

Agenda



- 1. Financial Informatics at Glasgow
- 2. A timeline towards impact
- 3. Before IAA
- 4. During IAA
 - 1. PPC-FI
 - 2. FAR-Market
 - 3. FAR-AI
- 5. After IAA
- 6. Lessons learned and advice





Financial Informatics

- Financial Informatics is a research theme hosted by the Information, Data and Analysis section
 - It represents a cross-cutting group of researchers in Computing Science working on the research and development of AI and Information Retrieval technologies applied to financial use-cases and data



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Dr. Javier Sanz-Cruzado

Financial Recommendation Systems

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We specialize in the development of sophisticated AI-powered services, which can collate, analyse and apply financial data in real-time



Three Research Pillars

Collate Diverse Information from Multiple Modalities

Market Data

Capture how traders value assets over time an in different market conditions

Asset
Fundamentals

Know your business, track company performance, scale and growth

First Party Statements How does a business view itself and its prospects?

News and Social Media How does the world view the business and its products? What impact do world events have

Customer Data Know your customer via their transactions, investments and situation data

Analyse to Build Models, Data Structures and Gain Insights

Event Detection

Identify financially relevant events, whether within-company or from the wider world

Technical Analysis Convert financial time series to human understandable KPIs

Knowledge Graphs Model financial facts and their relation ships, as well as how they evolve over time

Financial Language Models Build adapters for large language models, enabling them to better handle financial terminology

Content Categorization Develop AI models to classify and tag financial documents

Apply our technologies to tackle real financial use-cases

Asset Recommender Suggest suitable personalized investments for customers and advisors

Portfolio
Construction

Build optimized portfolios for investors

Robo-Advisors

Answer financial questions and provide advice

Automatic Trading Automatically manage investment portfolios in real-time

Sustainability Optimization Optimize back-end financial data services to reduce carbon costs







Passive Impact

A company reads one of your research papers and then implements some or all of it

...but how do you track it?



R&I Projects

Have a project with a company that can test your invention and maybe put it into production for you



Patents

Register a patent for an invention, companies can then pay a small fee to use your invention

Difficult to do for Computing Science, and the Uni budget for patents is small



Spin-out Company

Licence-Based: Create a piece of software or algorithm and then convince customers to pay a licence to use it

Product-Based: Create a real solution and market it to customers

IAA can help us here!



Our Timeline Towards Impact

The Foundation



- Before any IAA journey, you need to start from the core research
 - We need to be expert in something before we can think about impact
- In our case this was experience in developing recommender systems
 - Movies, Groceries, Venues to Visit (we have done a lot of this over the last decade)
 - Get your early research published, you will need to demonstrate your expertise in an area



Exploring Data Splitting Strategies for the Evaluation of Recommendation Models

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ABSTRACT

Effective methodologies for evaluating recommender systems are critical, so that different systems can be compared in a sound manner. A commonly overlooked aspect of evaluating recommender systems is the selection of the data splitting strategy. In this paper, we both show that there is no standard splitting strategy and that the selection of splitting strategy can have a strong impact on the ranking of recommender systems during evaluation. In particular, we perform experiments comparing three common data splitting strategies, examining their impact over seven state-of-the-art recommendation models on two datasets. Our results demonstrate that the splitting strategy employed is an important confounding variable that can markedly alter the ranking of recommender systems. making much of the currently published literature non-comparable, even when the same datasets and metrics are used.

