#### School of Computing FACULTY OF ENGINEERING



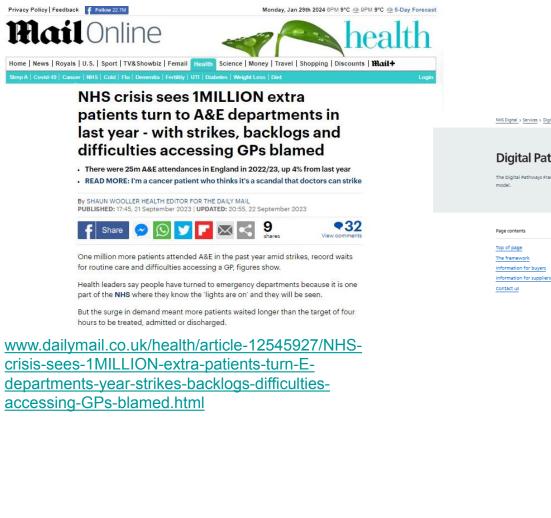
## Process Mining to Identify Pathway Improvement Opportunities

Leeds Digital Health 6.15 pm 31<sup>st</sup> January 2024

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### Healthcare in crisis ... are digital pathways the answer?

# **UNIVERSITY OF LEEDS**



NHS Digital > Services > Digital Services for Integrated Care > Digital Pathways Framework

#### **Digital Pathways Framework**

The Digital Pathways Framework will be the first framework to launch under the new Digital Services for Integrated Care

It is part of a new suite of frameworks designed to support modern general practice by providing standardised assured, interoperable digital systems for primary care settings

Solutions on the Digital Pathways Framework will be available to procure from the Buying Catalogue in early 2024.

It builds upon the Digital First Online Consultation Video Consultation (DFOCVC) Framework, which expires in March 2024.

#### The framework

The new Digital Pathways Framework will include new solutions to support patient pathways and assess patient needs. This includes initial online contact with a GP practice, navigation to the appropriate point of care, messaging and enabling patient interactions with the practice, and scheduling or booking appointment:

#### Products in the Digital Pathways Framework will cover these core Capabilities

- · online consultations and administrative request reporting online patient or service user consultation
- care navigation
- online administrative requests

#### In addition to at least one of the above, products may also provide:

- prescription ordering (for patients)
- communication management video consultation
- record viewing (for patients)
- cross organisational appointment booking

#### Information for buyers

The framework overview pack provides an overview of the framework and the digital tools that will enable modern general practice. One of the main commitments in the recovery plan

It also shows the main actions for ICBs to prepare for the framework launching in early 2024. The information in this pack should be read in conjunction with the system letter and updated checklist of actions, published on 13 September

https://digital.nhs.uk/services/digital-services-for-integratedcare/digital-pathways-framework



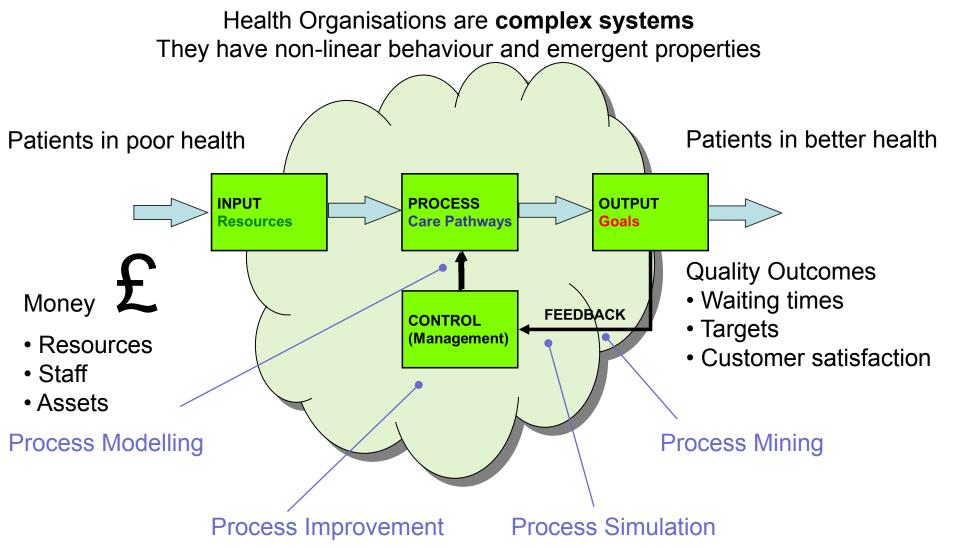
Digital technology is a significant part of our everyday lives improving the way we socialise, shop and work. It also has great potential to improve how the NHS delivers its services in a new and modern way; providing faster, safer and more convenient care.

