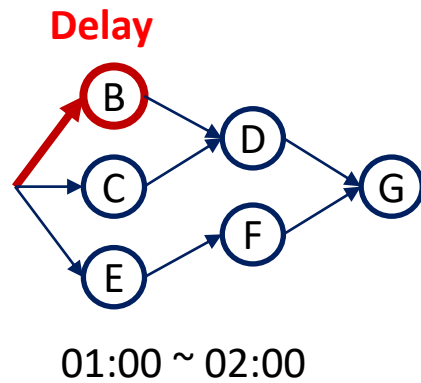


Prediction (PM +AI)

Process performance prediction

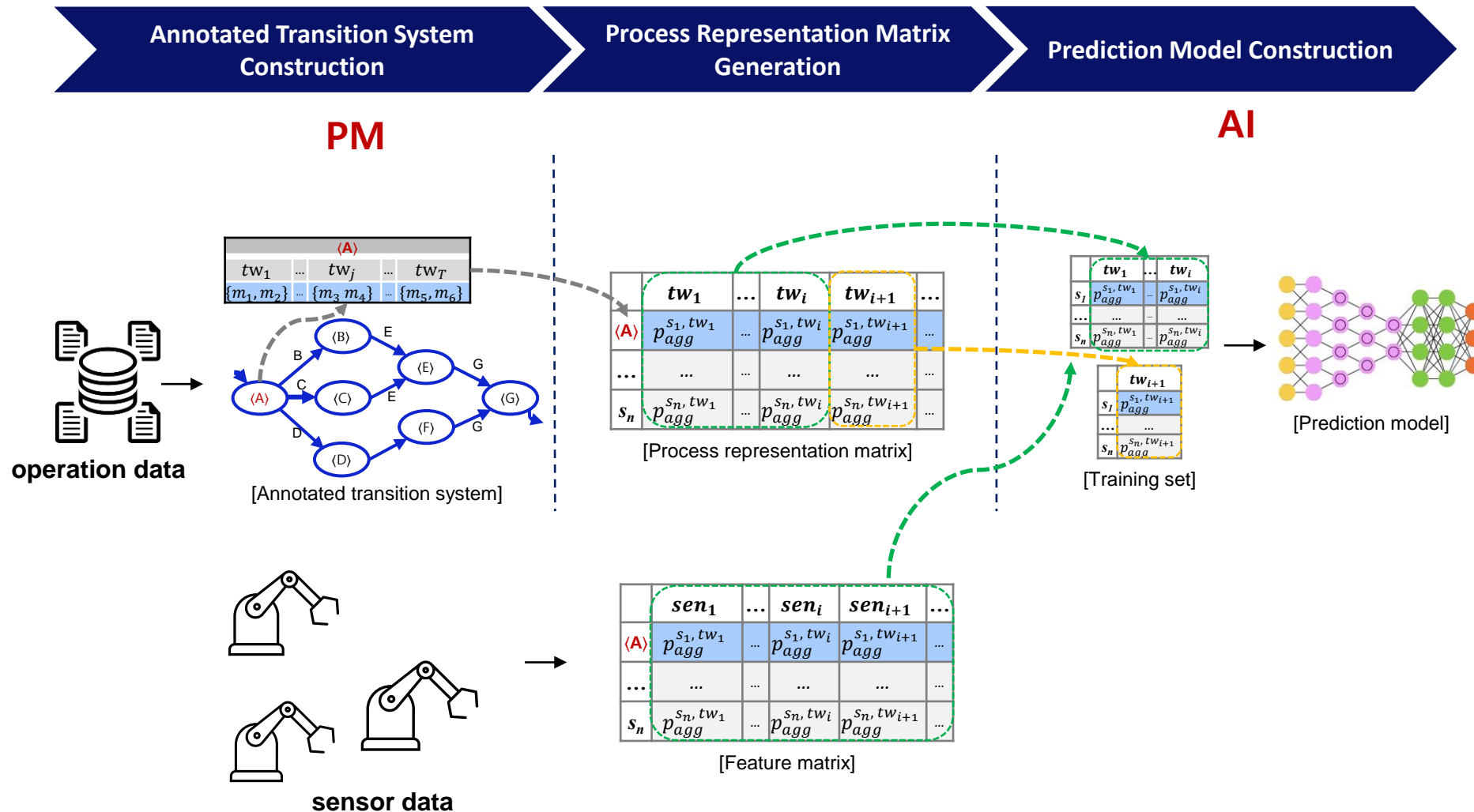
Next event prediction

Pattern: Delay on B → Delay on C → Delay on D

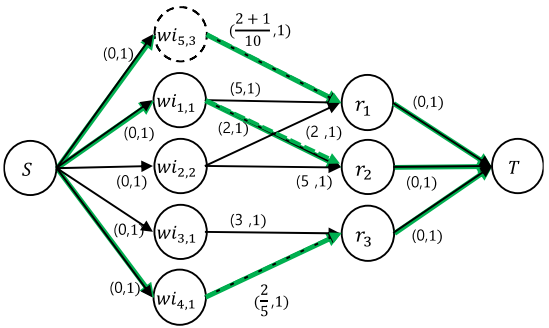
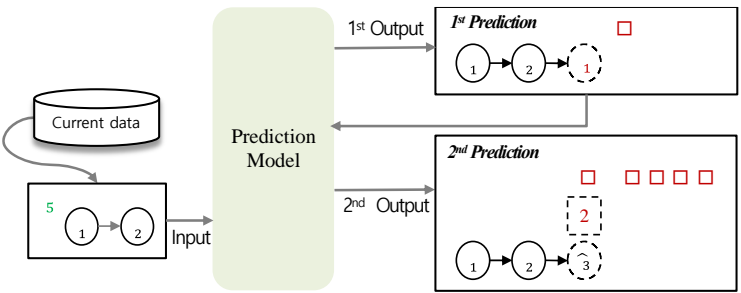


Prediction (PM +AI)

Predicting Performances in Business Processes using Deep Neural Networks



Adaptability(Process Mining + Optimization)



	t	t+1	t+2	t+3	t+4	t+5	t+6	$\sum w_i C_i$
r ₁								65
r ₂								5
r ₃								15

Here

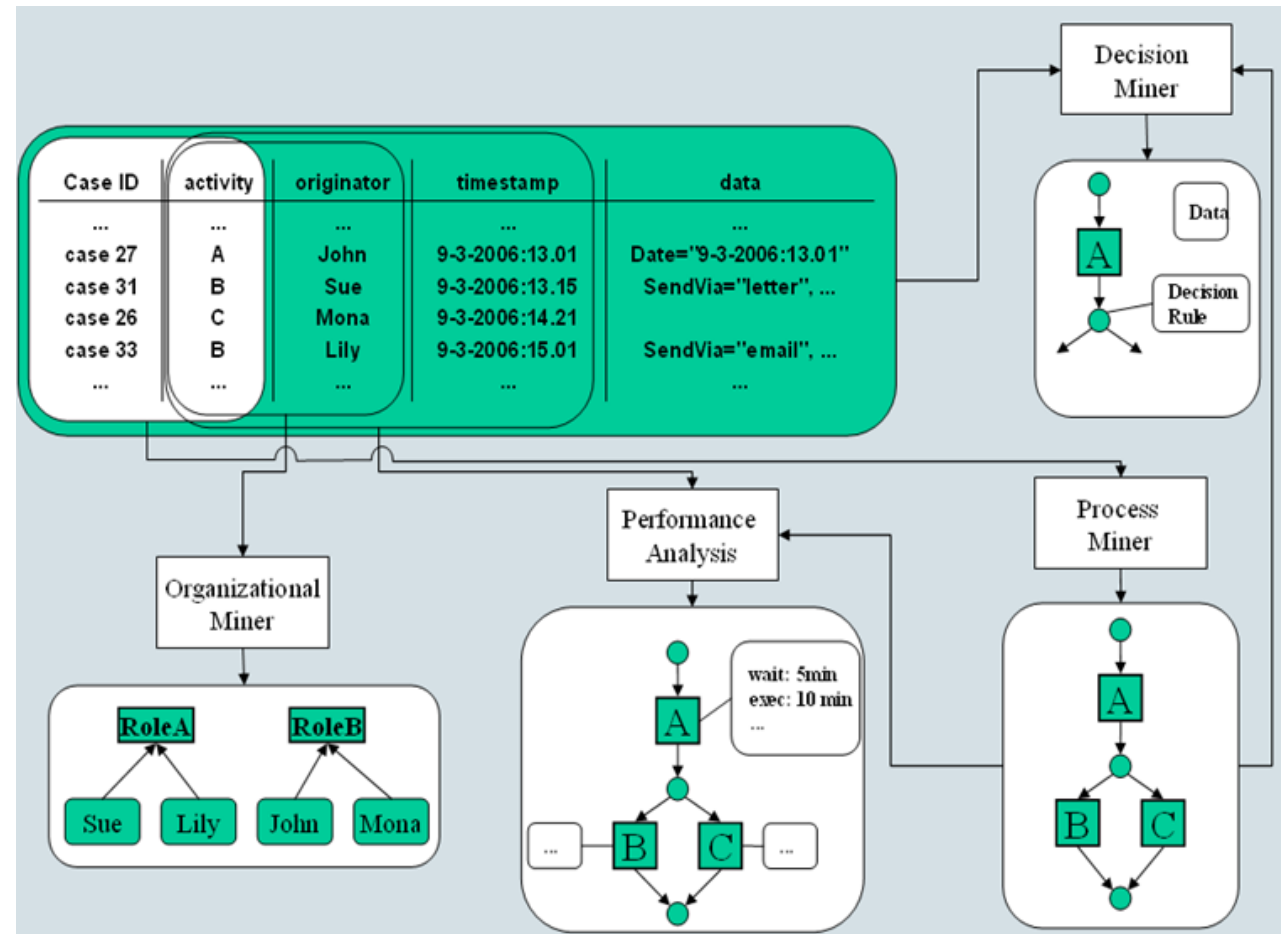
	t	t+1	t+2	t+3	t+4	$\sum w_i C_i$
r ₁						25
r ₂						2
r ₃						15

<Result of resource allocation>

Park, G., M., Song., "Prediction-based Resource Allocation using LSTM and maximum flow and minimum cost algorithm" In *International Conference on Process Mining (ICPM)*, Aachen, Germany, June 24-26, 2019.

Adaptability(Process Mining + Simulation)

Simulation Model Generation

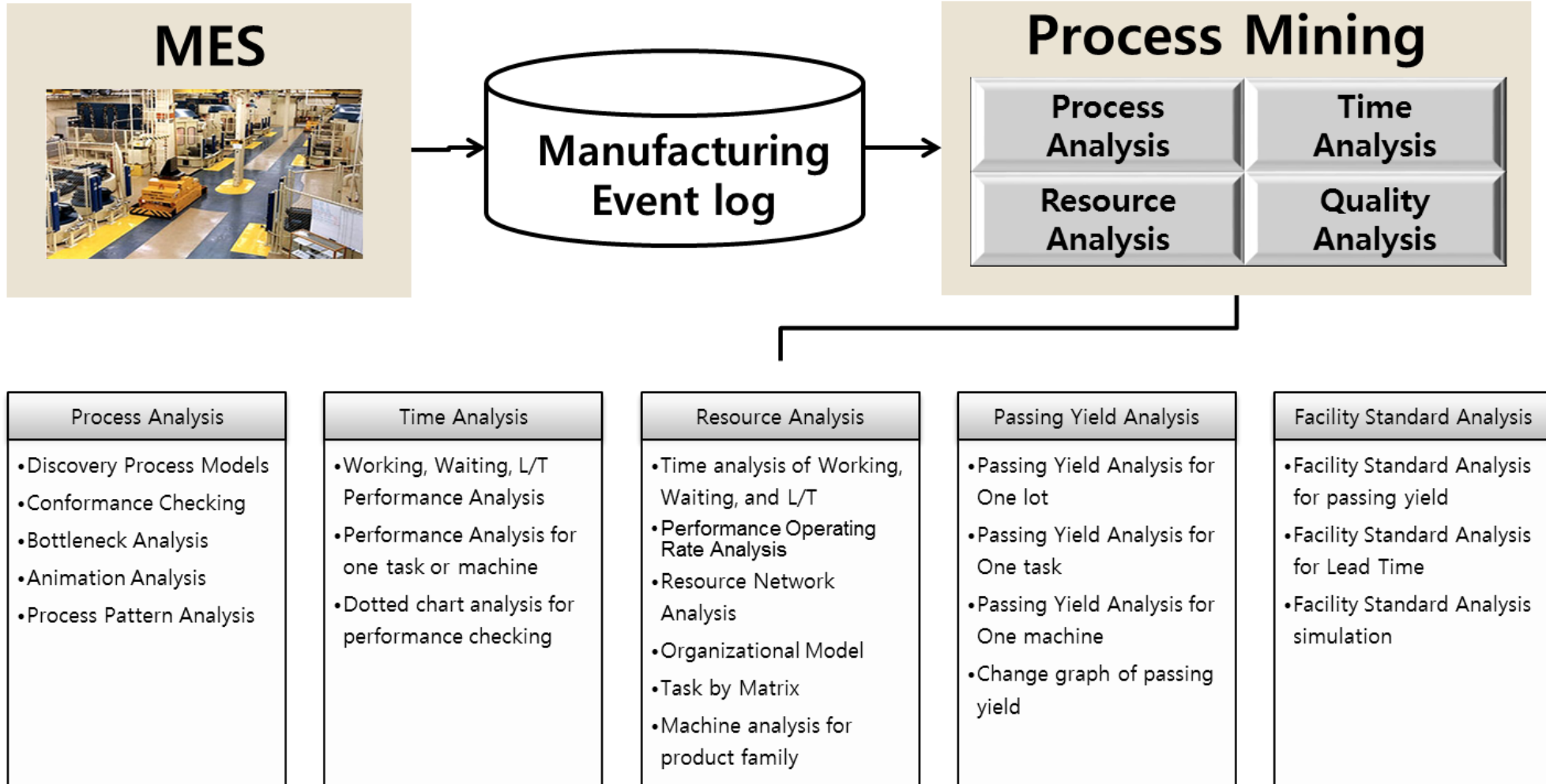


Applications in Korea

- Three major application areas
 - Manufacturing Industry: POSCO, Hyundai Heavy Industries, Samsung Heavy Industries, Samsung electronics, Samsung Electro-Mechanics, etc.
 - Healthcare Industry: SNU Hospital@Bundang, SNU Hospital@Boramae, Samsung Seoul Hospital, etc.
 - Finance Industry: KB



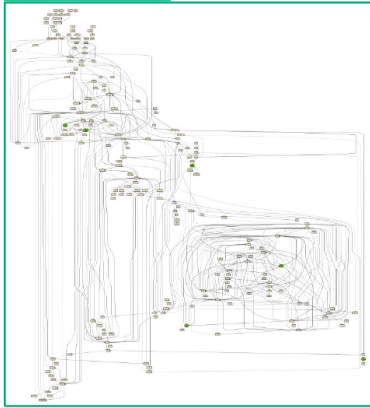
Manufacturing Process Analysis



MES 데이터 분석

Proposed Method – Motivation

Motivation Example

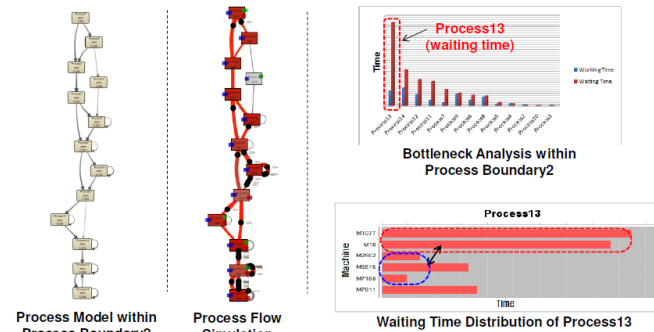


- We extracted a process model from process event logs with ProM.
- Unfortunately, the model is too complicated, so hard to understand.
- To find bottlenecks, we simulated the model with process logs but couldn't get any insight due to model complexity.

Process Model & Bottlenecks

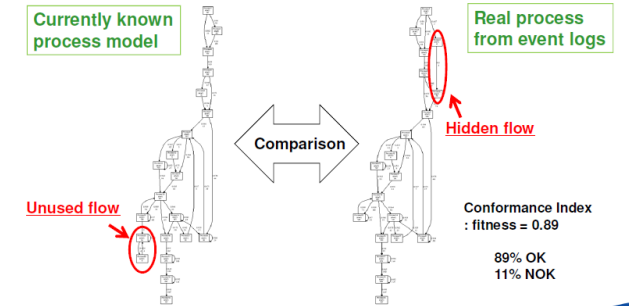
Case Study
 - Process model and bottleneck
 - Model conformance
 - Process visualization

- Step5. (Optional) Repeat Step3-4 to find a narrower bottleneck within each process boundary.



Process Mining Framework

- Phase II: Conformance Test
 - Find the difference between the currently known process model and the real event logs.
 - From this analysis, (1) unused flow and (2) hidden flow are found.