Information systems → Recommender systems; Test collec-

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and different scenarios (such as session-based recommendation [29] and sequential recommendation [14]). Many approaches have been proposed to solve these tasks over the last two decades, among which neural network-based recommendation models are currently very popular, due to their high effectiveness and adaptability to different sub-tasks and scenarios [29]. As the recommender systems field matures, advances in performance naturally become more incremental, leading to smaller increases in model effectiveness. This places more strain on the evaluation methodology's ability to distinguish between systems with similar performance, as researchers

and practitioners chase ever smaller performance gains. With the current influx of very similar neural network-based recommendation models being published, there needs to be increased emphasis placed on eliminating confounding factors that can lead to uncertainty during evaluation, otherwise it will be impossible to confidently determine whether gains are truly being made. In the Information Retrieval (IR) domain, standardization efforts such as TREC, and other evaluation initiatives like NTCIR, CLEF and FIRE laid down guidelines on what constitutes a sound evaluation



BETA-Rec: Build, Evaluate and Tune Automated Recommender Systems

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University of Glasgow

The field of recommender systems has rapidly evolved over the last

few years, with significant advances made due to the in-flux of deep

learning techniques. However, as a result of this rapid progress,

escalating barriers-to-entry for new researchers is emerging. In

particular, state-of-the-art approaches have fragmented into a large

number of code-bases, often requiring different input formats, pre-

processing stages and evaluating with different metric packages.

Hence, it is time-consuming for new researchers to reach the point

of having both an effective baseline set and a sound comparative

environment. As a step towards elevating this problem, we have

developed BETA-Rec, an open source project for Building, Evalu-

ating and Tuning Automated Recommender Systems. BETA-Rec

aims to provide a practical data toolkit for building end-to-end

recommendation systems in a standardized way. It provides means

for dataset preparation and splitting using common strategies, a

ABSTRACT

Shangsong Liang Yucheng Liang Guangtao Zeng Junhua Liang Sun Yat-sen Univeristy

Qiang Zhang University College London

Recommender Systems, Framework, Open-source, Toolkit

ACM Reference Format:

KEYWORDS

Zaigiao Meng, Richard McCreadie, Craig Macdonald, Iadh Ounis, Siwei Liu, Yaxiong Wu, Xi Wang, Shangsong Liang, Yucheng Liang, Guangtao Zeng, Junhua Liang, and Qiang Zhang. 2020. BETA-Rec: Build, Evaluate and Tune Automated Recommender Systems. In Fourteenth ACM Conference on Recommender Systems (RecSys '20), September 22-26, 2020, Virtual Event, Brazil. ACM, New York, NY, USA, 3 pages. https://doi.org/10.1145/3383313.

1 INTRODUCTION

Recommender systems that suggest items of interest to users based on available information such as purchases and interactions histories have been the subject of intensive research by both industry mo [2 E 12] In

Large Multi-modal Encoders for Recommendation

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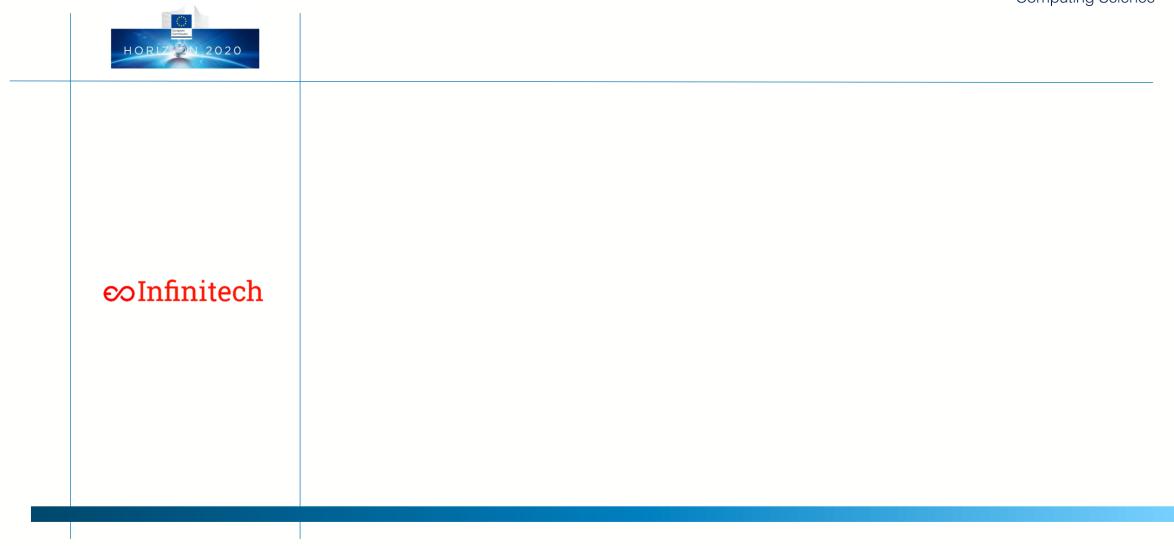
In recent years, the rapid growth of online multimedia services, such as e-commerce platforms, has necessitated the development of personalised recommendation approaches that can encode diverse content about each item. Indeed, modern multi-modal recommender systems exploit diverse features obtained from raw images and item descriptions to enhance the recommendation performance. However, the existing multi-modal recommenders primarily depend on the features extracted individually from different media through pre-trained modality-specific encoders, and exhibit only shallow alignments between different modalities - limiting these systems' ability to capture the underlying relationships between the modalities. In this paper, we investigate the usage of large multi-modal encoders within the specific context of recommender systems, as these have previously demonstrated state-of-the-art effectiveness when ranking items across various domains. Specifically, we tailor