Our NHS Long Term Plan will increase the range of digital health tools and services. People will be able to seek health information and support online, and choose whether they speak to a doctor on the phone or in person. A wide range of NHS-approved apps will help people get ongoing support to help them manage their health and wellbeing needs, backed up by face-to-face care when this is needed.

We are also investing in improving NHS IT systems and in developing new technology. We will make sure staff have the technology they need to do their jobs, and our systems can talk to each other and share vital information to support the delivery of care.

### The challenge The NHS is a "Complex System"

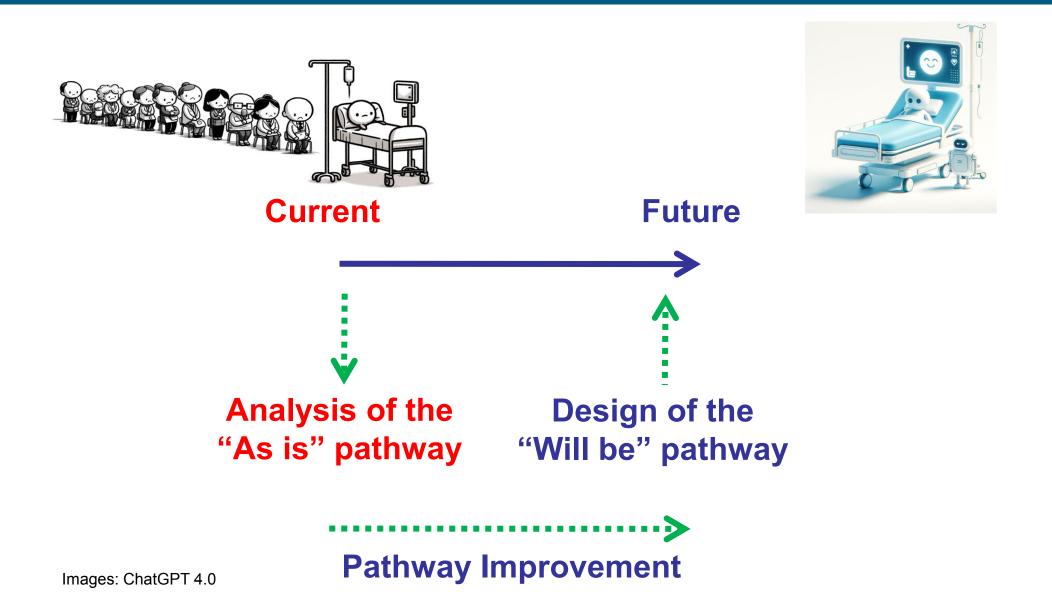




Book chapter: Johnson, O., 2019. General System Theory and the Use of Process Mining to Improve Care Pathways. *Studies in health technology and informatics*, 263, pp.11-22.



### Designing new Digital Pathways



### Care Pathways In design and in reality

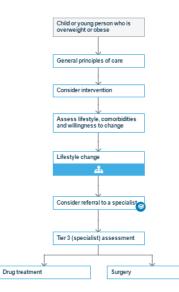


### Typical "As is" pathway



Care pathways are informed by clinical training and working practices within healthcare providers. Care pathways are implemented, mediated and recorded by health information systems.

#### Typical "Should be" pathway

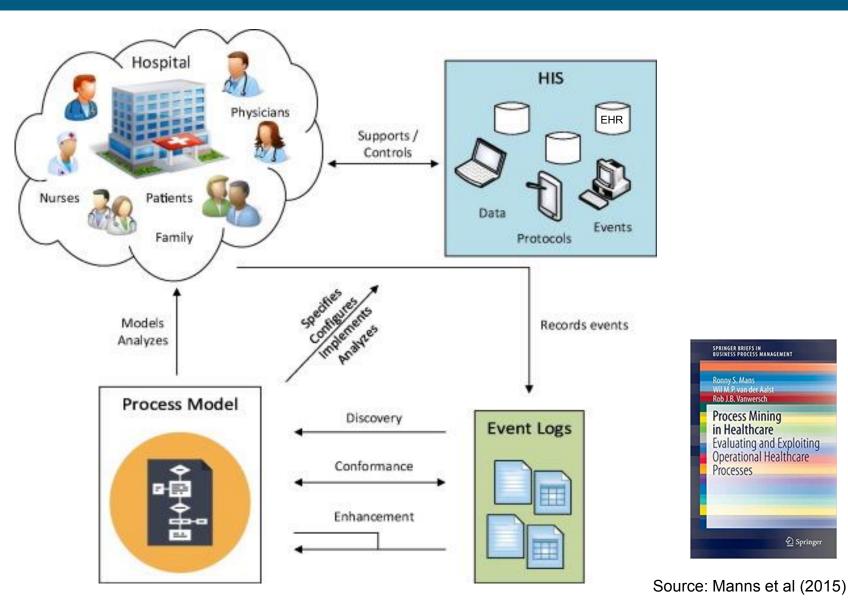


Example: Managing children and young people who are overweight or obese pathways.nice.org.uk