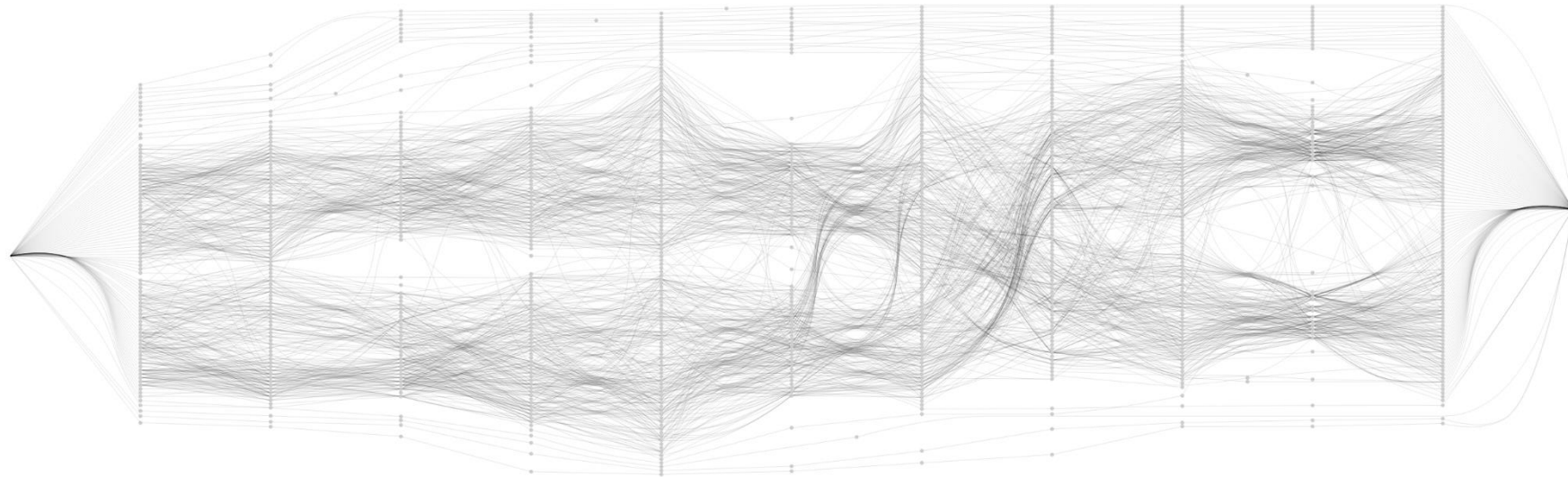


삼성전자 정일교 박사

BOB/WOW: 수율 기반 최적 설비 경로 도출 방법론

Transition System 모델 도출

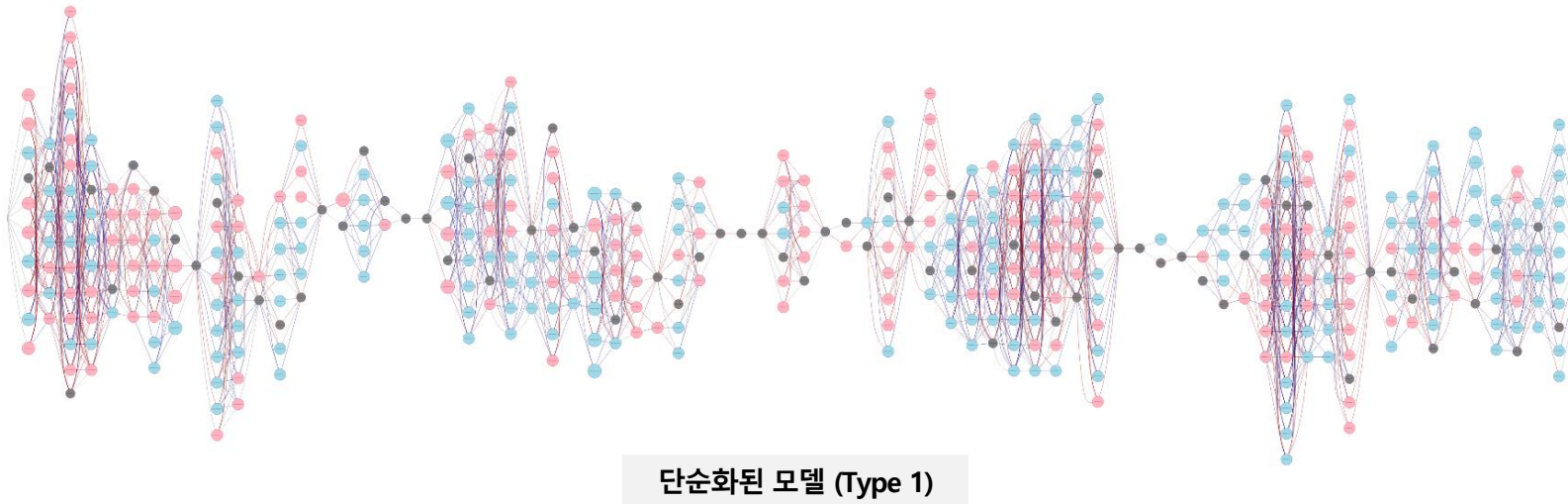
- 실제 공정 로그에서 생성된 FSM 모델은 일반적으로 매우 복잡하여 중요 설비를 고려한 모델의 단순화 필요
- 아래 예시는 전체 공정 중 11개의 공정에 대한 모델 도출 결과임



<실제 공정 로그를 활용한 FSM Model 도출 결과>

BOB/WOW: 수율 기반 최적 설비 경로 도출 방법론

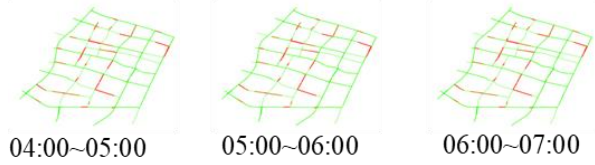
유의/더미 설비 분석 기발 설비 프로세스 모델 단순화



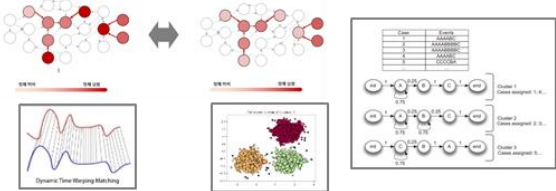
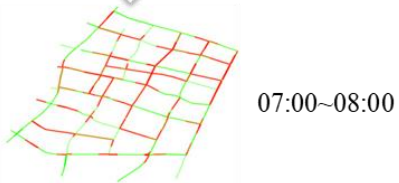
	단순화 전	단순화 후	비율
Node 수	1675	476	71.6% ↓
Arc 수	27462	2347	91.5% ↓

Congestion analysis in FAB

반도체 물류 내 비효율을 효과적으로 감지하고 이러한 비효율의 원인을 분석하여 개선을 위한 인사이트를 제공하는 실용적인 방법 개발



Predict

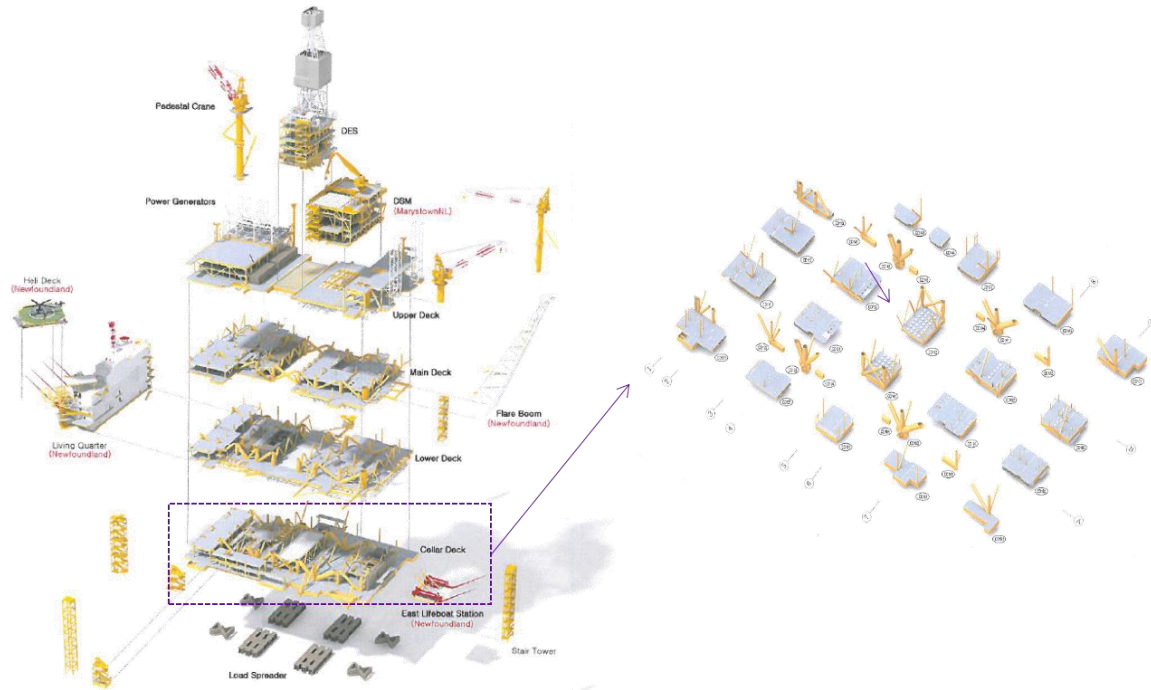


[Congestion analysis & prediction]

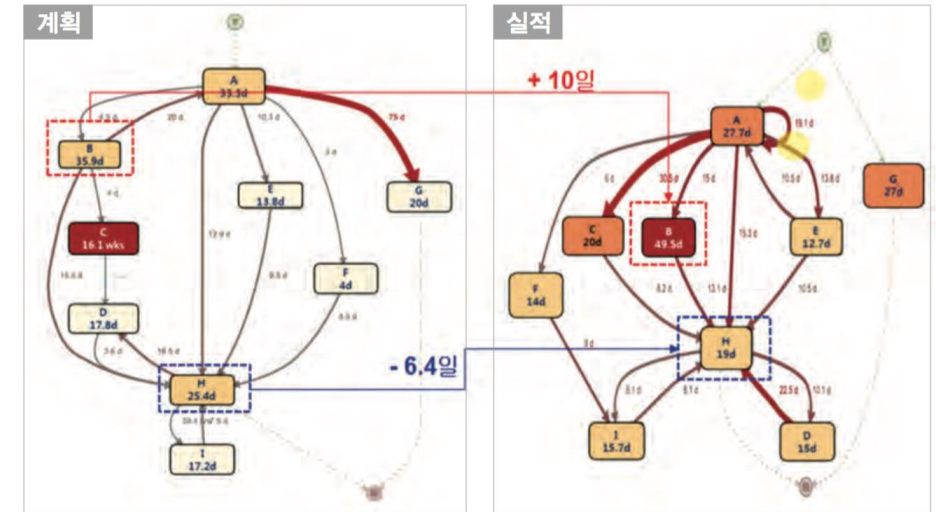


[복잡한 물류 흐름 예시]

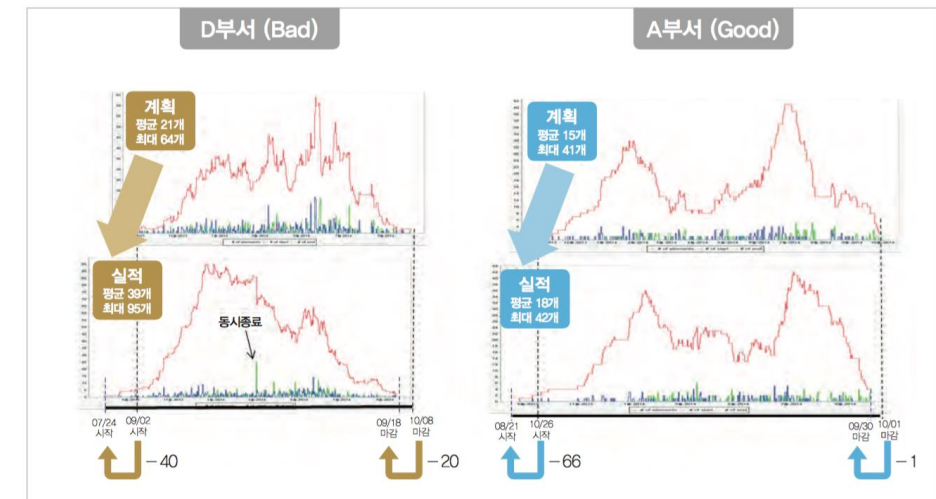
Bottleneck Analysis



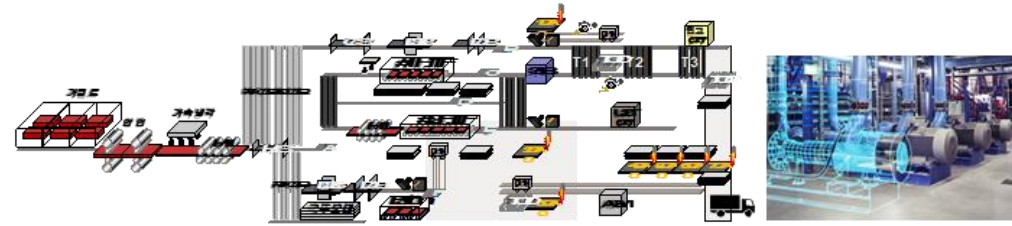
[그림] A 중공업 계획 실적 프로세스 모델 비교



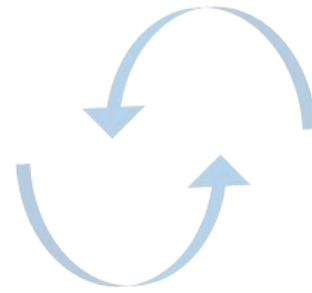
[그림] 부서별 부하 비교 분석



Process Mining & Digital Twin



3D Model + MR



Digital Twin



IoT



Simulation

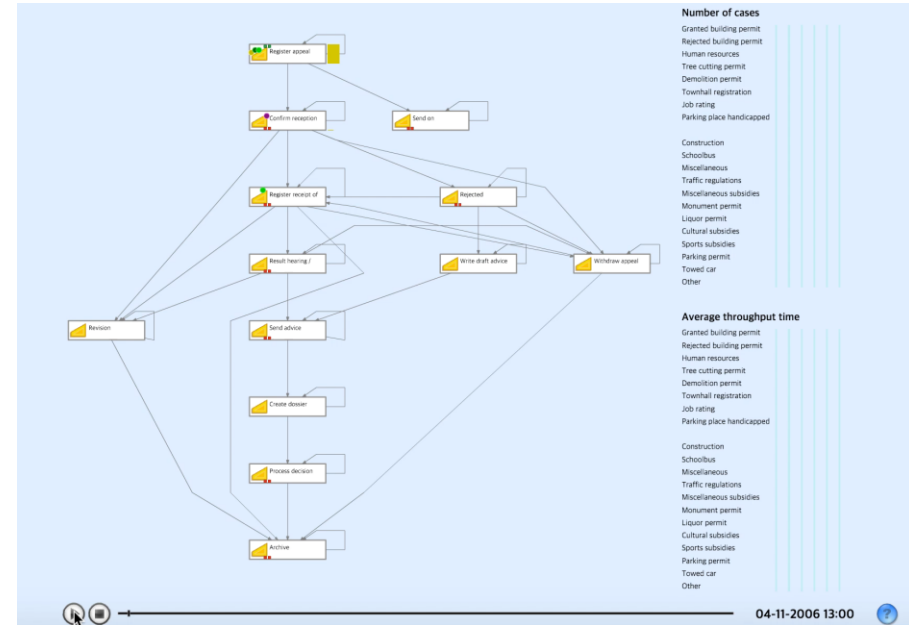
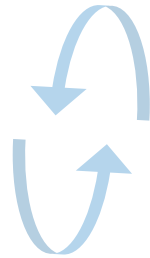


AI+ML



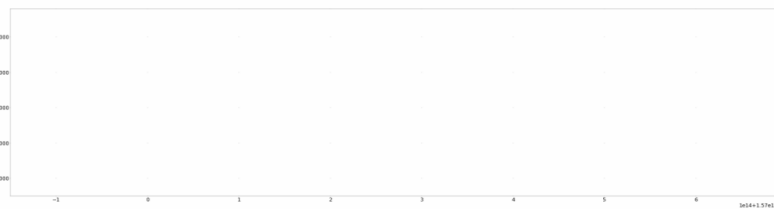
Optimization

Process Mining & Digital Twin

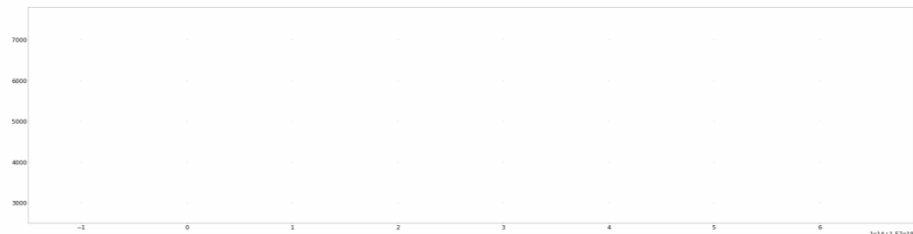


Futura

Visualization (Process Discovery + Animation)