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Traditional recommendation systems primarily rely on user-item interactions to provide personalised recommendations, which may not fully capture the rich information embedded in the diverse forms of data associated with items, such as images, texts, and audio [32]. Multi-modal recommendation systems address this issue by representing items using encodings from multiple modalities, and hence provide more effective recommendations [12]. By incorporating various modalities, multi-modal recommenders [2, 8, 13, 16] bridge the gap between general recommendation systems and the complexities of multimedia item content. However, the existing multi-modal recommenders [8, 13, 16] fuse the extracted multimodal features into user/item representations without sufficiently addressing the complex and inherent correlations between different modalities [25]. For example, MMGCL and LATTICE (as detailed in Table 1), fail to effectively fuse the multi-modal features. Contrary

Timeline





October 2019

April 2023

Infinitech Project



- Based on our research profile, we were invited to an EU H2020 project
 - These are both research and innovation projects
 - Usually focused on taking expertise and applying it for target use-cases, but are precommercialization
 - Can act as a good starting point to move research towards innovation



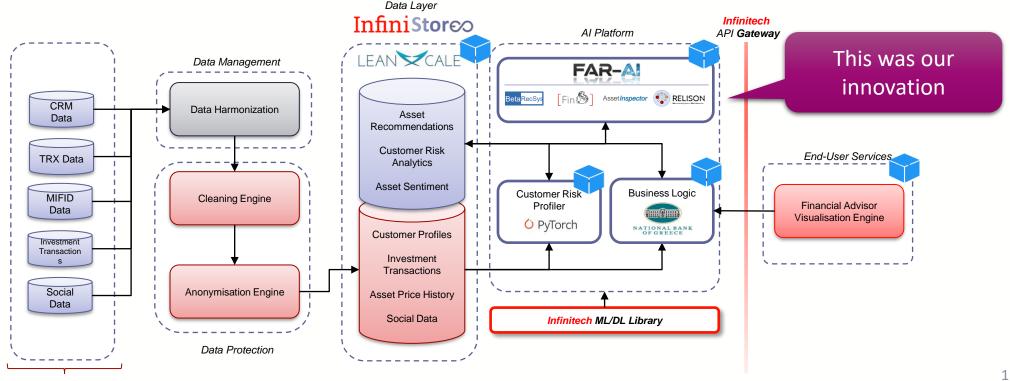
Goal: We were tasked to build a **financial recommender system** for banking customers (recommend investments)

- Primary investigators: ladh Ounis, Craig Macdonald, Richard McCreadie
- Dates: October 2019 March 2023