NB NICE withdrew their comprehensive set of care pathway guidelines in 2022



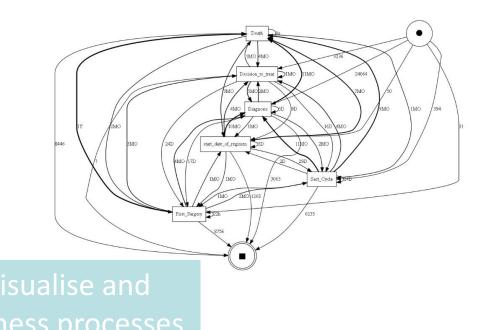
#### Healthcare processes

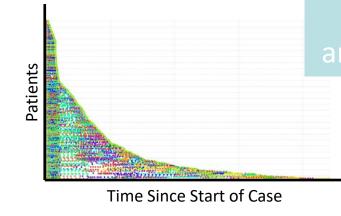


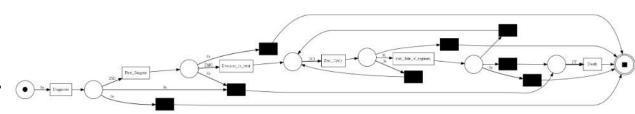


### What is Process Mining?

	Patient	Activity	Resource	Timestamp
0	patient 0	First consult	Dr. Anna	2017-01-02 11:40:11
1	patient 0	Blood test	Lab	2017-01-02 12:47:33
2	patient 0	Physical test	Nurse Jesse	2017-01-02 12:53:50
3	patient 0	Second consult	Dr. Anna	2017-01-02 16:21:06
4	patient 0	Surgery	Dr. Charlie	2017-01-05 13:23:09
5	patient 0	Final consult	Dr. Ben	2017-01-09 08:29:28

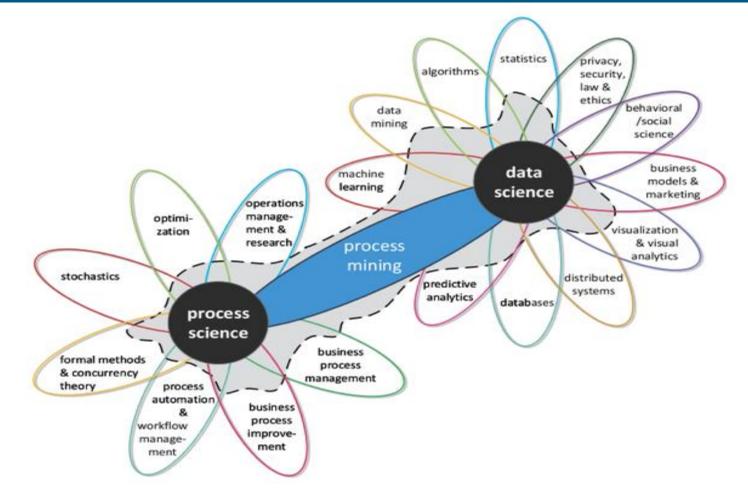








### Process Mining Research Field



Process mining is bridging the gap between classical process model analysis and data science analysis. Process mining focuses on understanding real business processes using real data. In classical data mining people usually ignore the process.

#### The Process Mining for Healthcare Manifesto



Based on a two-day brainstorming event in Hasselt, Belgium (July 2019) Identified ten Characteristics of healthcare that make Process Mining in health different And ten Challenges for future research...