Digitalization (IoT)

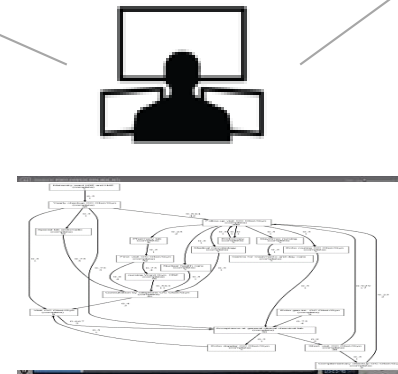
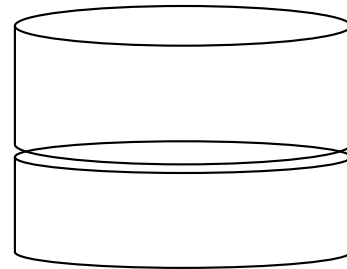
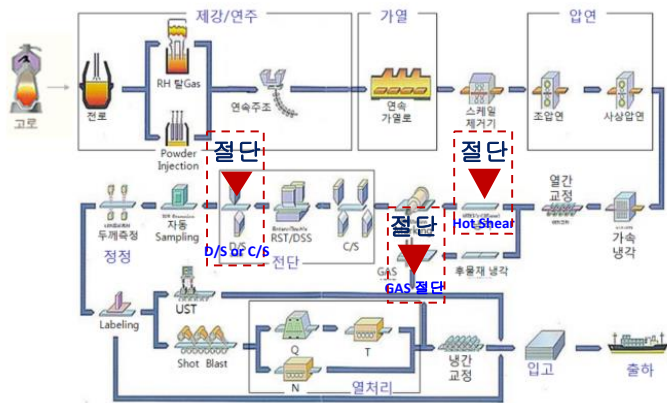


Prediction + Optimization + Simulation

Process Mining & Digital Twin

Adaptability - Self-Optimizing

압연 일자(일)	1	2	3	4	5	6	7	8	9	10	11	12	13	14
원 편성안	1	2	3	4	5	6	7	8	9	10	11	12	13	14
최적 편성안 (예시)	12	11	13	10	14	9	7	6	2	4	3	1	8	5



디지털 트윈 Platform




편성안 검증

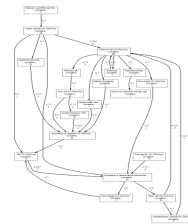
Process Mining & Digital Twin

Adaptability - Self-Optimizing

일자(일)	1	2	3	4	5	6	7	8	9	10	11	12	13	14
원 스케줄	1	2	3	4	5	6	7	8	9	10	11	12	13	14
Random 스케줄	10	11	4	2	12	3	9	13	6	5	8	14	7	1
LP 스케줄	1	11	4	5	8	2	6	12	3	7	13	9	10	14
AI 스케줄	5	14	6	1	12	2	9	3	13	7	10	4	11	8

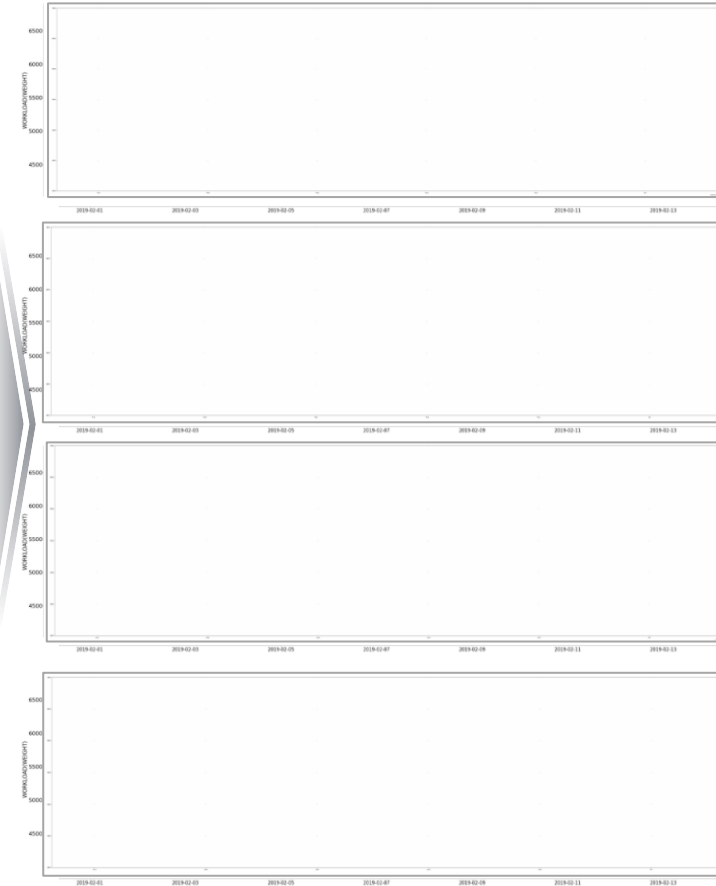
 : Roll 편성(1 turn)

Digital Twin



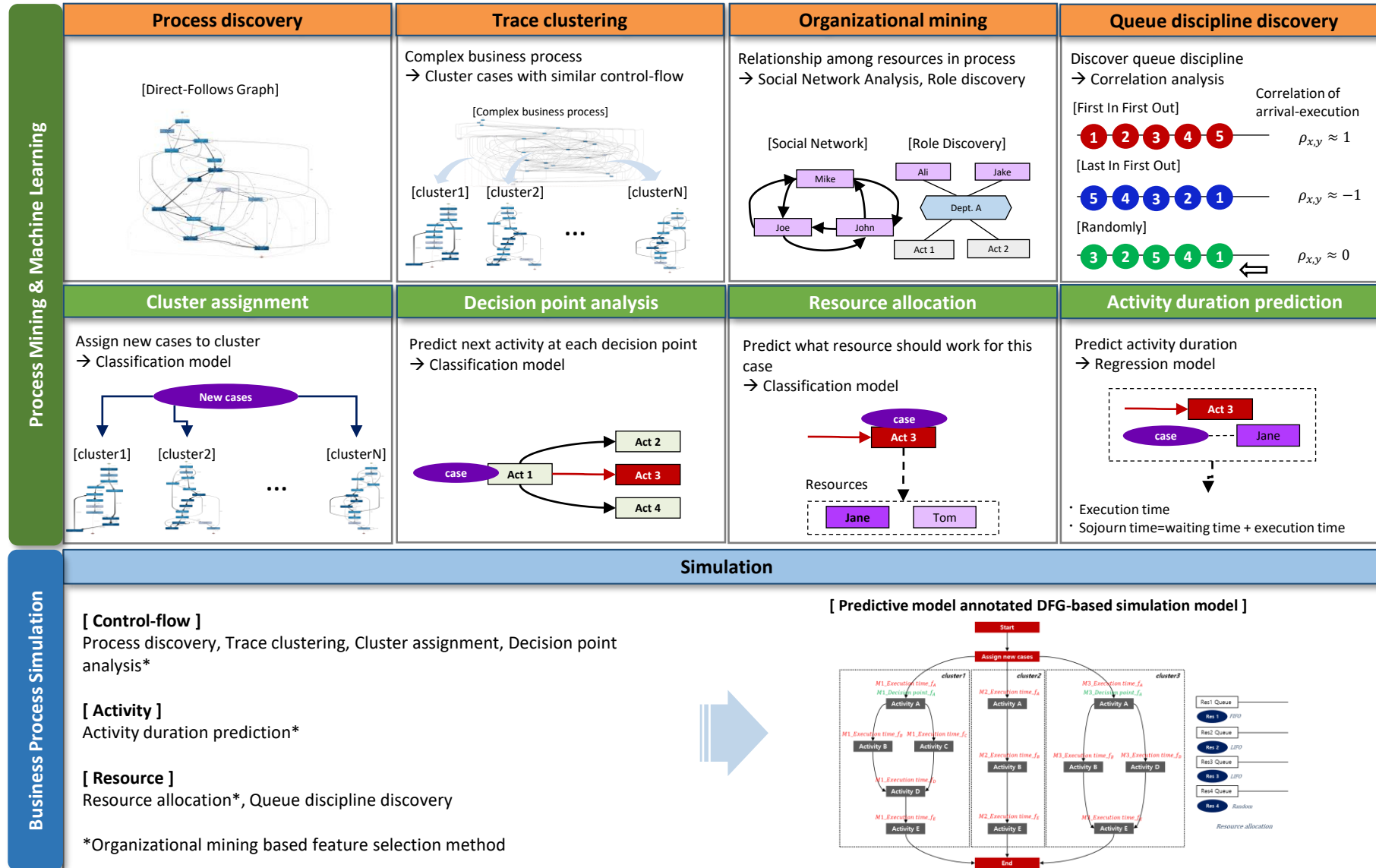
예측 공정
부하량

2019-02-01~2019-02-14



Process Mining & Digital Twin

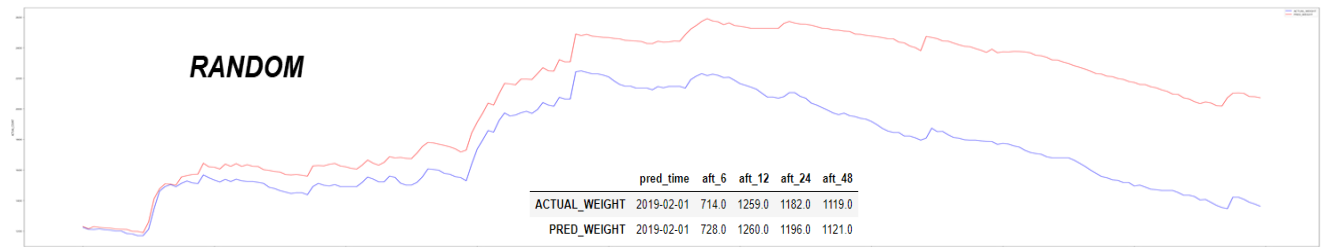
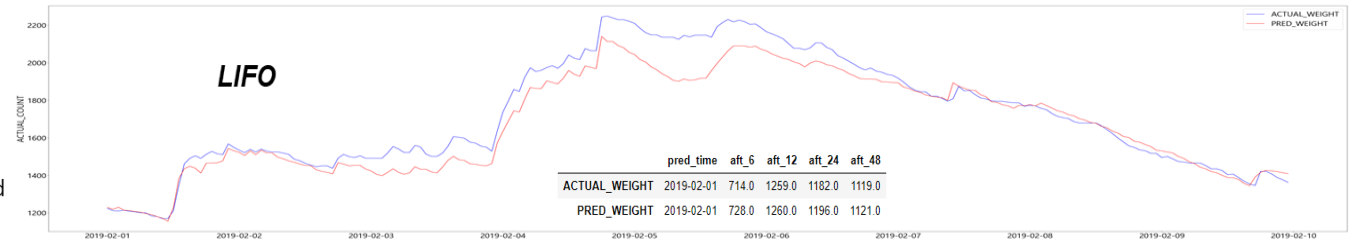
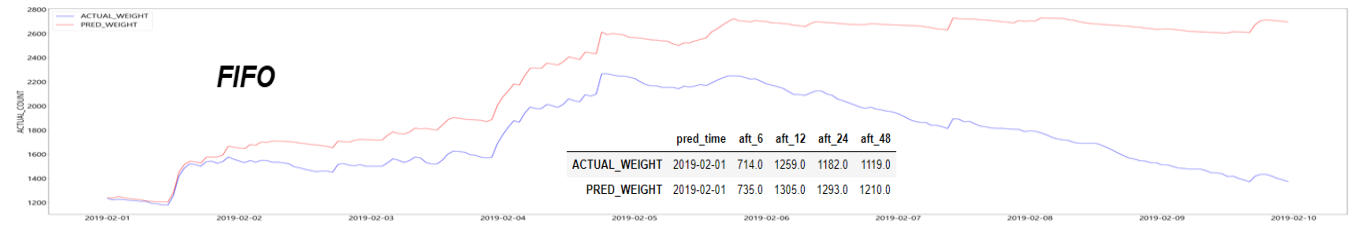
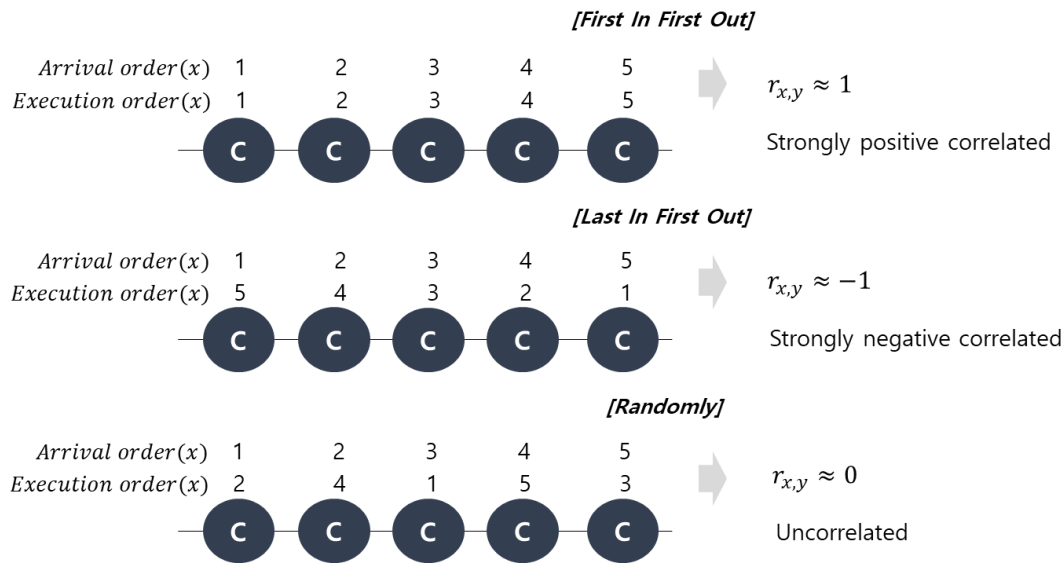
Simulation model generation framework



Process Mining & Digital Twin

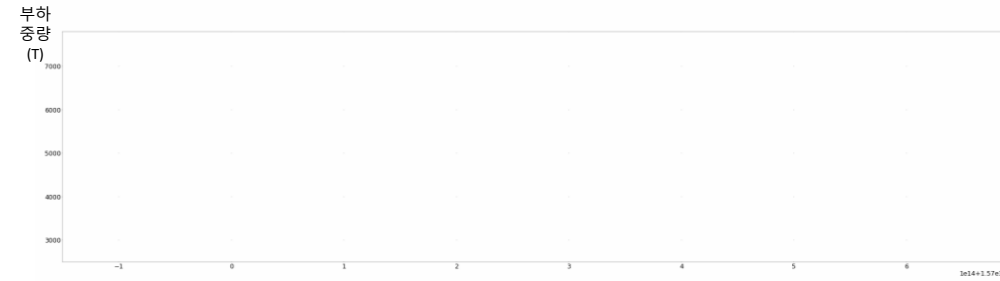
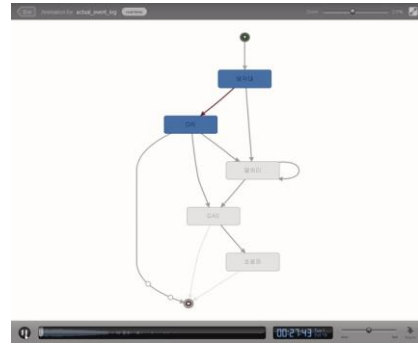
Simulation model generation framework

대기열 정책 발견 알고리즘(Queue Discipline Discovery)