Research to Prototype



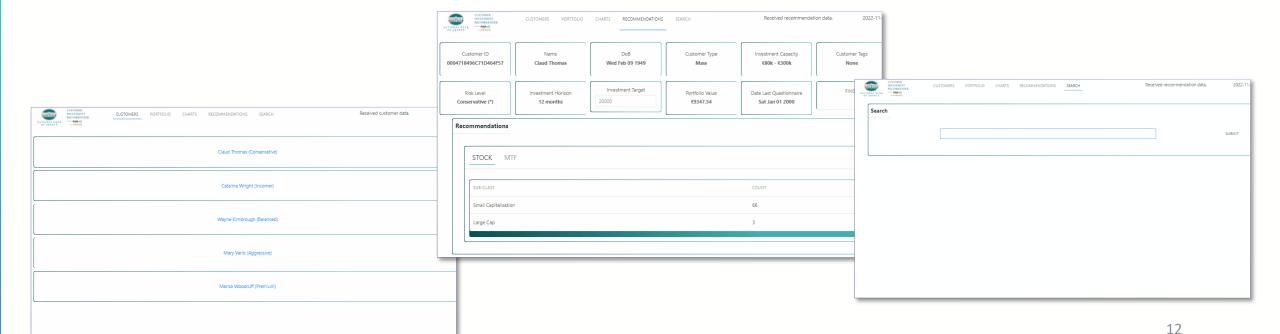
- As part of this type of project you usually advance your core technology and build a prototype for testing with a Pilot, in our case the National Bank of Greece
 - This is where you start to think about the practicalities of converting your innovation into a product (what you care about is often only a small part of a larger ecosystem)



Research to Prototype

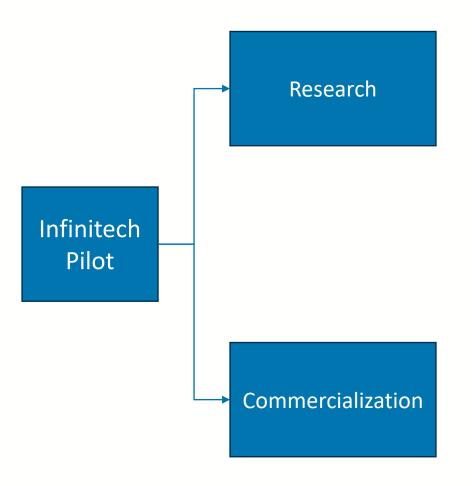


- As part of this type of project you usually advance your core technology and build a prototype for testing with a Pilot, in our case the National Bank of Greece
 - This is where you start to think about the practicalities of converting your innovation into a product (what you care about is often only a small part of a larger ecosystem)
 - If you are lucky, impact could happen here if the Pilot company wants to buy your solution, but usually more refinement is needed at this point



Infinitech ends March 2023... then what?



