Data Science and AI for Public Good: Lessons from cross-sectoral collaboration

Slava Jankin Mikhaylov
CSA to Essex County Council
University of Essex

- Cross-Sectoral Collaboration
- Case 1: Catalyst project
- Case 2: California Policy Lab
- Case 3: Essex Centre for Data Analytics

Cross-sectoral partnership

The world's most innovative economy ambition

Partnership between government, industry, and academia:

- Attract and retain Al talent;
- Improve digital infrastructure;
- Conducive business climate;
- Prosperity of society through Albenefits.

£1bn + £1.7bn from the Industrial Strategy Challenge Fund



Catalyst | A LOCAL PARTNERSHIP IMPROVING COMMUNITY SERVICES

Catalyst is funded by the Higher Education Funding Council for England.



University of Essex working in partnership with:











"Our ongoing partnership with the University of Essex puts us at the forefront of local authorities seeking innovative solutions to the challenges we face."

Councillor David Finch

LEADER OF ESSEX COUNTY COUNCIL AND CHAIR OF ESSEX PARTNERS

Boosting the relationship







Boosting the relationship









20 interviews: SCC, ECC, and UoE stakeholders

3 core issues:

- Challenges to knowledge transfer;
- Challenges to cooperation;
- Leadership challenges

```
ក្ខេង researchers បទ្ទេប management anomer of challenge problems someone benefits _8
buy ਨੂੰ funding big give runproblem analysts ਹੁੰ
```

Challenges to knowledge transfer:

- Weak personal connections (10) (Van Wijk 2008)
- Lack of social cohesion (7) (Inkpen 2005; Easterby 2008)
- Multiple/conflicting priorities (8) (Cummings 2003)
- Risk aversion (4) (Gho 2002)

Challenges to cooperation:

- Communication (4) (O'Flynn 2018)
- Lack of shared collaborative reality (1) (Argote 2000;
 O'Flynn 2018)
- Institutional forces (7) (McEvily 2003)
- Lack of flexible adjustments (3) (Chen 2004)

Leadership/management challenges:

- Role clarity (9) (Heen 2009)
- Low middle management and bottom level buy-in (5) (Easterby 2008)
- Misalignment between research and developing project objectives (4) (Easterby 2008)
- Lack of facilitative leadership (9) (Ansell and Gash 2007)



CALIFORNIA Experience

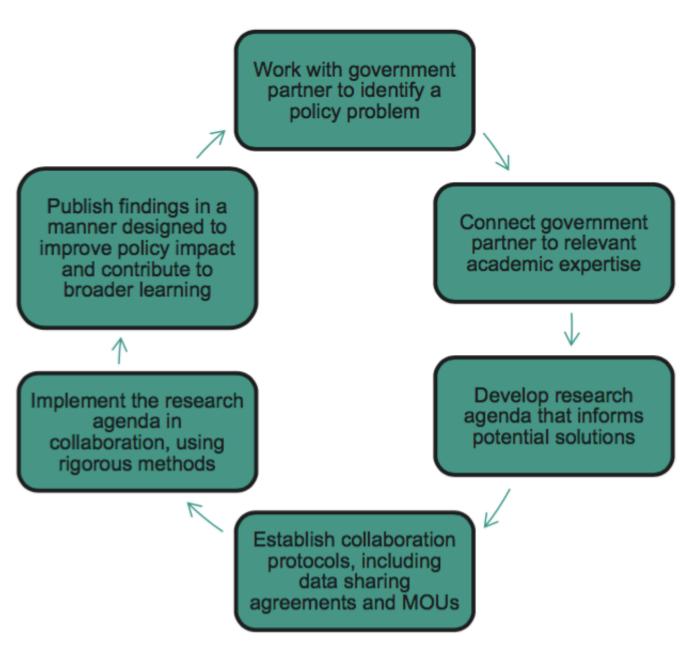
- Main barriers for academics: lack of access to data, communication challenges, questions not being aligned (with government), and a publish or perish mentality;
- Main barriers for civil servants: data silos, lack of capacity and time, changes in leadership and strategy, political pressure, and rapid pace of decision making.
- CPL solution is to fuse unity is through diversity



Governance

Key standards:

- Effective creation of data sharing agreements;
- Ensuring data confidentiality;
- Pairing government agencies with appropriate experts.