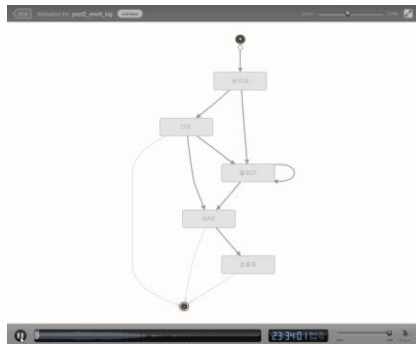


Process Mining & Digital Twin

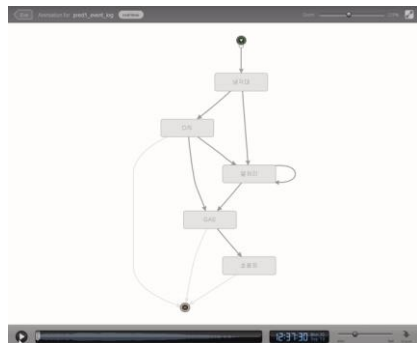
실제로그



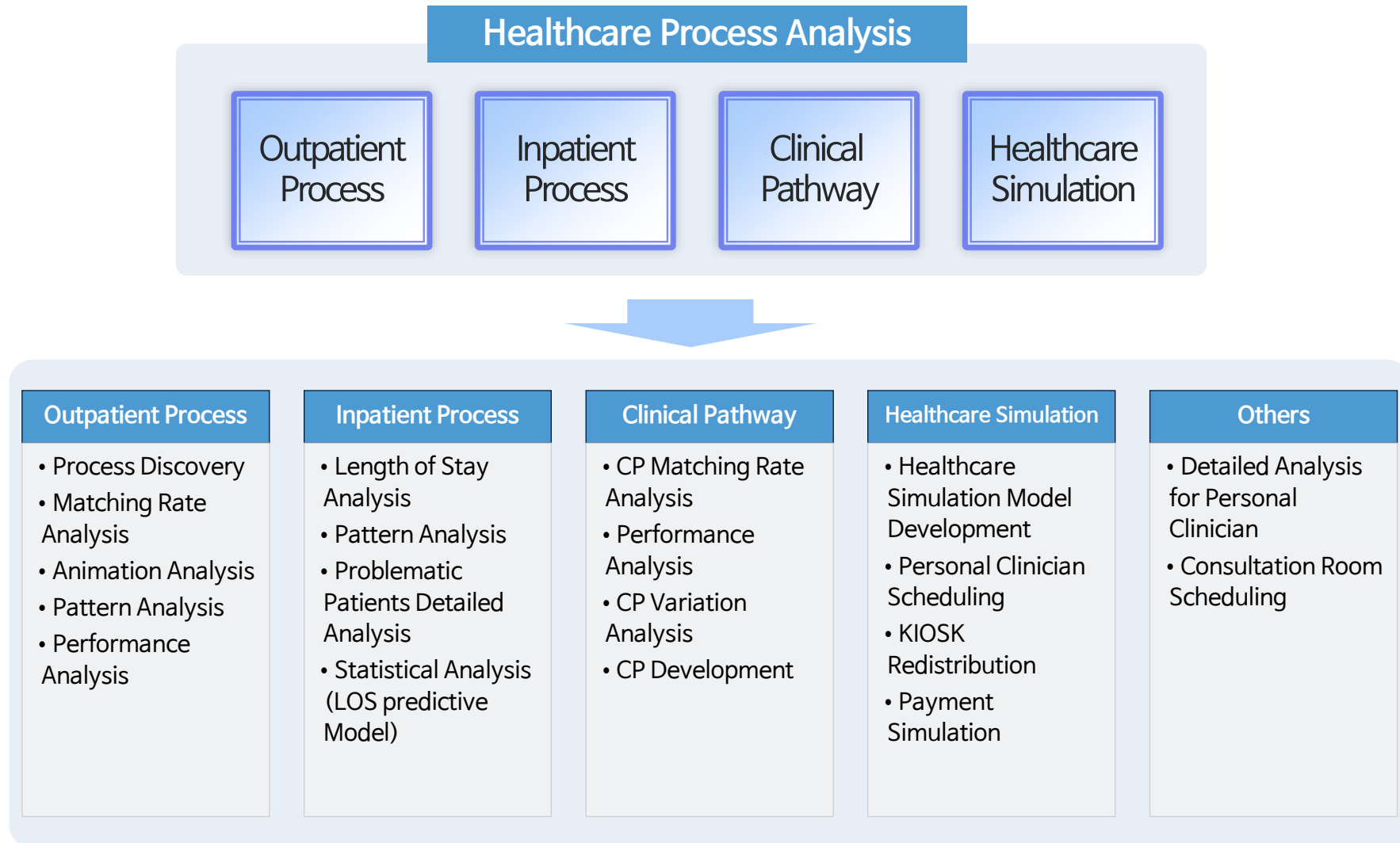
프로세스 마이닝
기반 예측
(초기 모델)



프로세스 마이닝
기반 예측
(개선된 모델)



Healthcare Process Analysis



의료 프로세스 분석

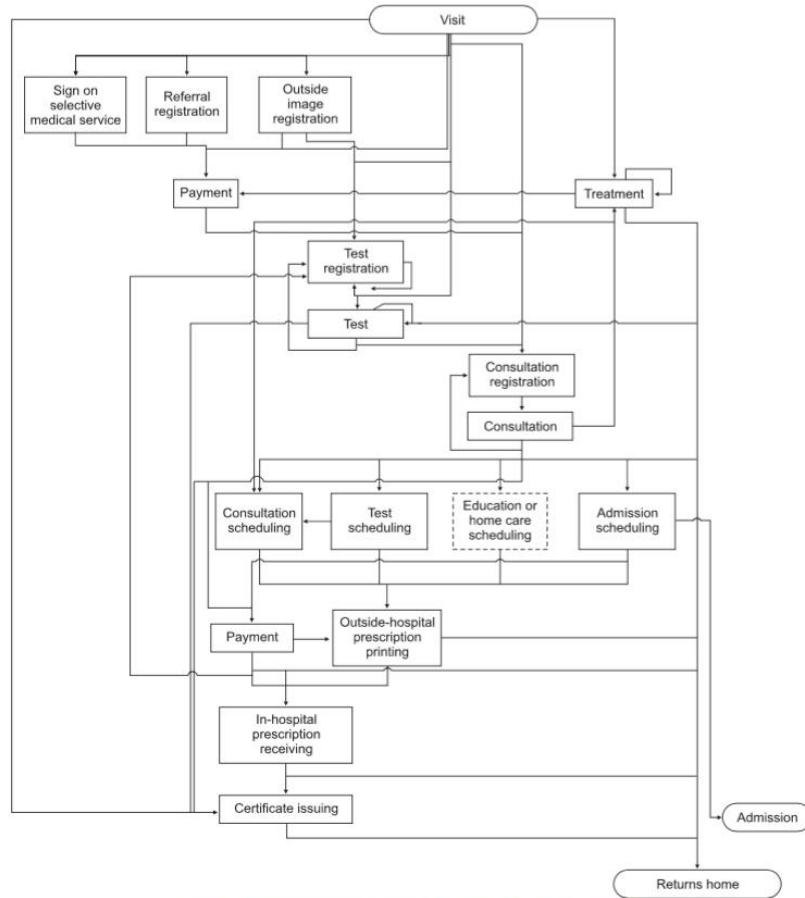


Figure 1. Outpatient care mega-process model derived from domain experts.

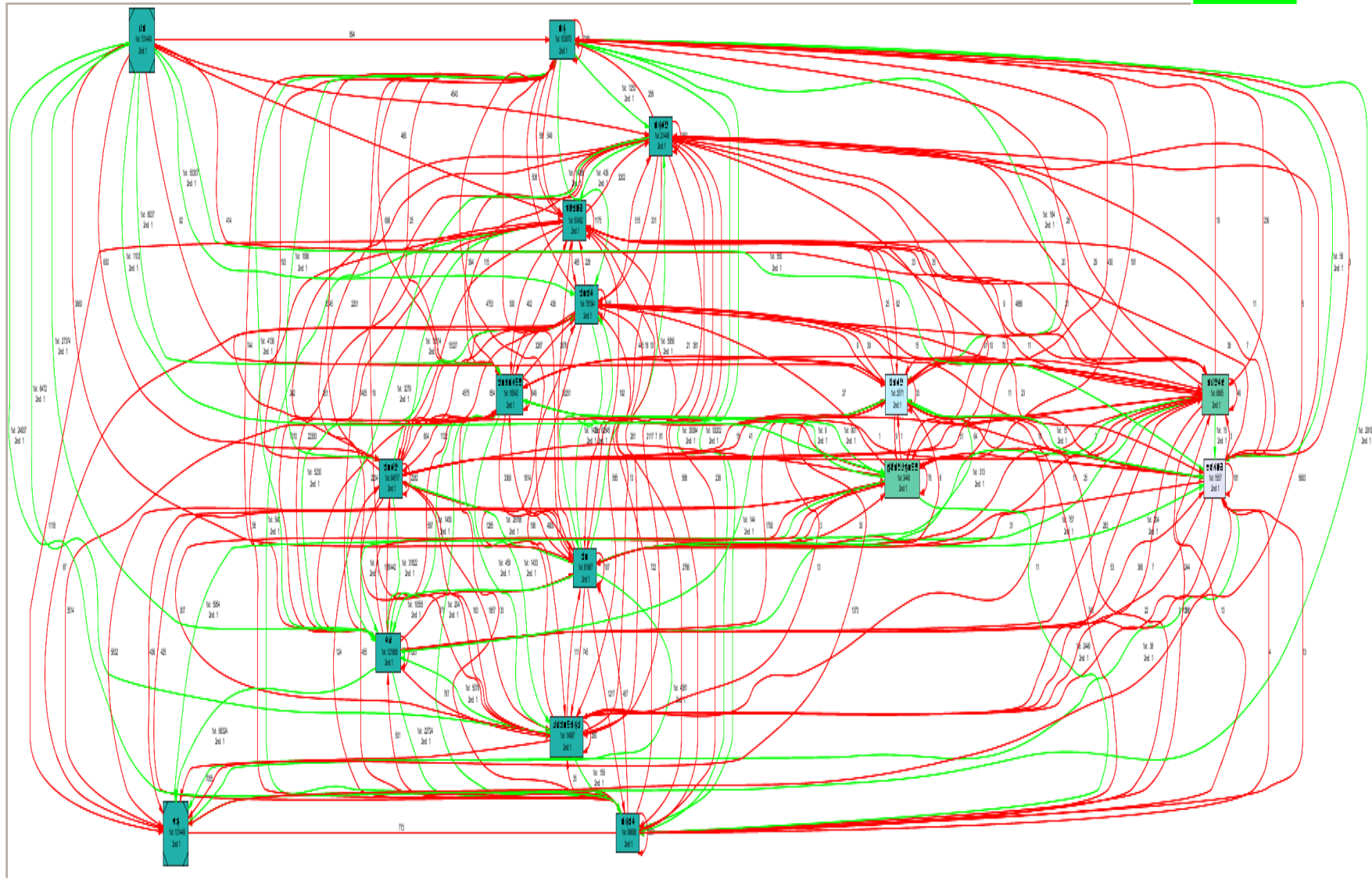
SNUB2012

파일 | 홈 | 삽입 | 페이지 레이아웃 | 수식 | 데이터 | 검토 | 보기 | 추가 기능

외부 데이터 가져오기 | 정렬 및 필터

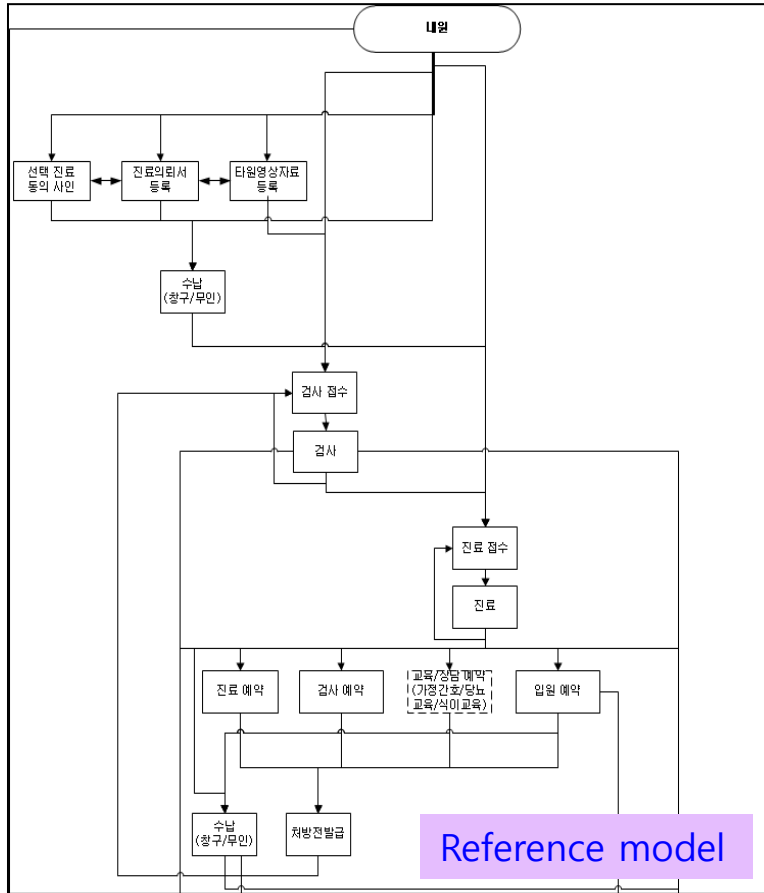
	A	B	C	D	E	F
1	CASEPID	ACTIVITYOID	RESOURCEOID	RESOURCEDEPT	EXECUTEDAT	RESERVEDAT
2	A001	검사접수	P001	내과	2012-05-09 9:54	2012-05-09 0:00
3	A002	검사	P001	내과	2012-05-09 9:54	2012-05-09 0:00
4	A002	검사접수	P001	내과	2012-05-09 9:54	2012-05-09 0:00
5	A003	검사	P001	내과	2012-05-09 9:54	2012-05-14 0:00
6	A003	검사접수	P001	내과	2012-05-09 9:54	2012-05-14 0:00
7	A004	진료접수	P002	소아과	2012-05-09 9:54	2012-05-09 10:10
8	A005	진료접수	P003	부인과	2012-05-09 9:54	2012-05-09 10:10
9	A006	수납	C001	원무	2012-05-09 9:54	

의료 프로세스 분석 - 모델 비교

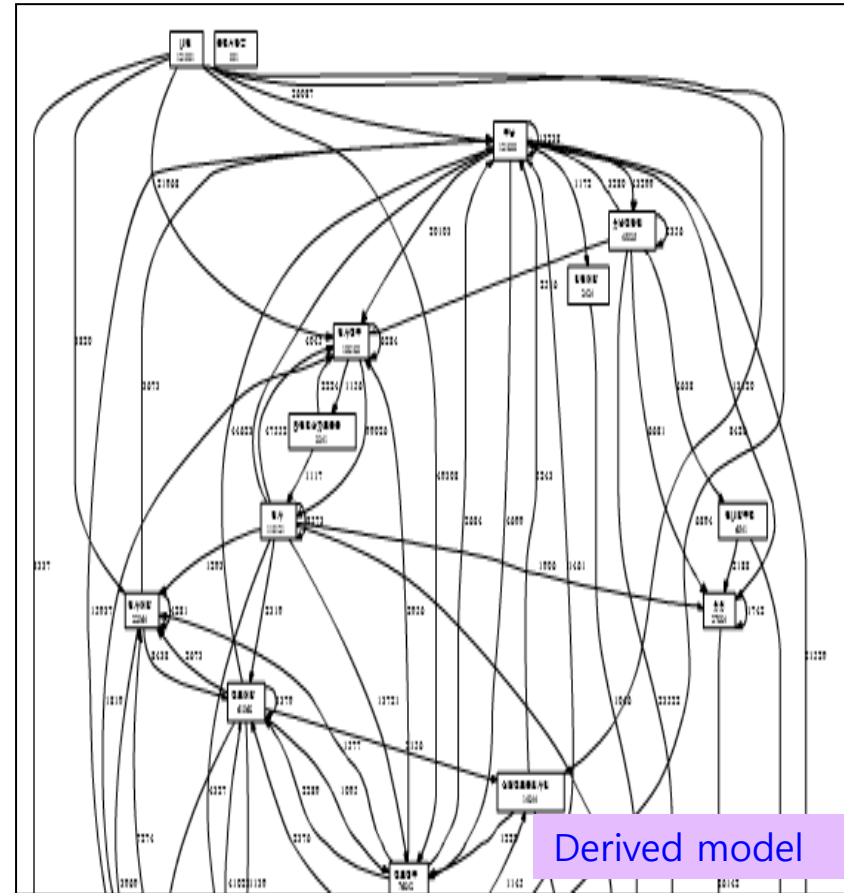


표준 의료 프로세스 정합도 측정

matching rate = 89.01%



Reference model

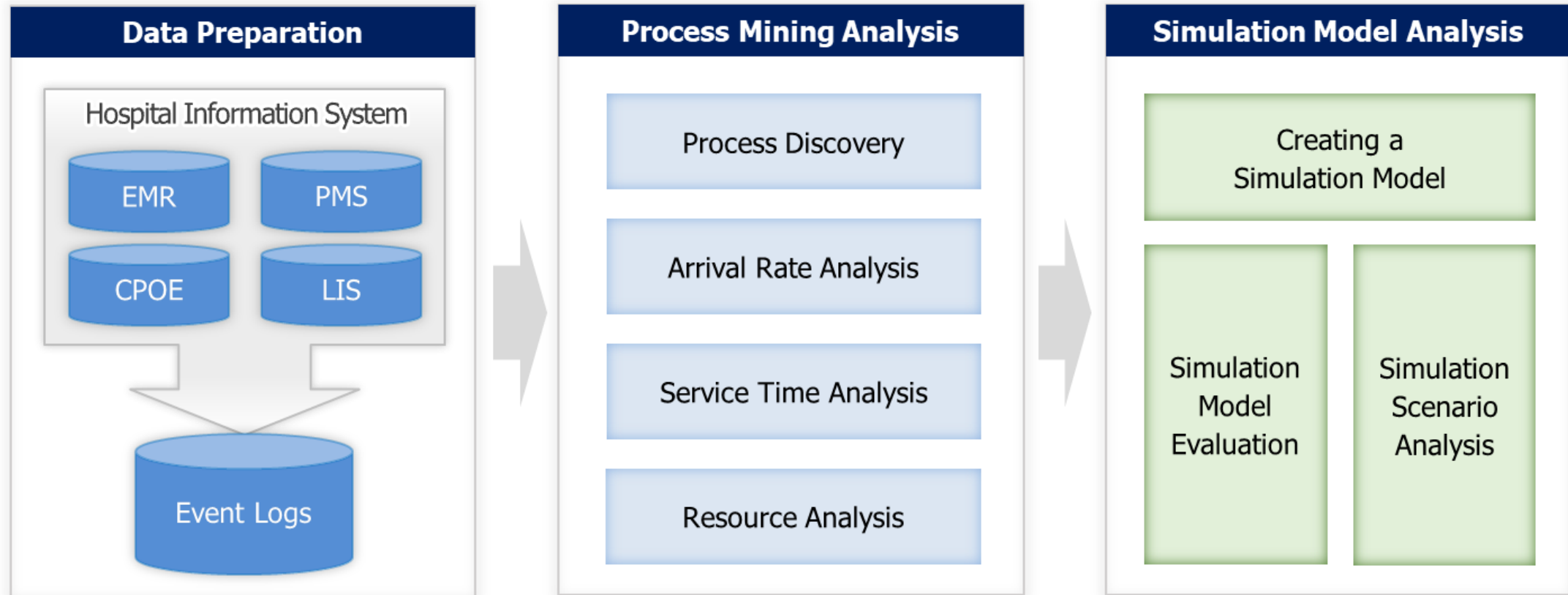


Derived model

미수납 패턴

Performance Sequence Analysis		Frequency	Time (Min)
p0	Consulting room assignments → Consultation	99	135
p1	Registration of an outside referral document → Registration of an outside referral document	16	38
p2	Consulting room assignments → Consultation → Issuance of prescription → Reservation for visit → Payment	16	71
p3	Preparation for test → Test → Payment → Test	13	679
p4	Consulting room assignments → Consultation → Payment	10	267
p5	Consultation → Consulting room assignments → Preparation for test → Test → Reservation for visit → Payment	10	63
p6	Consulting room assignments → Consultation → Reservation for visit → Payment	9	93