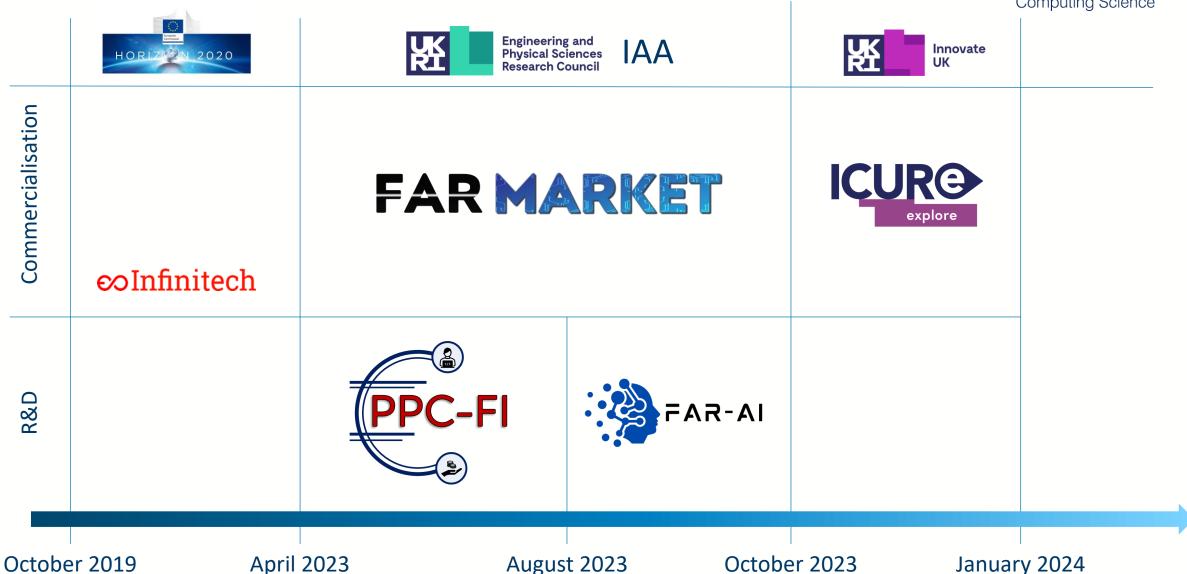


- Continue the financial informatics theme
- Keep partnership with the National Bank of Greece
- Improve our product

- License our technologies to NBG
- Potential to sell the product to third parties
- Creation of spin-out company

A timeline towards impact

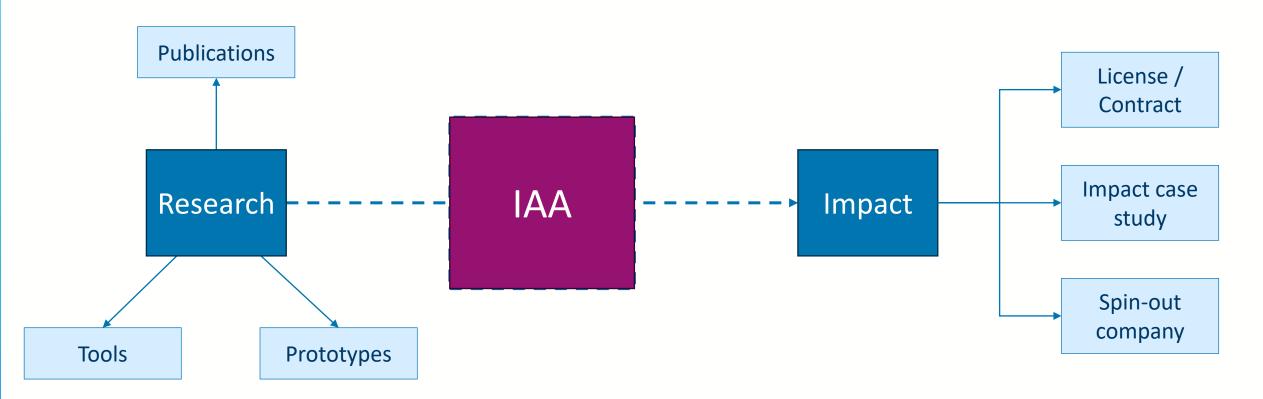






EPSRC Impact Acceleration Account (IAA)





EPSRC IAA funds **innovation** projects aimed towards the creation of **impact** based on **existing research**

EPSRC Impact Acceleration Account (IAA)



- EPSRC IAA are not research projects!
 - New academic research is not eligible
 - Funding academic-only conference attendance is not allowed
- They are an **intermediate step** towards impact
- 2 types of projects: standard calls and RA-led calls

| | Standard Call | RA-led Call | |
|---------------------------|---------------------------|---------------------|--|
| Primary investigator (PI) | Research & Teaching staff | Research Associates | |
| Funding | Up to £50,000 | Up to £20,000 | |
| PDRA time (estimated) | Up to 10 months | Up to 4 months | |
| PI salary allowed | No | Yes | |