#### Original Research

Process mining for healthcare: Characteristics and challenges

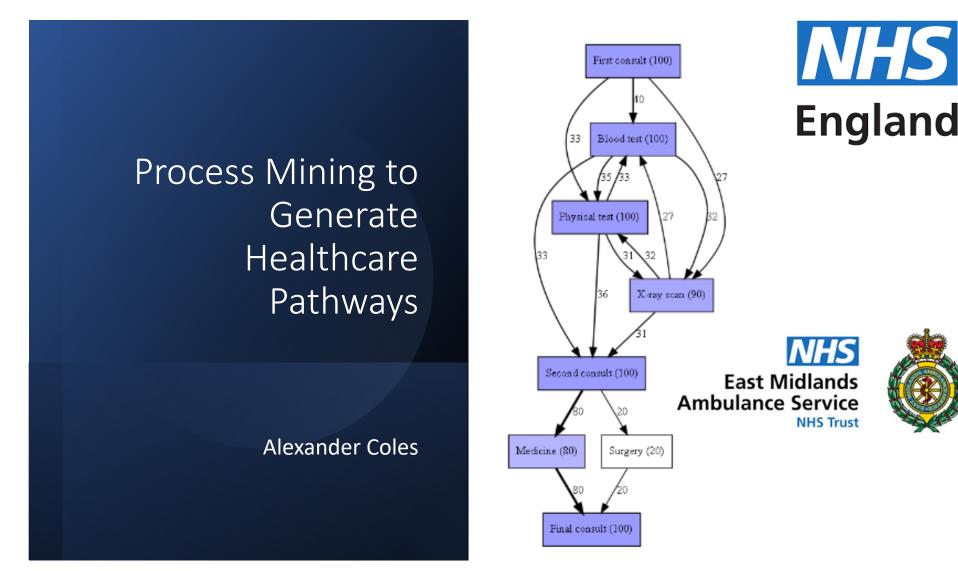
Jorge Munoz-Gama<sup>a,\*</sup>, Niels Martin<sup>b,c,\*</sup>, Carlos Fernandez-Llatas<sup>d,g,\*</sup>, Owen A. Johnson<sup>e,\*</sup>, Marcos Sepúlveda a,\*, Emmanuel Helm<sup>f,\*</sup>, Victor Galvez-Yanjari a,\*, Eric Rojas a, Antonio Martinez-Millana<sup>d</sup>, Davide Aloini<sup>k</sup>, Ilaria Angela Amantea<sup>1,q,r</sup>, Robert Andrews<sup>ab</sup>, Michael Arias<sup>z</sup>, Iris Beerepoot<sup>o</sup>, Elisabetta Benevento<sup>k</sup>, Andrea Burattin<sup>ai</sup>, Daniel Capurro<sup>j</sup>, Josep Carmona<sup>s</sup>, Marco Comuzzi<sup>w</sup>, Benjamin Dalmas<sup>aj, ak</sup>, Rene de la Fuente<sup>a</sup>, Chiara Di Francescomarino<sup>h</sup>, Claudio Di Ciccio<sup>i</sup>, Roberto Gatta<sup>ad,ae</sup>, Chiara Ghidini<sup>h</sup>, Fernanda Gonzalez-Lopez<sup>a</sup>, Gema Ibanez-Sanchez<sup>d</sup>, Hilda B. Klasky<sup>F</sup> Angelina Prima Kurniati<sup>al</sup>, Xixi Lu<sup>o</sup>, Felix Mannhardt<sup>m</sup>, Ronny Mans<sup>af</sup>, Mar Marcos<sup>v</sup>, Renata Medeiros de Carvalho<sup>m</sup>, Marco Pegoraro<sup>x</sup>, Simon K. Poon<sup>ag</sup>, Luise Pufahl<sup>u</sup>, Hajo A. Reijers<sup>m,o</sup>, Simon Remy<sup>y</sup>, Stefanie Rinderle-Ma<sup>ah</sup>, Lucia Sacchi<sup>t</sup>, Fernando Seoane<sup>g, am, an</sup>, Minseok Song<sup>aa</sup>, Alessandro Stefanini<sup>k</sup>, Emilio Sulis<sup>1</sup>, Arthur H. M. ter Hofstede ab, Pieter J. Toussaint ac, Vicente Traver d, Zoe Valero-Ramon d, Inge van de Weerd<sup>°</sup>, Wil M.P. van der Aalst<sup>x</sup>, Rob Vanwersch<sup>m</sup>, Mathias Weske<sup>y</sup>, Moe Thandar Wynn<sup>ab</sup>, Francesca Zerbato<sup>r</sup>

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#### Process Mining for Healthcare: Characteristics and Challenges

Process Mining for Healthcare (PM4H)	Characteristics (D)	D1. Exhibit Sobstantial Variability D2. Value the infrequent Behaviour D3. Use Guidelines and Protocols D4. Break the Glass D5. Consider Data at Nultiple Abstraction Levels D6. Involve a Multiplexiplinary Team	D7. Focus on the Patient D8. Think about White-box Approaches D9. Generate Sensitive and Low-Quality data D10. Handle Rapid Evolutions and New Panadigms
	a wide range of experts Challenges IC	C1. Design Dedicated/Tailored Methodologies and Framoworks C2. Discover Beyond Discovery C3. Mind the Concept Drift C4. Deal with Reality C5. Do it Yourself (DIY) C6. Pay Attention to Data Quality	<ul> <li>C7. Take Care of Privacy and Security</li> <li>C8. Look at the Process through the Patient's Eyes.</li> <li>C9. Complement HBs with the Process Perspective.</li> <li>C10. Evolve in Symbiosis with the Developments in the Healthcare Domain</li> </ul>