Healthcare Process Simulation: Process Simulation Methodology



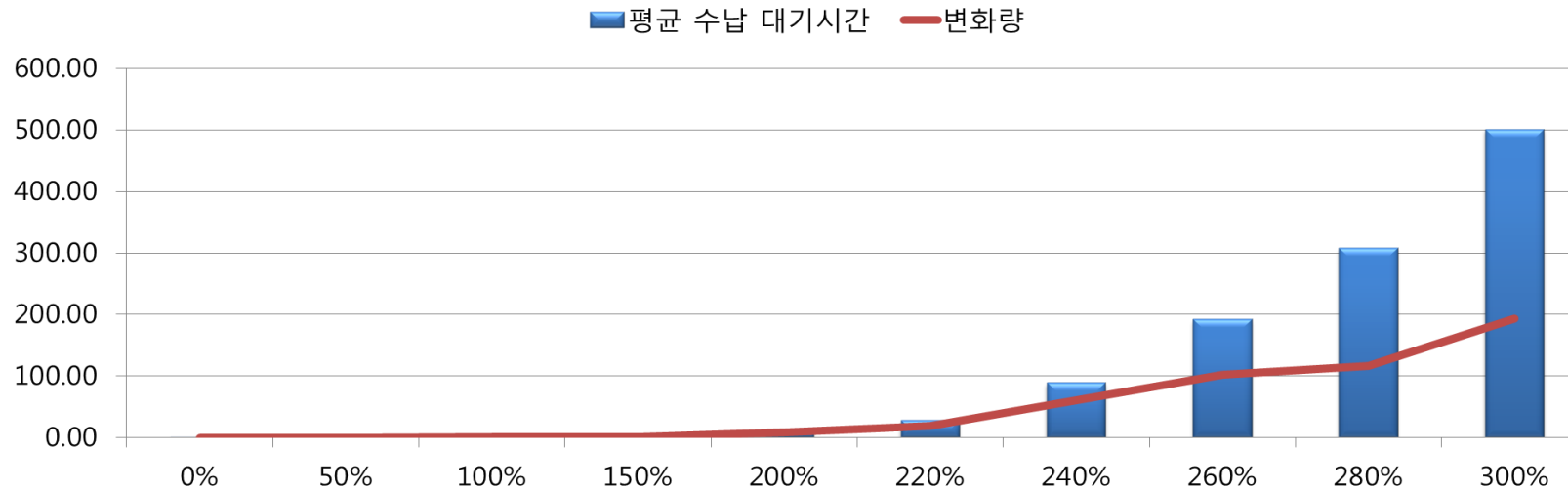
<Process Simulation Analysis Framework>

Simulation Analysis - KIOSK



Kim, E., Kim, S., Cho, M., Song, M., Yoo, S. "Simulation of optimal number of outpatient payment KIOSK estimation using process mining.", *Proceedings of the 2013 Korean Society of Medical Informatics Spring Conference*, Asan Hospital, Korea, June 13-14, 2013.

Simulation Analysis - KIOSK



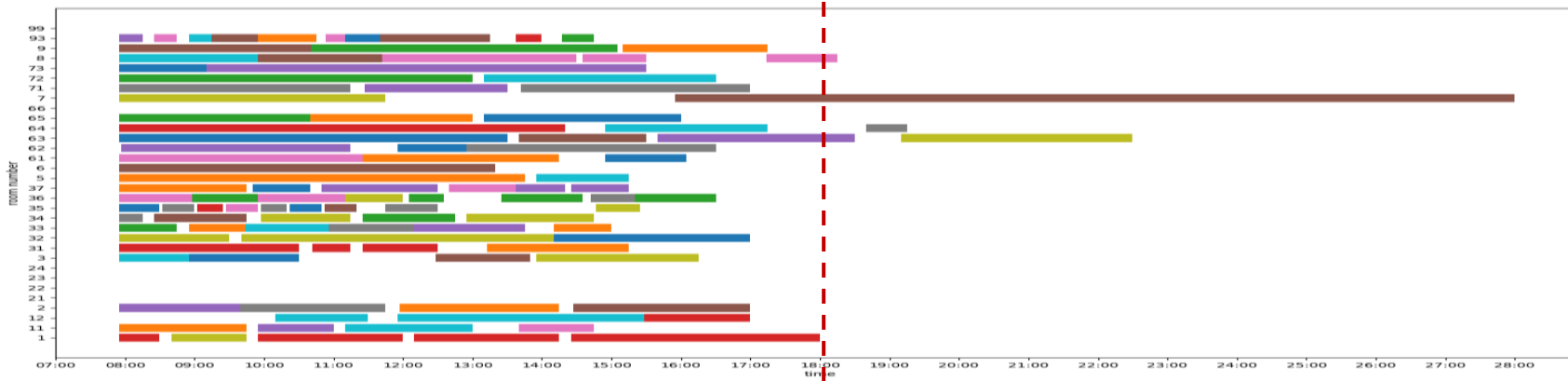
Increasing ratio	0%	50%	100%	150%	200%	220%	240%	260%	280%	300%
# payment	1265.7	1894.8	2529.3	3167.5	3801.4	4030.1	5293.4	4542.6	4801.7	5067.3
Total time	10.17	10.17	10.18	10.21	10.34	10.57	11.27	12.59	14.80	18.51
Waiting time (s)	0.00	0.03	0.24	1.46	9.40	27.98	89.01	191.70	307.89	500.61

Kim, E., Kim, S., Cho, M., Song, M., Yoo, S. "Simulation of optimal number of outpatient payment KIOSK estimation using process mining.", *Proceedings of the 2013 Korean Society of Medical Informatics Spring Conference*, Asan Hospital, Korea, June 13-14, 2013.

Process Mining + Optimization + Scheduling

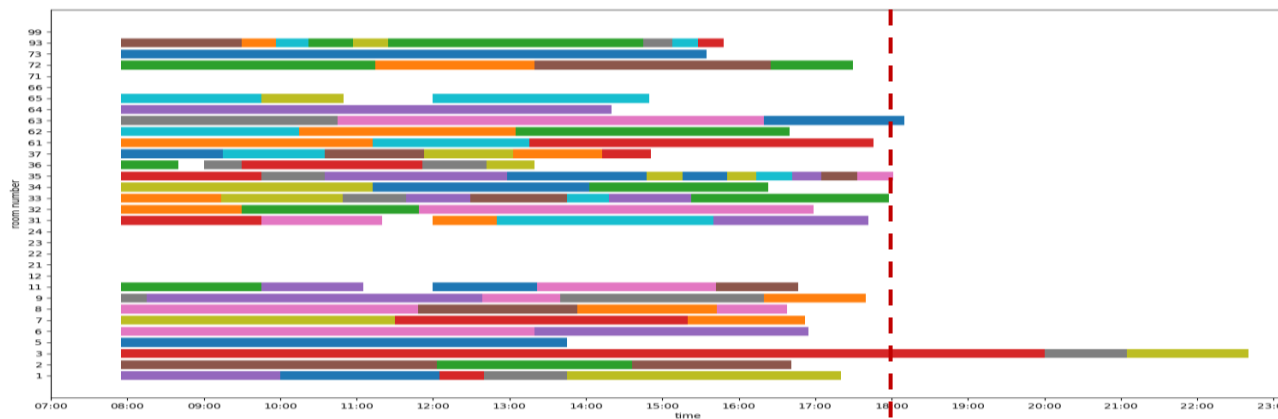
■ 실제 수술실행 현황 (2018.11.13)

- 26개의 수술실 109개 수술 할당, 오버타임 16시간 발생



■ 스케줄러에서 계획한 수술일정 (2018.11.13) – 실제 수술에 걸린 시간으로 스케줄링

- 24개의 수술실 109개 수술 할당, 오버타임 5시간 46분 발생



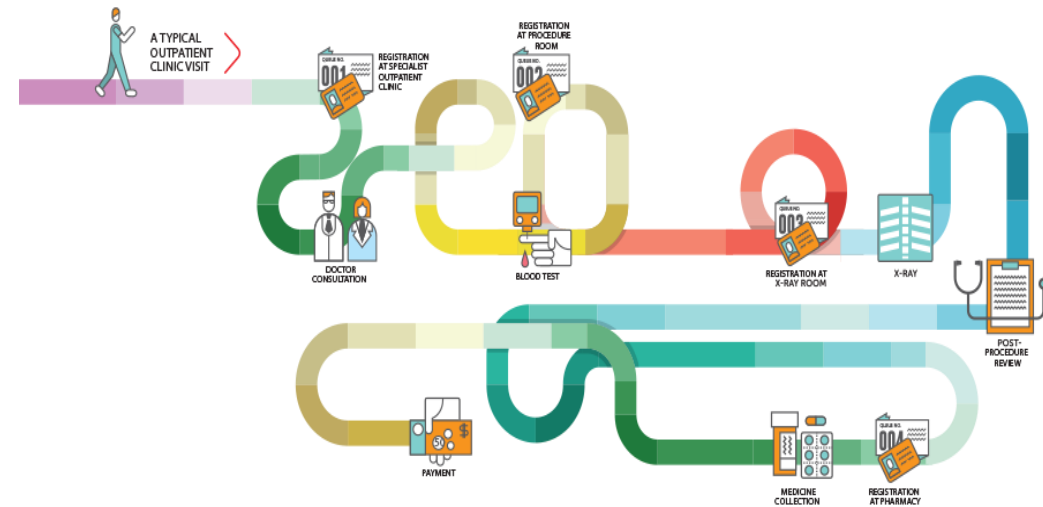
Using OPERA (OPERating room optimizing scheduler)
By 김병인 교수 (logistics.postech.ac.kr)

Clinical Pathway

CP(Clinical Pathway): 질병에 따라 정해진 표준 진료 지침

CP의 목적

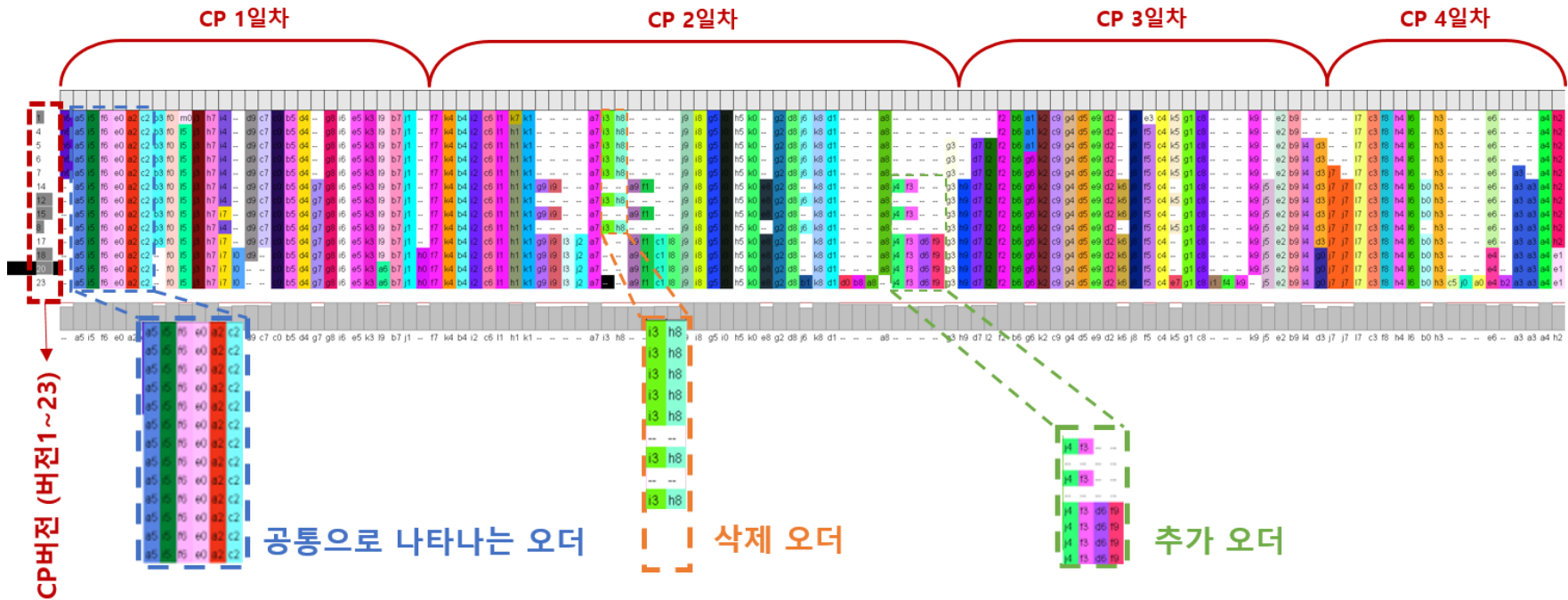
- 선진 치료 과정을 선택하여 일관적인 진료 시도
- 입원기간, 진단과정, 치료에 대한 표준을 설정
- 진료 과정 중에 가장 힘든 난관을 극복할 수 있는 방안 발견
- 모든 의료인에게 공통적인 진료계획을 제시, 자신의 역할 인지
- 의무기록 간소화
- 진료계획에 대한 교육을 통해 환자 만족



기존 CP 이해 및 평가 (Conformance) – Reference model 분석

CP Master 변화 분석

- Trace alignment
 - Trace들을 순서를 맞춰 최적으로 줄지어 배열하는 Process mining 기법
 - Trace들에서 공통으로 나타나는 이벤트와 차이가 나는 이벤트를 시각적으로 표현
- Trace alignment를 활용한 Reference model의 변화 시각화



Cho, M., Kim, K., Lim, J., Baek, H., Kim, S., Hwang, H., Song, M., Yoo, S., "Developing data-driven clinical pathways using electronic health records: The cases of total laparoscopic hysterectomy and rotator cuff tears." International Journal of Medical Informatics, Vol. 133, 104015, 2020.