Essex Centre for Data Analytics (ECDA)







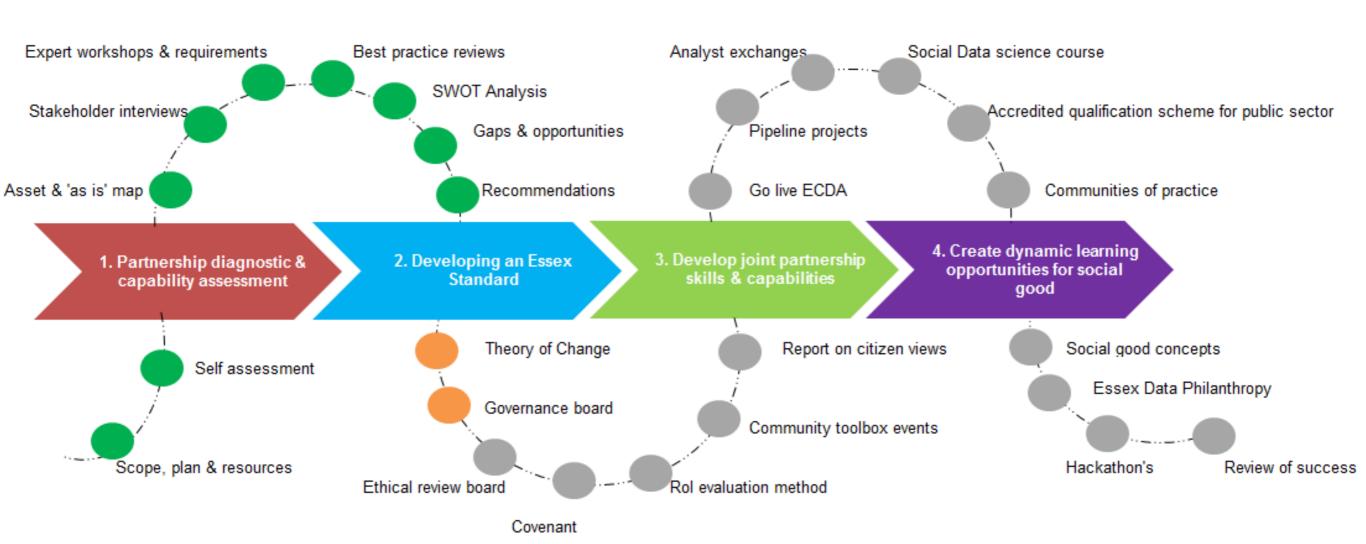
ECDA

The Vision: To make Essex national leaders using the power of data science and AI to tackle public policy challenges

The Aims:

- To make Essex a place that is an exemplar for the integration of data across public bodies.
- To have the skill, capability and technology to undertake predictive analytics based on ethical, high standards.
- To have a sustainable data infrastructure.
- To have the best data science / analytical capabilities in the UK to benefit our people and communities.

ECDA: Pipeline



ECDA: Self-assessment



What we do well

- Highly skilled workforce & availability of training
- Predictive analytics experience and application
- Leadership Buy in
- Credibility & reputation

• What we need to do better

- Sustainability & operational ability
- Citizen voice & transparency
- Infrastructure
- Impact



Opportunities

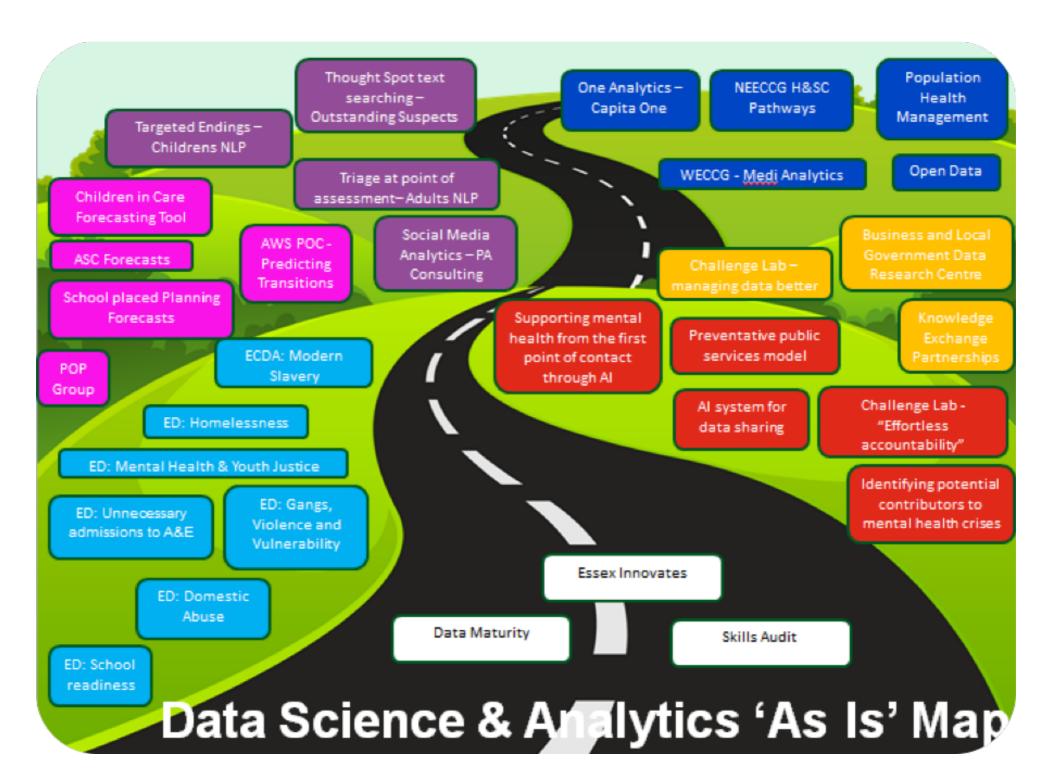
- Strategic Alignment & place shaping
- Data expeditions
- Connectivity
- Speciality dishes