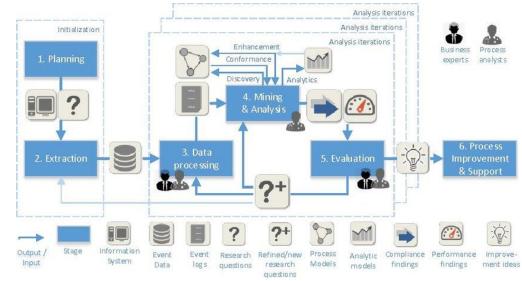
Munoz-Gama, J., Martin, N., Fernandez-Llatas, C., Johnson, O.A., Sepúlveda, M., Helm, E., Galvez-Yanjari, V., Rojas, E., Martinez-Millana, A., Aloini, D. and Amantea, I.A., 2022. Process mining for healthcare: Characteristics and challenges. Journal of Biomedical Informatics, p.103994







#### EMAS Progress/Targets



PM2 Process Mining Methodology. Van Eck, M. L., Lu, X., Leemans, S. J. & Van Der Aalst, W. M. (2015), Pm: a process mining project methodology, in 'International conference on advance information systems engineering', Springer, pp. 297–313



Job Cyle from Public Services Committee (2023), 'Emergency healthcare: a national emergency.', https://committees. parliament.uk/publications/33569/documents/187215/default/[Accessed 25 September 2023

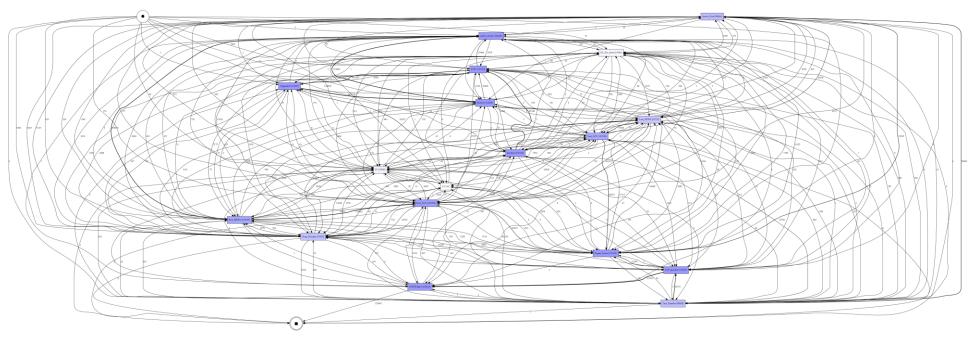
#### <u>Focus</u>

#### Process Discovery

- Category 2 makes up the majority of Ambulance Calls.
- Time to Scene, Time on Scene and Hospital Handover are of great interest to EMAS.
- How do attributes effect these job cycles (clinical category, Highest Qualification on scene, IMD etc)?

### All Category 2

Every Trace Variant for Category 2 Data



#### **Business Rules**

- Some timestamps are not automatically recorded, allowing some human error in recording.
- We constructed business rules with our clinical experts to ensure that we focus on cases where key events are recorded in the expected chronological order

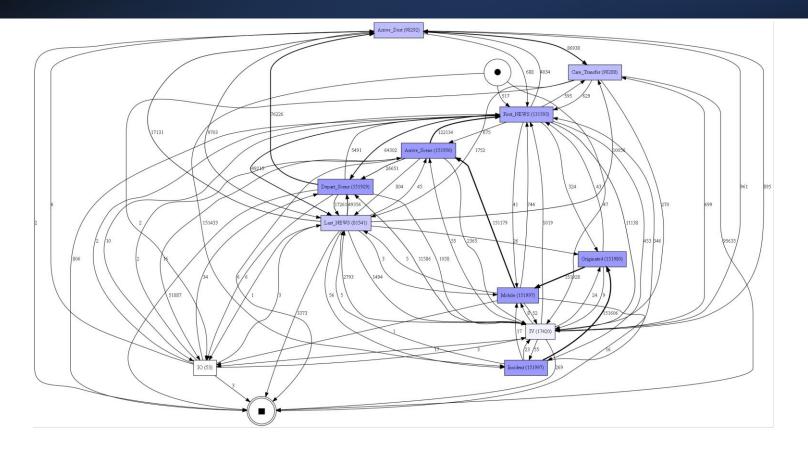
#### Business Rules – Remove Cases with Missingness