기존 CP 이해 및 평가 (Conformance)

- Reference model vs. 환자 로그 비교

CP 활용도 평가: CP Master와 실제 환자의 오더 기록의 유사도 분석

- 정합률

- Fitness 계산 기법을 활용한 지표로 CP Master와 실제 환자의 오더 기록의 유사도 수치화

$$Matchrate \text{ (정합률)} = \frac{1}{2} \left(1 - \frac{M_{CP}}{N_{CP}} \right) + \frac{1}{2} \left(1 - \frac{R_{Log}}{N_{Log}} \right)$$

- N_{CP} : 표준 CP에 포함되어 있는 오더 코드 수
- M_{CP} : 표준 CP에는 포함되어 있지만, 이벤트 로그에 나타나지 않는 오더 코드 수
- N_{Log} : 처방 데이터에 포함되어 있는 오더 코드 수
- R_{Log} : 처방 데이터에는 포함되어 있지만, 표준 CP에는 포함되어 있지 않은 오더 코드

<표준 CP, 오더 로그의 매칭 예시>

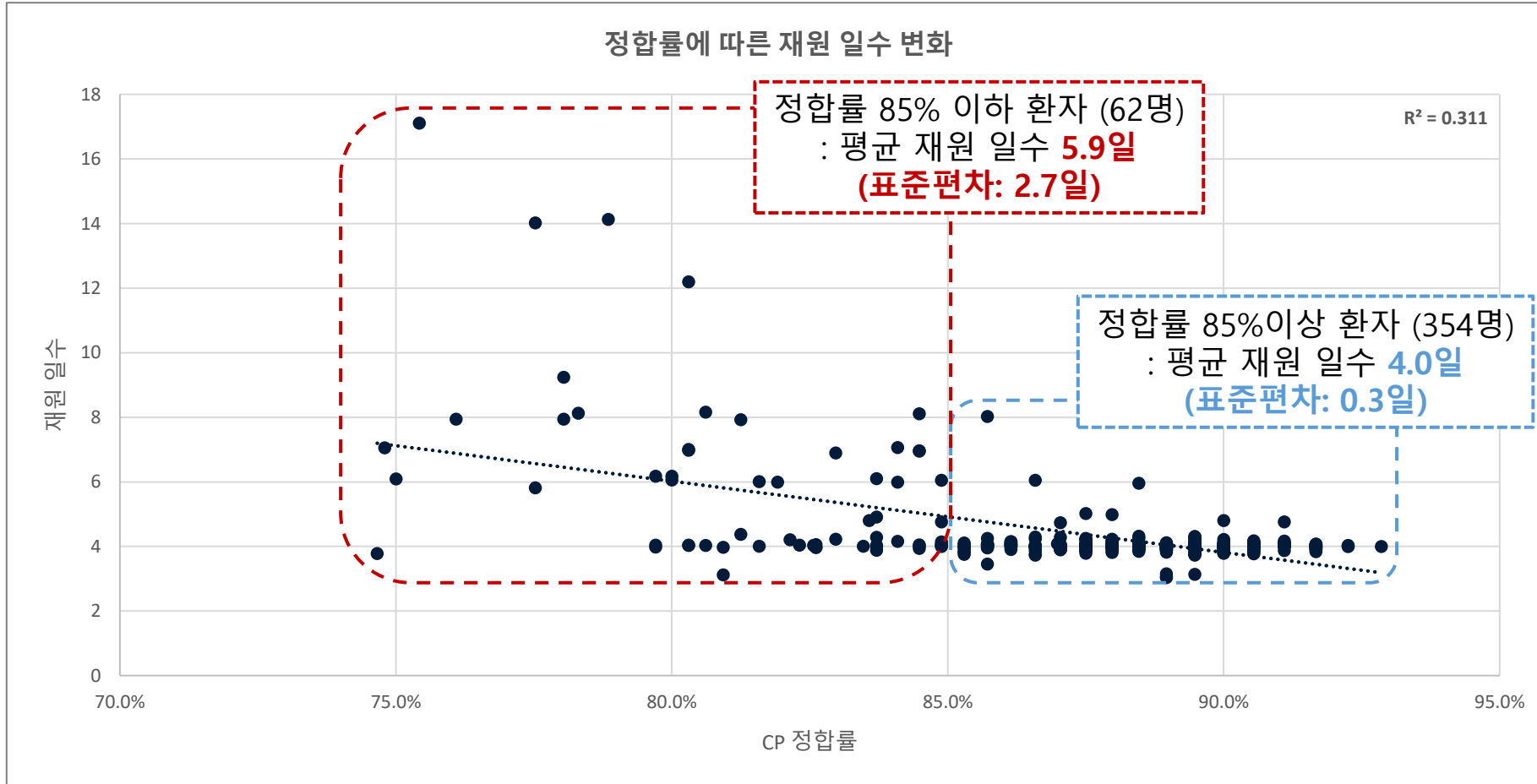
CP와 로그에 모두 나타남
 CP에만 나타남
 로그에만 나타남

작업	처치				검사			투약			식이					매칭결과				정합률
	T1	T2	T3	-	Te1	Te2	Te3	M1	M2	M3	I1	I2	I3	-	-	N_{CP}	M_{CP}	N_{Log}	R_{Log}	
표준 CP	T1	T2	T3	-	Te1	Te2	Te3	M1	M2	M3	I1	I2	I3	-	-	12	0	12	0	
환자1	T1	T2	T3	-	Te1	Te2	Te3	M1	M2	M3	I1	I2	I3	-	-	12	0	12	0	
환자2	T1	T2	T3	-	Te1	Te2	-	M1	M2	M3	I1	-	-	-	-	12	3	9	0	
환자3	T1	T2	T3	T4	Te1	Te2	Te3	M1	M2	M3	I1	I2	I3	I4	I5	12	0	15	3	
환자4	T1	-	T3	T4	Te1	Te2	-	M1	-	M3	I1	I2	-	I4	-	12	4	10	2	

기존 CP 이해 및 평가 (Conformance)

- CP 효과 평가

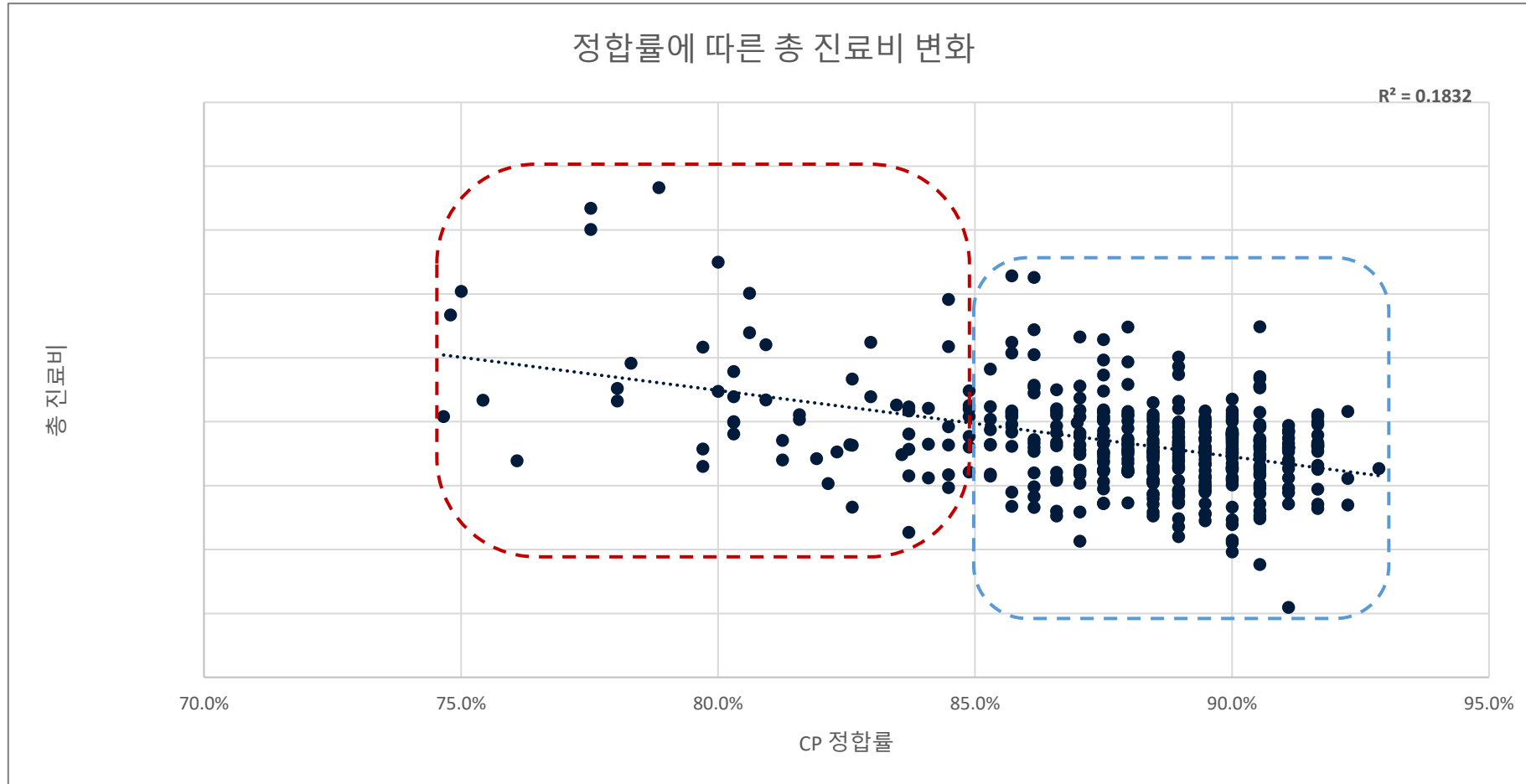
CP와 자원일수의 연관성 분석



기존 CP 이해 및 평가 (Conformance)

- CP 효과 평가

CP와 진료비의 연관성 분석

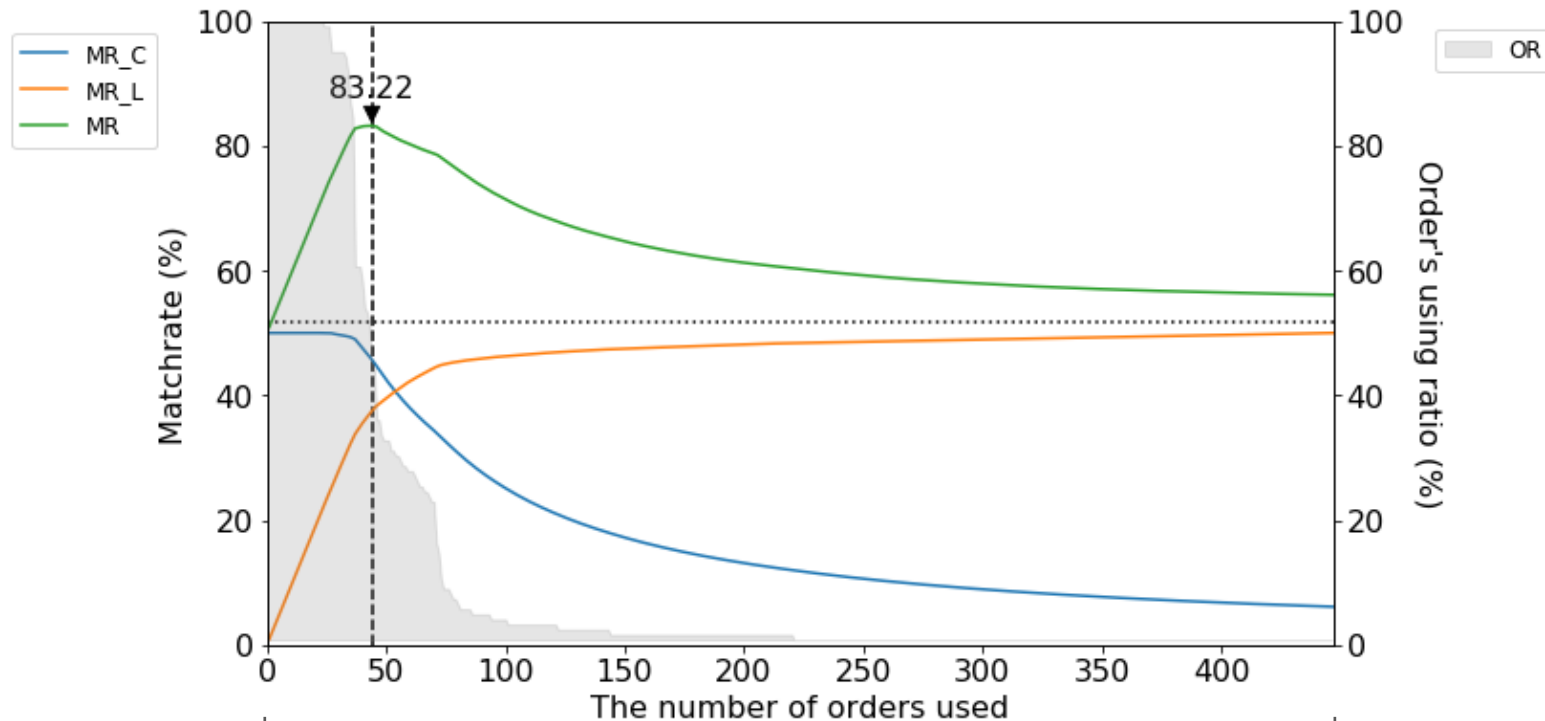


데이터 기반 CP 개발 (Discovery)

- 데이터 기반 CP 개발 방법론

오더 실적용률 & 정합률 기반 CP 개발

- 실적용률이 51.64% 이상인 오더들로 CP Master를 구성할 때 환자들의 오더 로그와 CP Master의 유사성이 가장 높음



수술 1일 전		수술일		수술 1일 후		수술 2일 후		수술 3일 후	
Original	New	Original	New	Original	New	Original	New	Original	New
D11AX	D11AX	A02BA03	A02BA03	A02BA03	A02BA03	A02BA03	A02BA03	A04AA01	A04AA01
M01AH01	M01AH01	A04AA	A04AA	A04AA	A04AA	A04AA01	A04AA01	B05XA03	B05XA03
N01BB02	N01BB02	A04AA01	A04AA01	A04AA01	A04AA01	B05XA03	B05XA03	L2017	L2017
N02AA05	N02AA05	B05BB01	B05BB01	B05XA03	B05XA03	N02AA01	N02AA01	L3001	L3001
N02BE01	N02BE01	B05XA03	B05XA03	N02AA01	N02AA01	R05CB02	R05CB02	L3092	L3092
RG9045	RG9045	J01DB04	J01DB04	R05CB02	R05CB02	J01DB04	J01DB04	N02AA01	N02AA01
		N02AA01	N02AA01	J01DB04	J01DB04	L2017	L2017	L3046	L3046
		R05CB02	R05CB02	L2017	L2017	L2201	L2201		J01DB01
		B05BB02	B05BB02	L2201	L2201	A02BX		RG6061R_pre	RG6061R
		M01AH01		A02BX		L3001		RG6062R_pre	RG6062R
		N02AB03		D03AX		L3046		RG6063R_pre	RG6063R
		N02AX02		L3001		L3092		A02BX	A02BX
		B05XA03_pre		L3046		M01AH01		M01AH01	
				L3092		N02AB03		M03BX	
				M01AH01		N02AX02		N02AX02	
				N02AB03		N02AXS2		N02AXS2	
				N02AX02		RG6061L_pre		RG6061L_pre	
				N02AXS2		RG6061R_pre		RG6062L_pre	
				RG6061L_pre		RG6062L_pre		RG6063L_pre	
				RG6061R_pre		RG6062R_pre		A02BA02_pre	
				RG6062L_pre					
				RG6062R_pre					
				RG6063L_pre					
				RG6063R_pre					

- Original & New CP Master의 공통 오더
- CP Master에 포함될 필요가 없는 오더
- 기존 CP Master에서 제거되어야 할 오더
- 기존 CP Master에 추가되어야 할 오더