The EPSRC IAA process





- Identify impact route
- Engage collaborators
- Write proposal
- Gather support letters
- Send to IAA team

- Develop project
- Meet with collaborators
- Impact and outreach activities
- Engage with commercialization team (if needed)

- Write final report
- Report impacts (Enlighten KE)

Our IAA projects



Towards licensing / impact case study



Towards commercialization / spin-out company





The PPC-FI project





PPC-FI: Personalized Portfolio Construction for Financial Investments

- RA-led EPSRC IAA project
- **Funding:** £17,586
- **PI:** Dr. Javier Sanz-Cruzado
- April 2023 July 2023

Partner:



Development of personalized portfolio recommendation and optimization algorithms for retail customers

Project concept

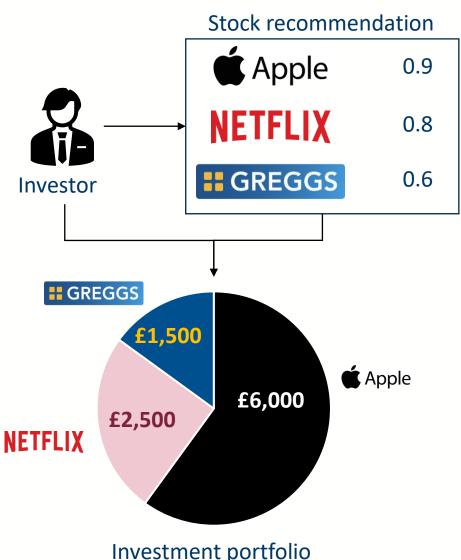


Improve and refine an existing prototype to license it to external partners

- No major research required
- Allowed continuous collaboration with NBG
- Potential application of research within NBG's daily operation (impact case study)

Project description

- Development of personalized portfolio construction algorithms
 - Identify financial assets for retail investors
 - And the amount of money they should invest on each
 - Adapted to needs of the investor (risk, horizon, capacity)
- Integration on the Infinitech prototype



Leasons learned...



- Focus on **impact**, not on research
- This is a **collaboration** with a partner, not a R&D contract
 - Get a strong support letter (including actual or in-kind costs from your collaborators)
 - What are the benefits for the University?
 - What are your partners contributing?
- Identify the benefits for the PDRA
 - IAA does not fund contract extensions
 - Clearly estate what advantages leading this project has for you.
- Ask for support from the IAA team (Keith Dingwall)
 - Pre-submission (form review)
 - Post-submission (improvement points and panel feedback)

The FAR-Market project





- EPSRC IAA project
- **Funding:** ~£35,000
- PI: Dr. Richard McCreadie
- June 2023 September 2023

Identify potential markets into which we could sell our product

Project Concept





- Tell them about our product
- Have them
 - Identify competitor companies
 - Interview potential customers
 - Write a report

You will need support from IP and Commercialisation to get this type of project approved



AUTOMATION IN PORTFOLIO & WEALTH MANAGEMENT



FEBRUARY 2024, VERSION 1 UNIVERSITY OF GLASGOW RESEARCH



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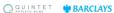
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Research methodology

This research conducted by findexable was compiled through primary and secondary research. Primary included a series of interviews with senior executives within investment management at large global and regional banks in Europe. Secondary research concentrated on current trends in AI deployment across the wealth management industry and a global AI benchmarking exercise. Interviews were conducted between July and October 2023.















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2 AUTOMATING WEALTH



Lessons Learned...



Its really difficult to get the University to pay an external contractor:

- Wrote the initial proposal in August 2022
- Submitted in January 2023
- Approved in March 2023
- Spent 6 months fighting with finance and legal over contract terms
- Project eventually started in around September 2023
- Report was delivered in January 2024

By the time we eventually finished this, it was already out of date:

- Our view on the product had changed
- We had a better idea of the landscape and had done our own market discovery by going to trade shows – which was much more useful

I don't recommend this route!