# Filtered to Only Including Category 2 we originally had 152,525 patients:

Lost 528 Patients to Business Rules

Remaining Patients after removing patients that don't follow rules 151,997

### **Business Rules Applied**



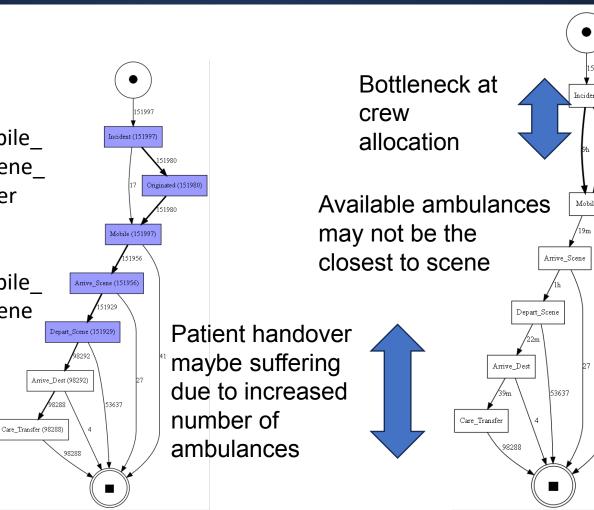
Filtered to Only Including Category 2 we originally had 152,525 patients: Lost 528 Patients to Business Rules Remaining Patients after removing patients that don't follow rules 151,997

#### Remove "Optional" Events

8 Variants

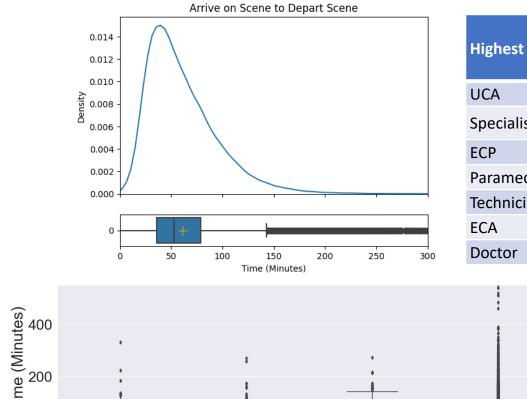
~64% follow Incident\_Originated\_Mobile\_ Arrive\_Scene\_Depart\_Scene\_ Arrive\_Dest\_Care\_Transfer

~35% follow Incident\_Originated\_Mobile\_ Arrive\_Scene\_Depart\_Scene

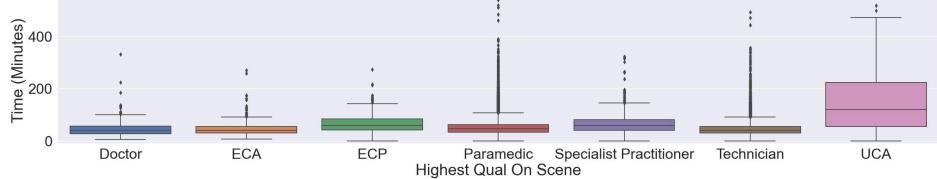


Originated

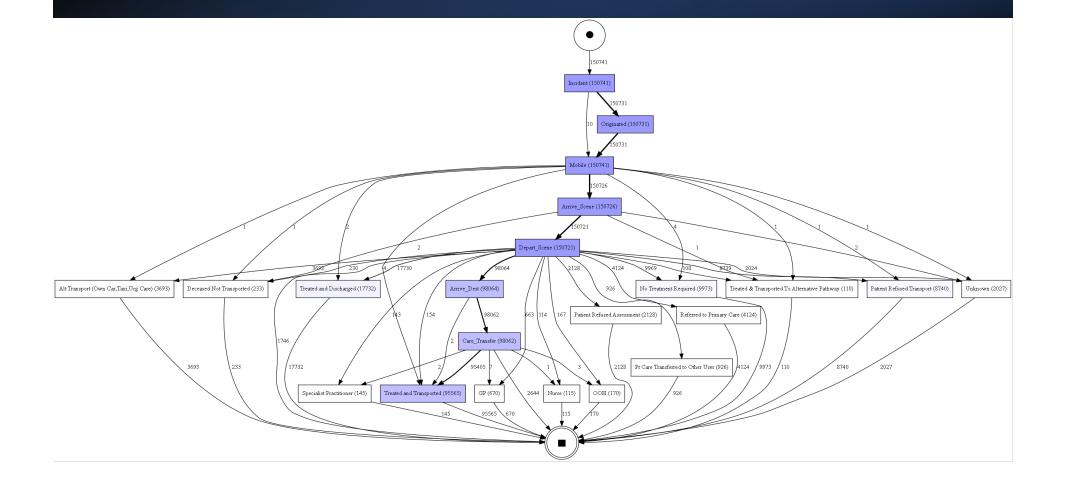
#### Analysis – Ambulance Arrive on Scene to Ambulance Depart Scene (Time on Scene)