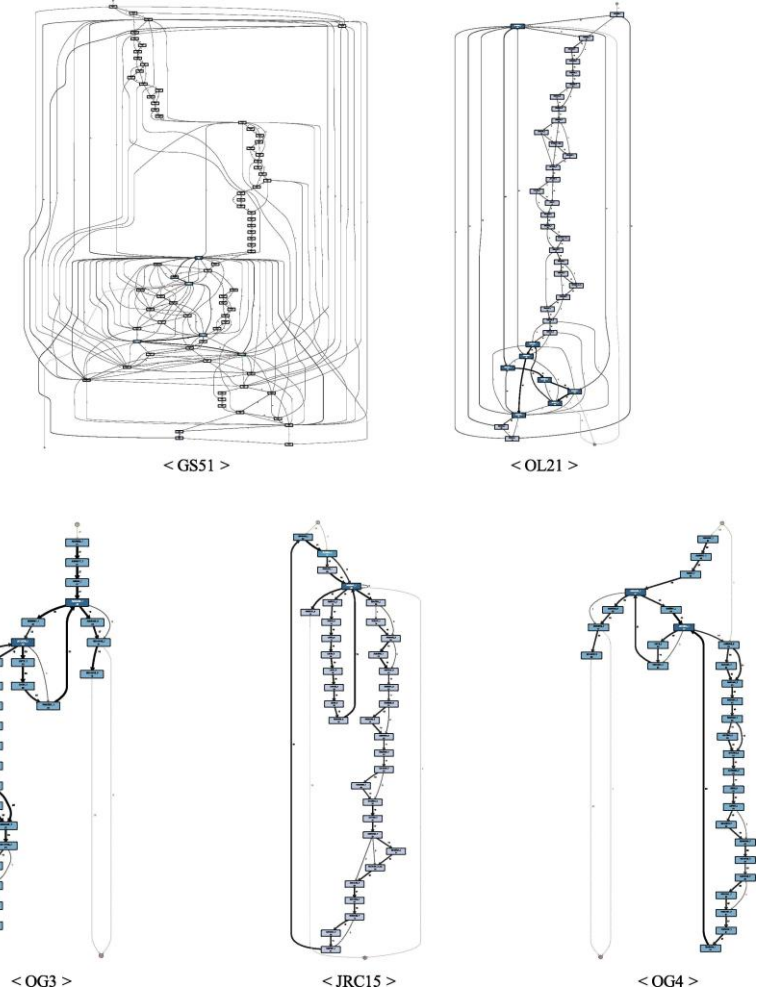
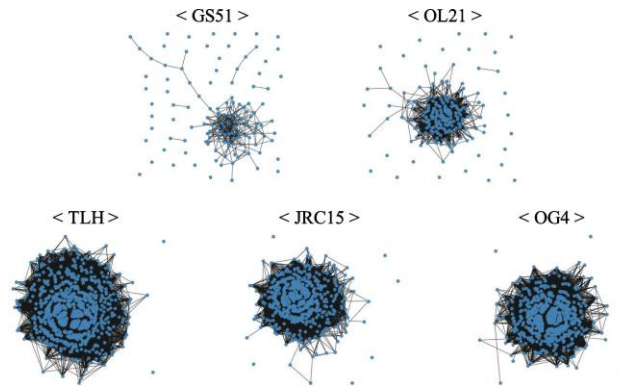
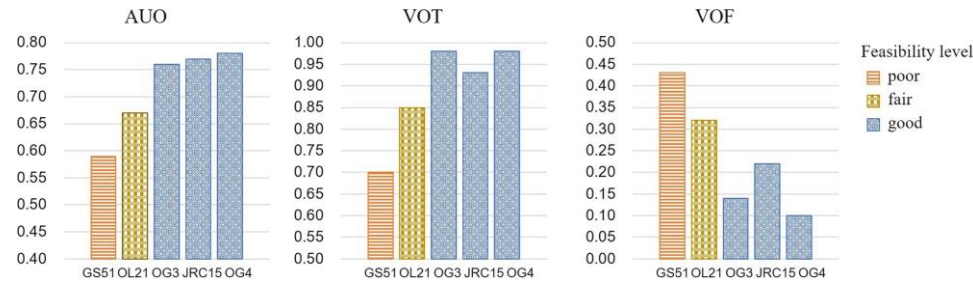
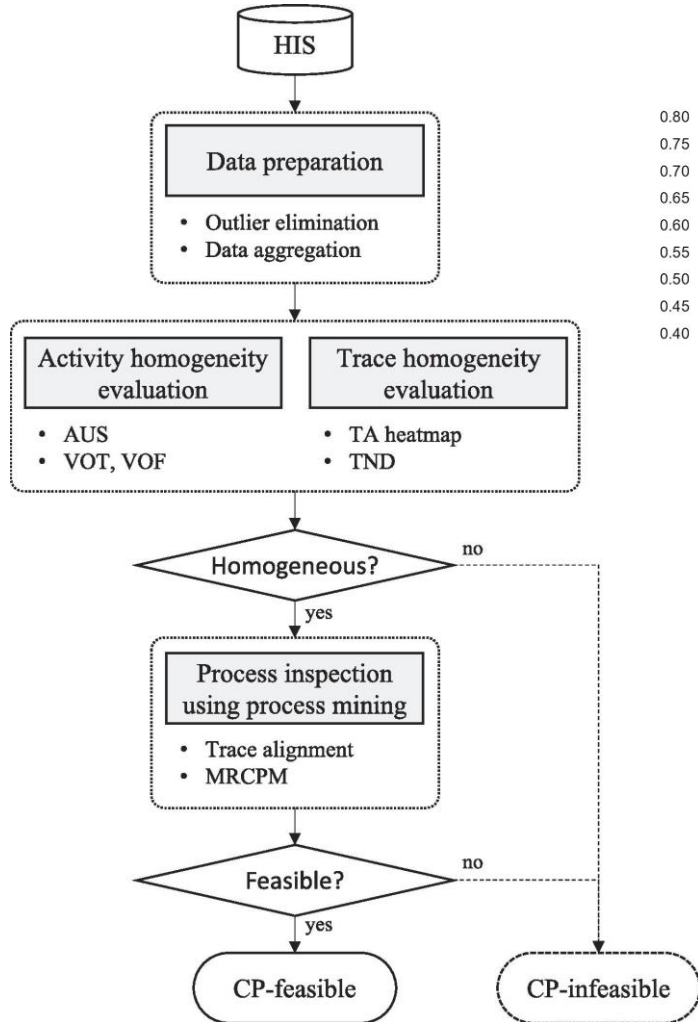
실적용률이 가장 높은 오더만 CP Master에 포함

CP Master에 실적용률이 높은 오더 하나씩 추가

CP Master에 모든 오더 포함

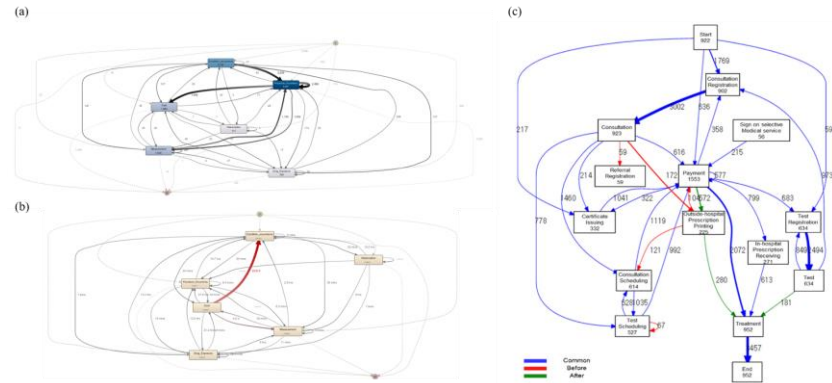
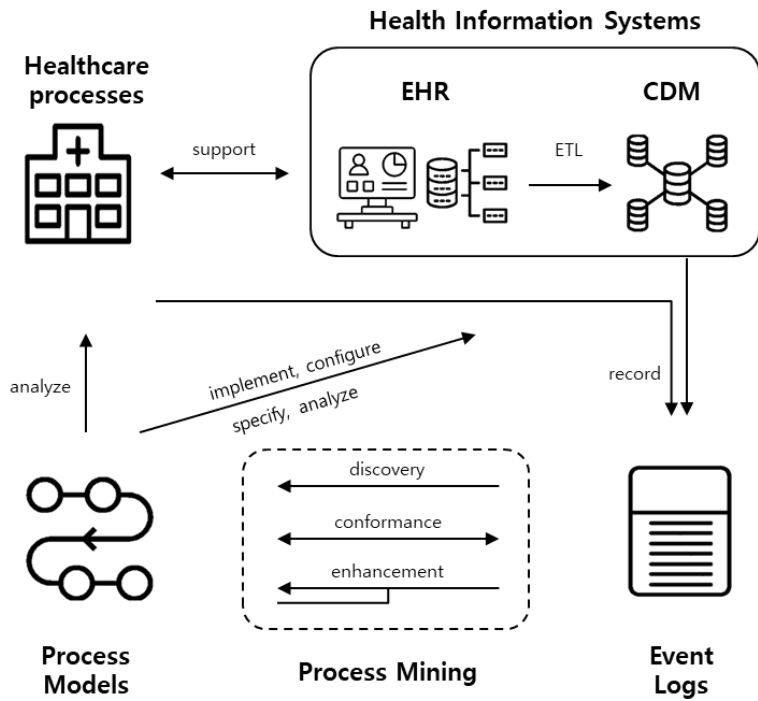
데이터 기반 CP 개발 (Discovery)

- 표준화 가능성 평가

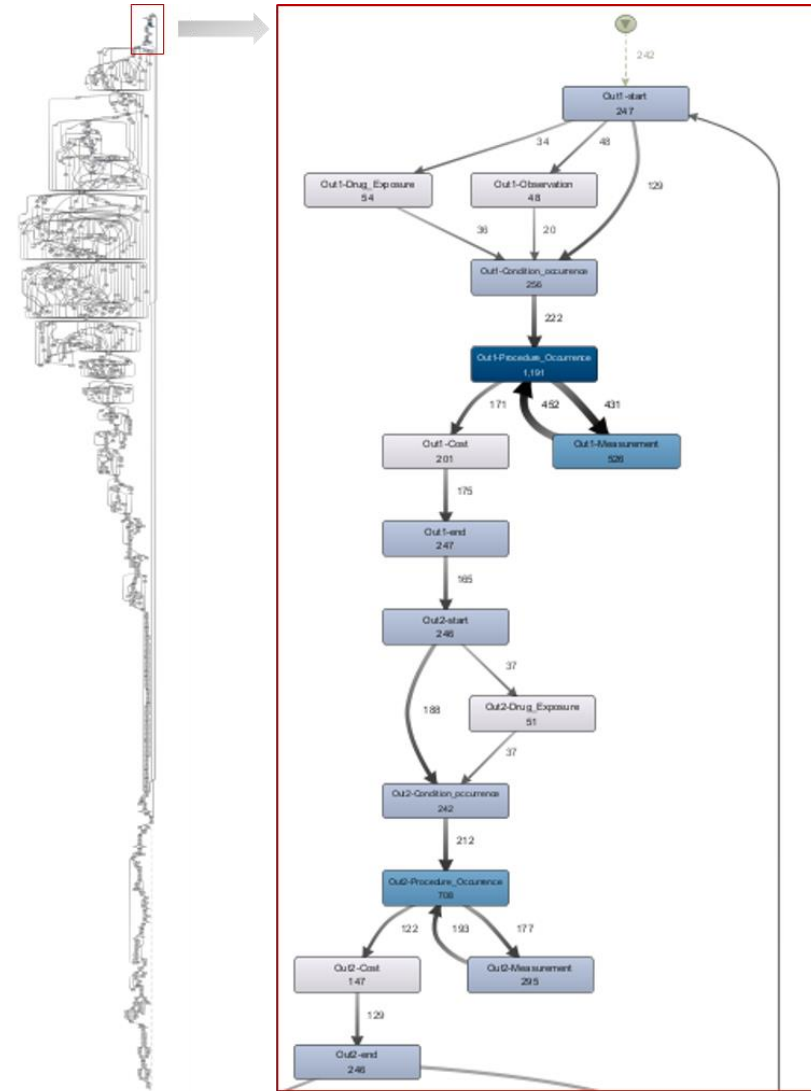


Lim, J., Kim, K., Song, M., Yoo, S., Baek, H., Kim, S., Park, S., Jeong, W., "Assessment of the feasibility of developing a clinical pathway using a clinical order log", Journal of Biomedical Informatics, Vol. 128, 104038, 2022.

OMOP CDM + Process Mining



Outpatient Process Analysis

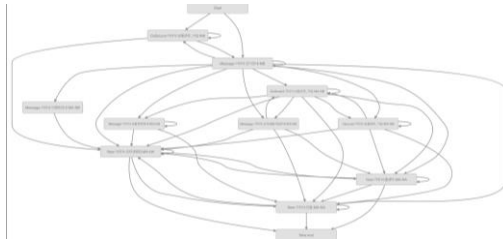


Patient journey

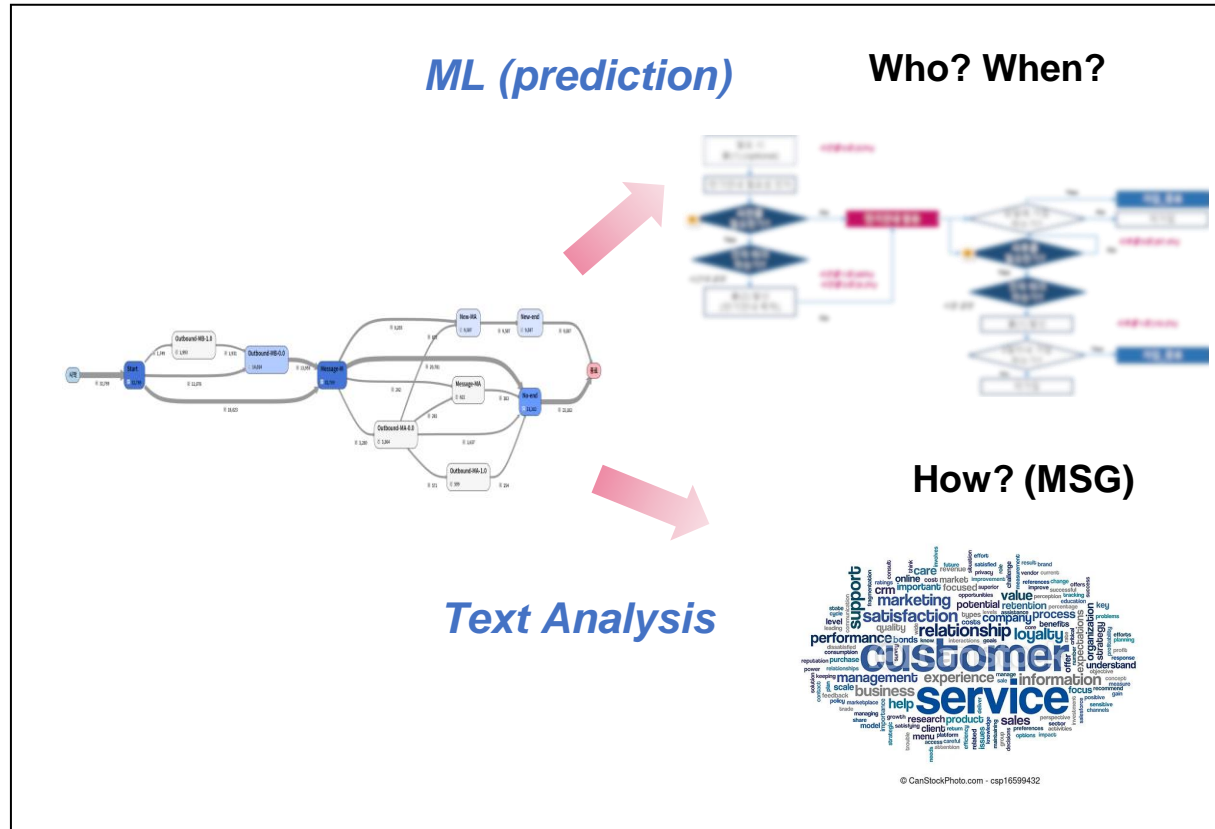
Finance Case: CJ Analysis

Cases	Logs
1	(A,John),(B,Sue),(C,John),(D,Carol)
2	(A,John),(C,Mike),(B,John),(D,Sue)
3	(A,Carol),(E,Mike),(D,Sue)
4	(A,Pete),(C,Carol),(B,Clare),(D,Pete)
5	(A,John),(E,Carol),(D,Clare)

Process Mining



AS-IS CJ



ML (prediction)

Who? When?