Considerations

- Infrastructure & data sharing platform
- Governance
- Ethical review & assurance

ECDA: Roadmap



ECDA: Opportunities

Evidence Based
Decision Making:
Making Data
Accessible; Single
View of Customer

Place Shaping: Geo-Spatial Analysis; Mapping; Hot-spot Analysis

of Resources: Risk Stratification; Predictive Risk Modelling; Segmentation PERFORMANCE REPORTING* SCENARIO TESTING

DATA VISUALISATION KEY DRIVER ANALYSIS
DEEP DIVES EXPLORATORY DATA ANALYSIS

FORECASTING & DEMAND MANAGEMENT

STRATEGIC NEED ASSESSMENTS EVALUATION

PERFORMANCE BENCHMARKING BEST PRACTICE REVIEWS

GEO-SPATIAL ANALYSIS & MAPPING*

MACHINE LEARNING ARTIFICIAL INTELLIGENCE
PROJECT SCOPING STATISTICAL SIGNIFICANCE TESTING

RISK STRATIFICATION*

PATHWAY MODELLING

DATA MANAGEMENT DASHBOARD DESIGN

DATA INTEGRATION REPORT WRITING (PROGRAMMING)

NATURAL LANGUAGE PROCESSING*

POPULATION HEALTH MANAGEMENT

Demand

Management:

Forecasting; Pathway Modelling; Scenario testing/Simulation

Key Added-Value Skills for Development

Improving

Productivity and

Practice: Qualitative Analysis; Sentiment

Analysis; Natural

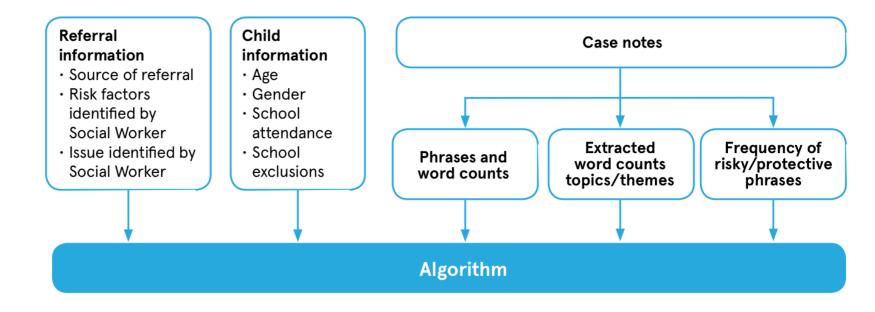
Language Processing

ECDA: NLP in policy

- Children's social care
- Given the text of the initial referral and assessment, and structured data relating to the case, could we predict whether the case would be re-referred and escalated if it were closed?
- ML and NLP

Cabinet Office
Behavioural Insights Team

Figure 5: The inputs of the machine learning algorithm used to detect escalated closed cases.



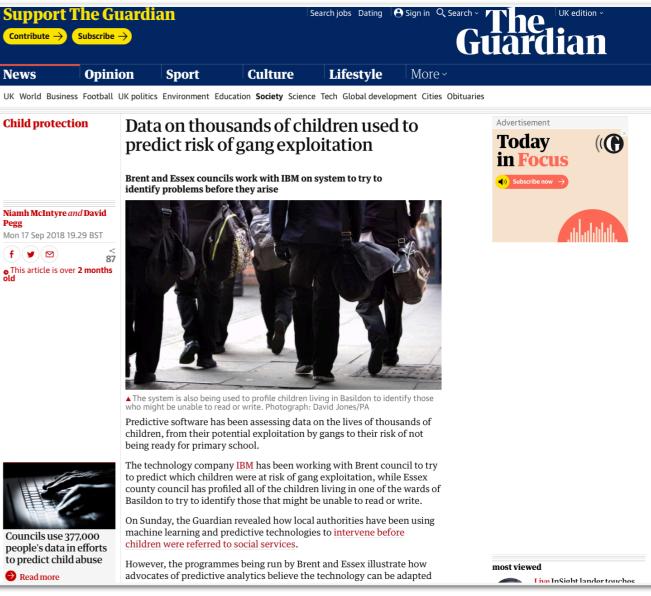
ECDA: Risk stratification

- Hackney & Thurrock councils and tech company Xantura developed a predictive system to identify vulnerable families in need of additional council support.
- Brent and IBM developed a risk model to identify children at risk of criminal exploitation.









ECDA: Essex Standard

Openness, transparency, and ethics underpin ECDA work

Thank you! s.mikhaylov@essex.ac.uk