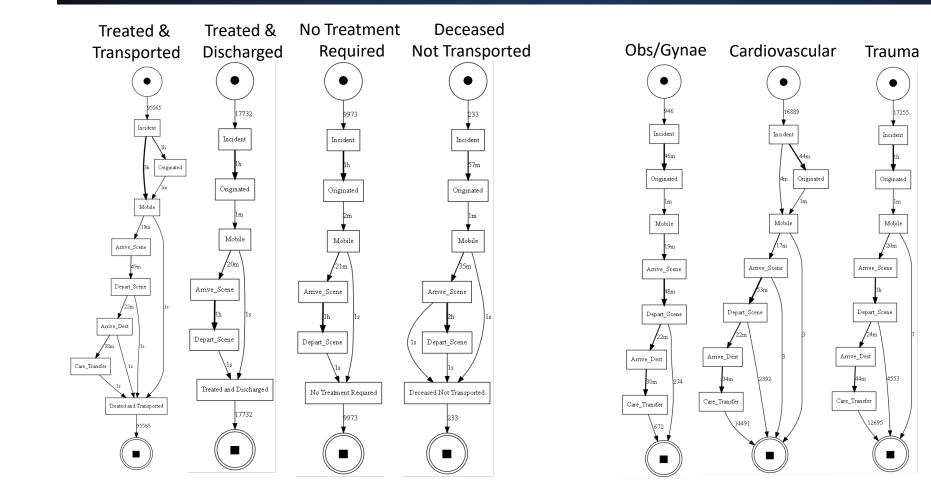
Highest Qualification on Scene	Mean effect on mean time between Arrive_Scene and Depart_Scene (minutes)
UCA	97.898471
Specialist Practitioner	14.543051
ECP	13.385187
Paramedic	1.229887
Technician	-3.018392
ECA	-4.630408
Doctor	-10.980780



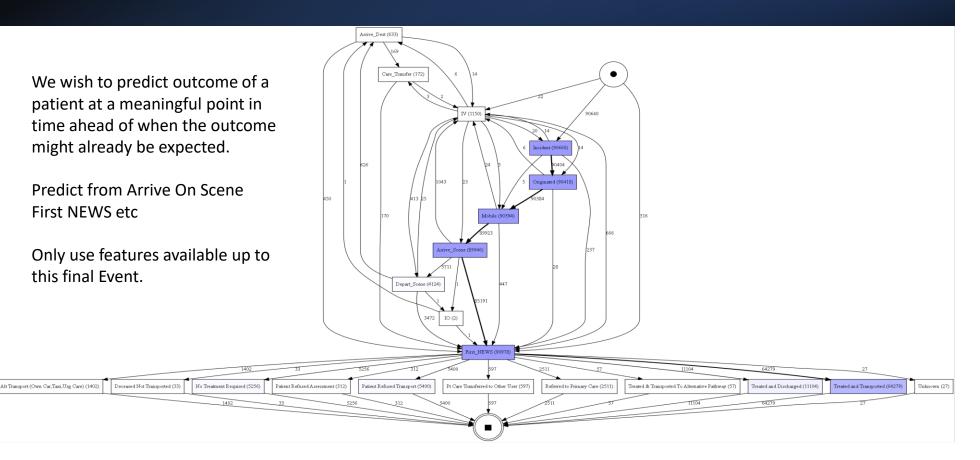
### Add Outcome As Events



### Filter by Outcome or by Category



#### Applying Machine Learning to Predict Job Cycle Outcome



#### Modelling - Decision Tree to Predict Outcome

- Class imbalance with Treated and Transported ٠
- Under sampled and grouped other classes into an umbrella class. ٠

Leaf Colour by Class Feature Importances Feature 100% 100% Treated and Other Warning Score (NEWS) Transported Age of Patient event RFWS <= 2.5 girl = 0.48 samples = 44055 Hour of Arrival on Scene Day of the Week Highest Qualification on Scene

**Decision Tree** 

#### Importance Initial Clinical Category of Call 0.29 First Recorded National Early 0.23 0.17 0.10 Index of Multiple Deprivation (IMD) 0.075 **Decile at Incident Location** 0.065 0.038

### Findings & Resources

- Observed bottlenecks at crew allocation and handover.
- Measured the effect of different attributes on the response time, time spent on scene and handover time
- Tried to predict whether a patient was to transported to hospital. More data required to do this accurately.