Text Analysis

How? (MSG)

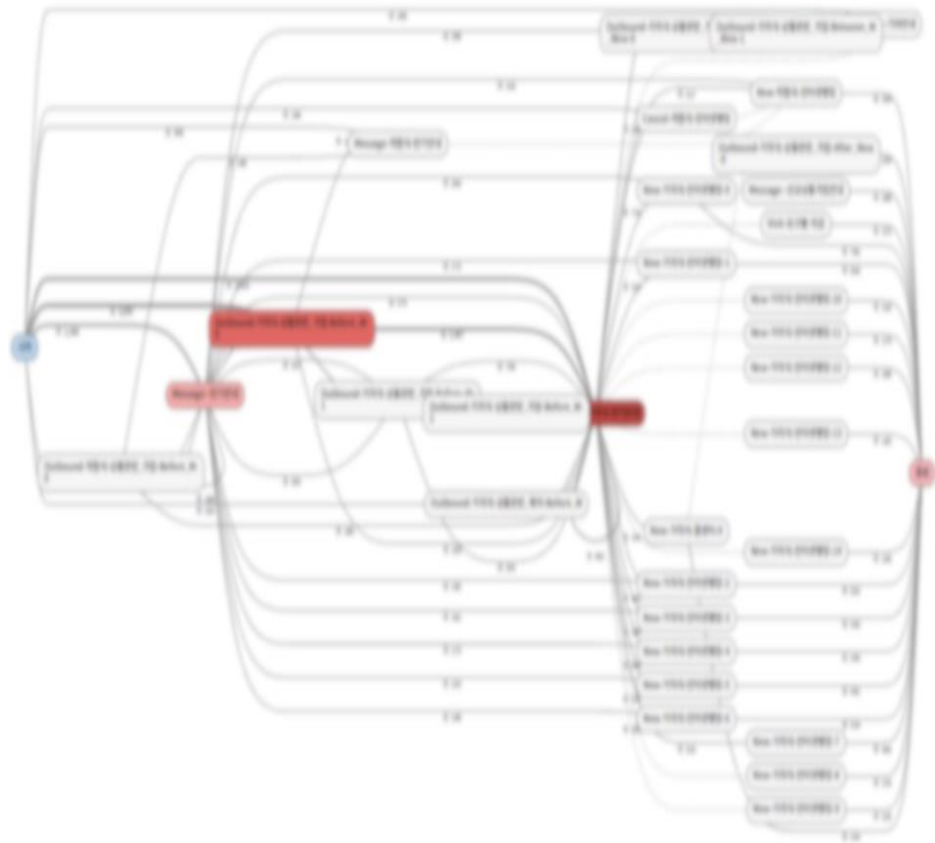


TO-BE CJ

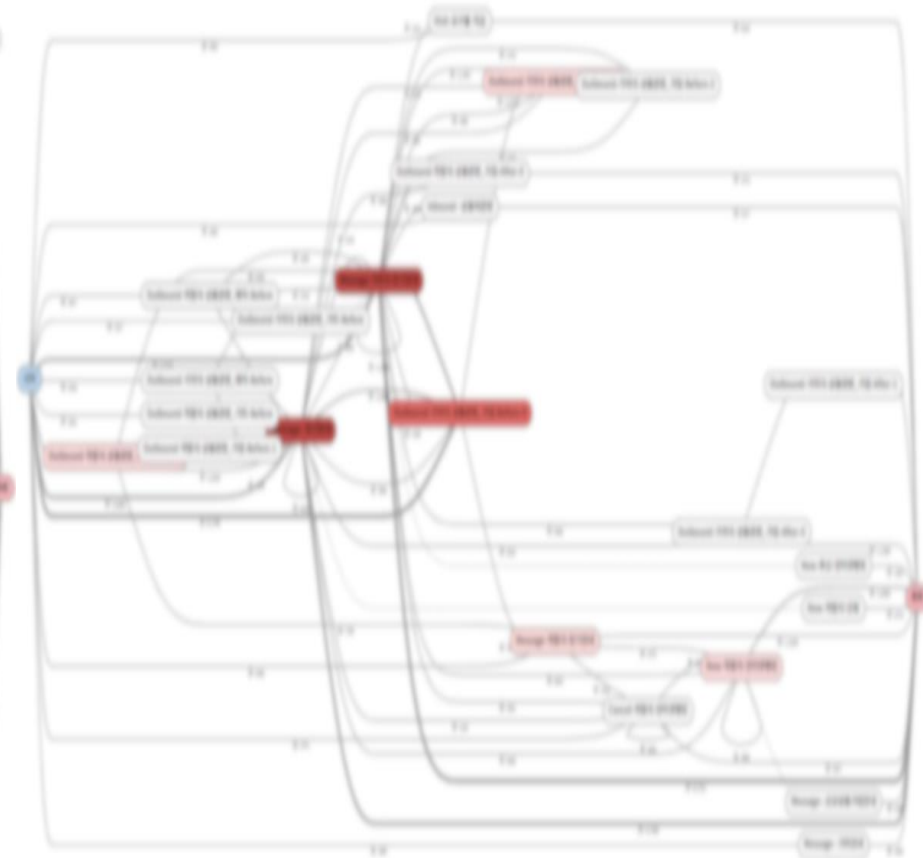
As-is CJ 도출

가입과 미가입 케이스의 CJ가 다름을 확인

가입 Process



미가입 Process



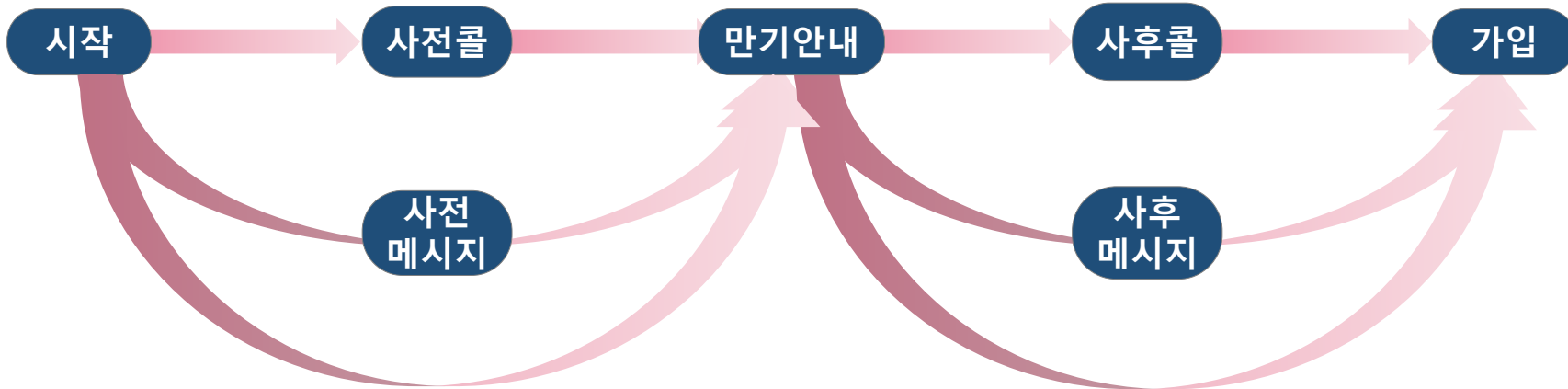
To-be CJ 도출

만기안내 전 사전콜/사전메시지, 만기안내 후 사후콜/사후메시지

만기안내가
필요한 고객



To-be CJ map



누구에게 언제 사전콜/사후콜을 해야 하는가? → AI 예측
사전/사후 메시지의 내용은 무엇인가? → Text Analysis

AI 기반 예측 모델링: Who? When?

EDA (Exploratory Data Analysis)

- 세션-사후 콜 4회가 가입을 성공에 효과 있는 profile이 많이됨

이전콜 1회이상 2회이상	이전콜 3회이상 4회이상	Profile
1. 세션-사후 콜	1. 세션-사후 콜	<ul style="list-style-type: none"> • 세션콜은 4회이상 효과 • 사후콜은 4회이상 효과 • 2달에는 사후콜 완료 • 핵심은 세션-사후 콜 4회
2. 세션-사후 콜	2. 세션-사후 콜	
3. 세션-사후 콜	3. 세션-사후 콜	
4. 세션-사후 콜	4. 세션-사후 콜	

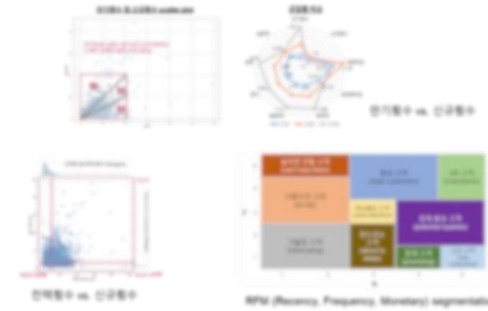
Who? 콜 효과가 큰 profile

- 오전 가입을 늘린 profile: MVP-2017년 1월 10일~10월 31일, 1월 1~31일
- 오후 가입을 늘린 profile: MVP-4075-1년 1월, 1월 1~31일

Demographic, Geographic, 가입을 도우려는, 가입이 성공한, 가입이 실패한

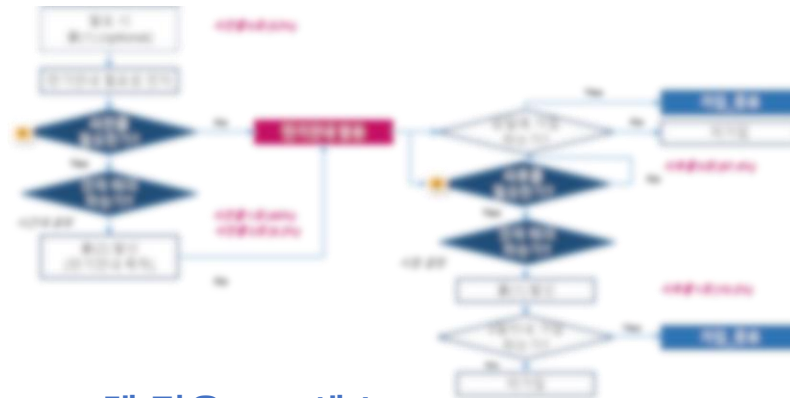
	성별	연령	지역	가입 성공	가입 실패
성별	남성	여성	남성	여성	남성
연령	10-19	20-29	30-39	40-49	50-59
지역	서울	경기	충청	전라	제주

When? 콜 시간대

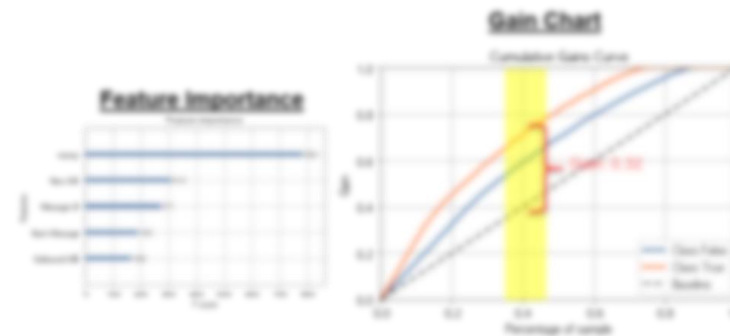


다양한 고객 군집 정보 활용

AI 기반 예측 모델링



모델 적용 프로세스



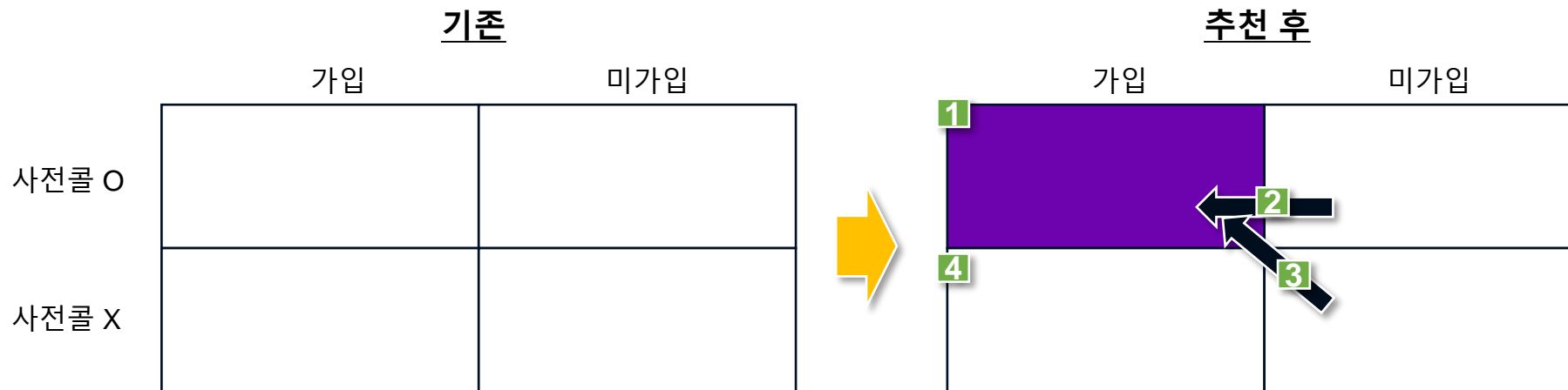
XGboost 기반 모델링

AI 기반 예측 모델링: 기대 효과

가입률 증가: 2배

사전콜 수행비중 증가: 2배

최빈시간대 변경: 월요일 → 금요일



- 1 사전콜 O → O, 가입 유지: 가입확률을 더 높이는 시간대 추천
- 2 사전콜 O → O, 미가입 → 가입: 가입확률을 높이는 시간대로 변경 추천
- 3 사전콜 X → O, 미가입 → 가입: 콜 수행 추천, 시간대는 새로 추천
- 4 사전콜 X → X, 가입 → 가입: no action (현상유지)

우수 직원 vs. 일반 직원

신규 실적이 많은 직원은 고객 向, 일반 직원은 은행 向 키워드가 많은 경향



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Conclusion: Process Management

- **Process Discovery: PM (discovery)**
- **Process Monitoring:**
PM (discovery + conformance checking) + Real Time/Big DATA
- **Process Prediction: PM + ML + AI**
- **Process Optimization: PM + OR**
- **Process Improvement: PM + AI + SIMULATION**

Conclusion: Digital Twin

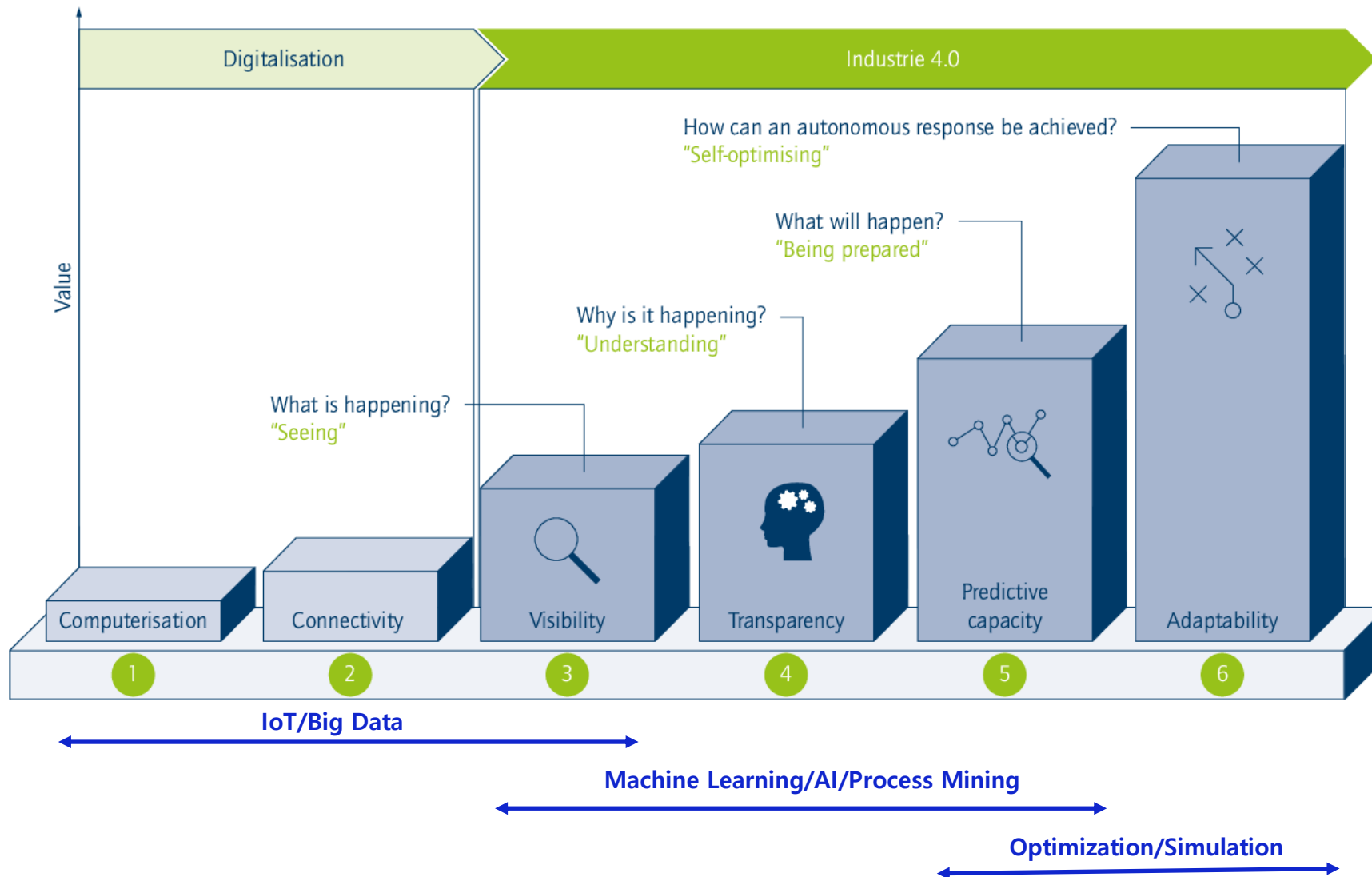
Data Driven Rapid Model Generation

- Process Model* Discovery: PM
- Rules/Prediction Model Discovery: PM + ML + AI

DT Monitoring & Improvement

- Process Monitoring: PM (Discovery + Conformance)
- Process Optimization: PM + OR
- Process Improvement: PM + AI + SIMULATION

Conclusion: Stages in the Industrie 4.0 development path



Agenda

Keynote

Prof. van der Aalst: Object-Centric Process Mining: Moving from 2D to 3D process analytics

제조

삼성전자: 반도체 FAB 공정의 프로세스 마이닝 적용 사례

LG CNS: 제조업에서의 프로세스 마이닝을 통한 제조 업무 프로세스 혁신 및 개선 모델 탐색

KAI/IMP: Digital 시대의 경영, Data를 활용하여 기업행동을 재설계하라

서비스

신한은행: 프로세스 마이닝 기법을 활용한 고객 경험 개선 사례

한림대학교 성심병원: 병원은 왜 프로세스 마이닝을 필요로 하는가?

SK텔레콤: 고객접점 MOT로부터의 프로세스 및 IT자원 최적화

솔루션

퍼즐데이터: 프로세스 마이닝으로 하는 모니터링, 분석, 최적화, 검증! 프로세스 인텔리전스 플랫폼

Thank you

프로세스 마이닝 창시자 윌 반 데르 알스트(Wil van der Aalst) 교수 초청
'프로세스 마이닝 워크샵'

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