The FAR-AI project





FAR-AI Deployment of AI-based Financial Asset Recommendation System

- RA-led EPSRC IAA project
- **Funding:** £10,658
- PI: Dr. Javier Sanz-Cruzado
- August 2023 October 2023

Building a financial asset recommendation platform for the UK banking market

Project concept



Prepare an existing prototype for demonstration and commercialization

- No new research
- Make easier to reach interested third parties
- Potential for spin-out creation

Project description

- Collection of UK financial set
 - Remove constraints from external data
 - Adapt to the target market
- Development of new prototype features
- Celebration of a workshop on Scotland Fintech Festival
 - Title: "What's my investment? Automated recommendations for investors".
 - Date: October 5th 2023



Prototype demonstration



Workshop materials

Lessons learned...



- Most lessons from PPC-FI project still apply here
- Although desired, you don't need an external partner for IAA
- In that case, find support from IP & Commercialization team (Darian Brookes)
- Building a product is not enough
 - Impact cannot be created if we keep the products in our labs
 - Do not restrict to academic venues
 - Explore festivals and events to engage with interested parties
 - Show your product to others
- Be aware of the project timeline



ICURe Explore





- Innovate UK
- Funding: £35,000 (up to £15,000 for salary, £20,000 for market discovery)
- PI: Dr. Richard McCreadie
- October 2023 January 2024

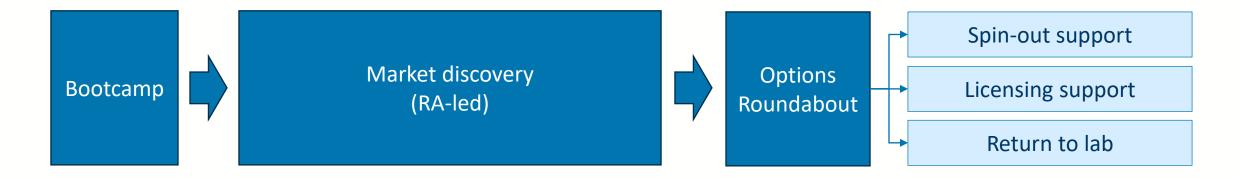
Explore the potential commercialization of research technologies

- The team
 - Entrepreneurial lead: PDRA / Early career researcher (Dr. Javier Sanz-Cruzado)
 - Principal Scientific advisor: Senior academic (Dr. Richard McCreadie)
 - **Technology Transfer Officer:** Commercialization manager (Darian Brookes)
 - Business advisor: Usually external to the University (Richard Braidwood)

Programme concept



- 3-month support for market discovery activities
- Led by the entrepreneurial lead



- What do we mean by market discovery activities?
 - Learn the current state of the market
 - Engage with potential customers
 - Identify partners, competitors
 - Establish the position of your research in the market

Our ICURe Explore journey

University of Glasgow School of Computing Science

We explored the market by attending trade shows and conferences all over the world

7 EVENTS



>50
COMPANIES

>20 COUNTRIES



Next steps



Towards licensing / impact case study

- Academia-industry research collaborations
- Further IAA collaborations

Towards commercialization / spin-out company

- Scottish Enterprise High-Growth Spinout programme
- Talk with investors



A PDRA Point of View



Pros

- Growth opportunities
 - Funding application training
 - Leadership and management skills
- Positioning yourself outside University
- Further collaborations with external partners

Cons

- Intermediate steps towards impact need to secure funding before end of project
- Very short projects (up to 4 months without external funding)
- No time for generating research outputs

An Academic Point of View



Pros

- Innovation funding can give you momentum to turn some research you have been doing into impact – which is not only good for the country but your profile
- Even if it does not work, you learn more about what customers actually care about!
- Successful innovation activities can lead to impact case studies

Cons

- Its significantly more work added on to your normal schedule
- It might all be for nothing most start-ups fail!
- Opportunity cost: working on this means less <everything else>
- The university is not that well set-up to enable innovation (yet)

Questions?



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