#### **Resources**

PM4PY: <u>https://pm4py.fit.fraunhofer.de</u> BUPAR: <u>https://bupar.net</u>, Celonis: <u>https://www.celonis.com</u> Disco: <u>https://fluxicon.com/disco</u>, Prom: <u>https://promtools.org</u>

#### Project Report and Code available on NHS England's GitHub

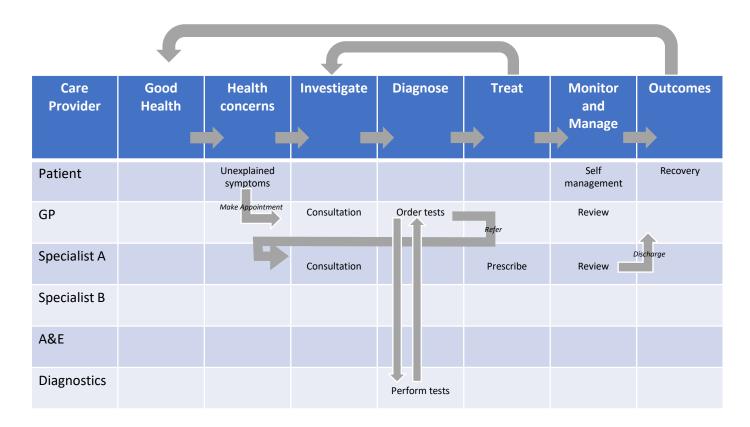
#### https://github.com/nhsengland/ProcessMining

Process Mining to Generate Healthcare Pathways
Alexander Coles Paul Carroll Martina Fonesca
June 2023
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2       Process Mining Case Study with East Midlands Ambulance Service       4         2.1       The UK Ambulance Service       4         2.2       East Midlands Ambulance Service Data       5         2.3       Process Discovery.       7         2.4       Conformance       8
2.5       Enhancement and Further Analysis       13         2.5.1       Outcome Prediction       13         3       Conclusion       13
1 Introduction
The increased digitisation of events/records within process flows is facilitating their analysis for evaluation and improvement. Examples of events range from accepting a purchase order in a business, receiving an email in a chain and administering medication in a hospital. With all this information now being electronically

# Care Pathways in the NHS follow typical patterns

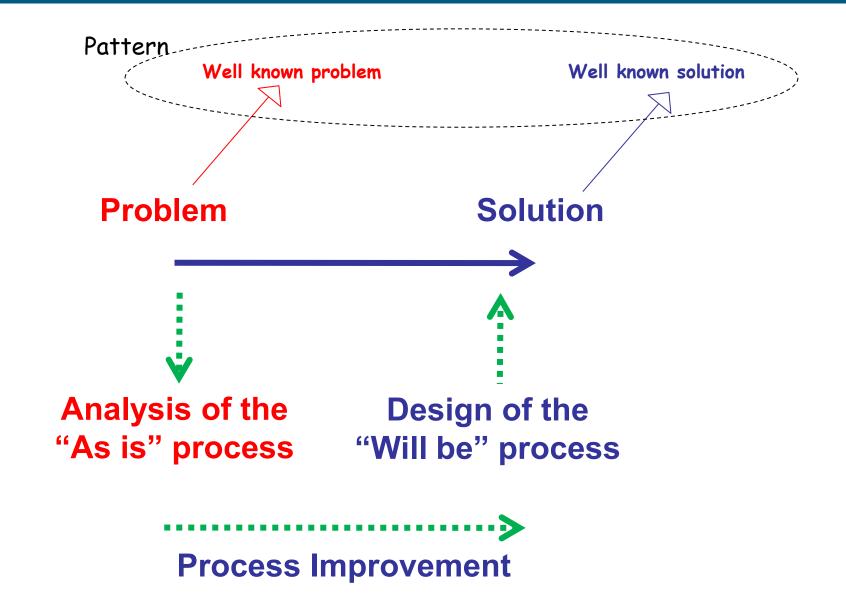


Most care pathways will follow one or more standard template patterns Linear, Cycle, Referral, Sub-process, Parallel process Opportunity to share and transfer learning within the Digital Health and NHS community.





### Designing new Digital Pathways



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## Process Mining to Identify Pathway Improvement Opportunities

Leeds Digital Health 6pm 31<sup>st</sup> January 2024

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