

Project Fellows: Nathan Coulson (Birkbeck, Univ. of London), Rosa Lavelle-Hill (Univ. of Nottingham), Tobias Richter (Univ. of Cologne)
Technical Mentor: Liliana Millán | **Project Manager:** Sara Guerreiro de Sousa | **Partner Team Coordinator:** Cristina Faro

Background

IEFP is the government institute responsible for reducing unemployment in Portugal (6.6% in 2019)

In 90 job centers across Portugal, dedicated job counsellors face 3 key constraints in recommending the best interventions for job seekers:

- Limited time**
- Limited experience**
- Manually searching** through a large list of possible interventions

IEFP counselling process



Problem statement

Which **interventions should job counsellors recommend to job seekers** to help them find meaningful employment?

Data

9 years of IEFP data (2010-2019)

3.1m job seekers

101m transactions

255 training courses

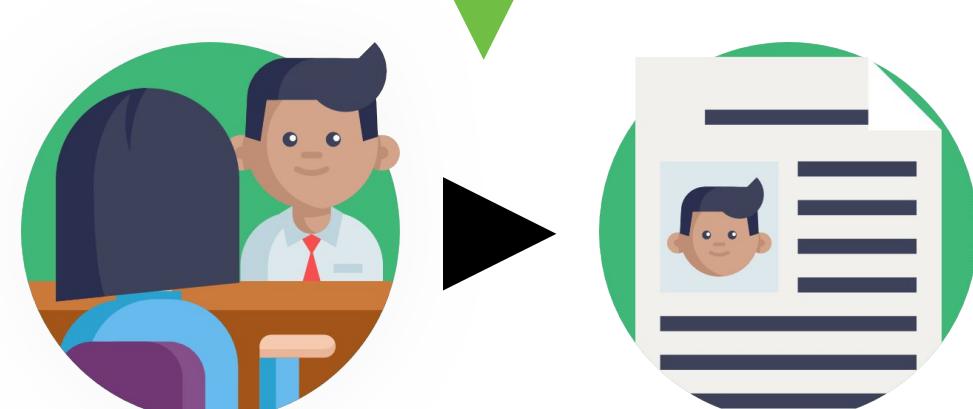


Solution

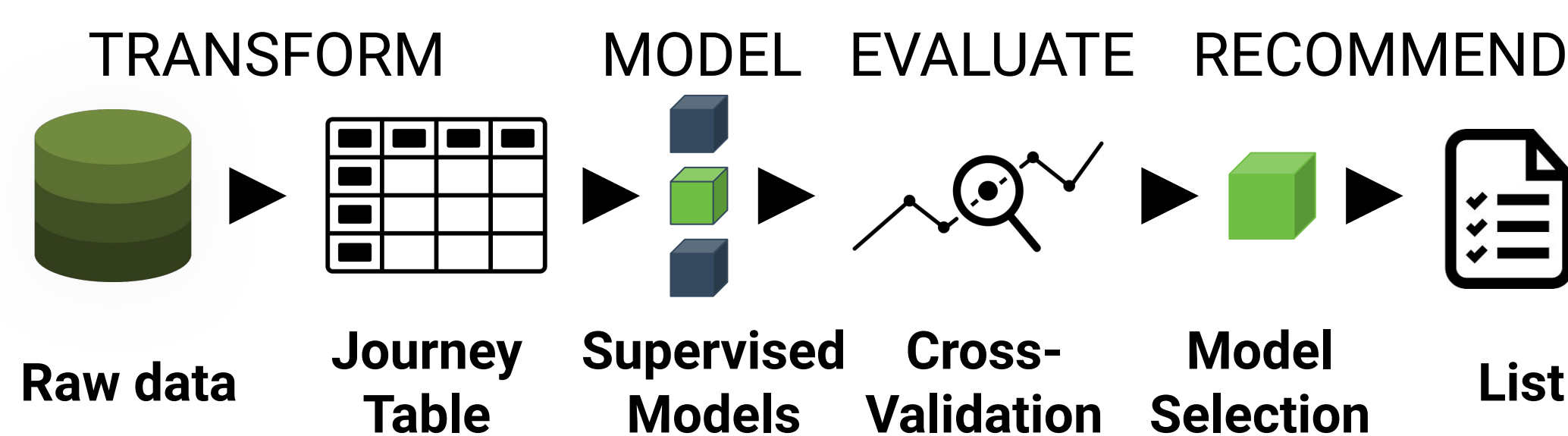
A system which suggests **relevant and effective interventions** for a job seeker

Relevant: meet the needs of the job seeker

Effective: optimised for an outcome: i.e. time to employment



Approach



Transform

Our raw data source consists of over 50GB of transaction records spread out over 16 Oracle DB tables.

A key challenge was defining a set of rules, through domain knowledge, that facilitates the transformation from **transactional data** into a modelling table consisting of **user journey** data.

ID	Journey (10)		Demographics (25)		Interventions (60)		Outcomes (3)
	Journey	Start	Age	Education	CV Workshop	Internship	Time to job
23	1	03/01/2017	31	lower	22/07/2017	-	168
23	2	16/02/2018	32	lower	-	15/05/2018	-
56	1	04/05/2017	51	higher	08/05/2017	22/11/2017	402

Model

Train models on different employment outcomes

Demographics, Preferences, Interventions → **train** → **target** → Fast employment, Long-term employment, Skills match

Find set of interventions that optimize the desired outcome

Demographics, Preferences, Interventions → **fix** / **permute** → **rank predictions** → Top-k interventions

Evaluate

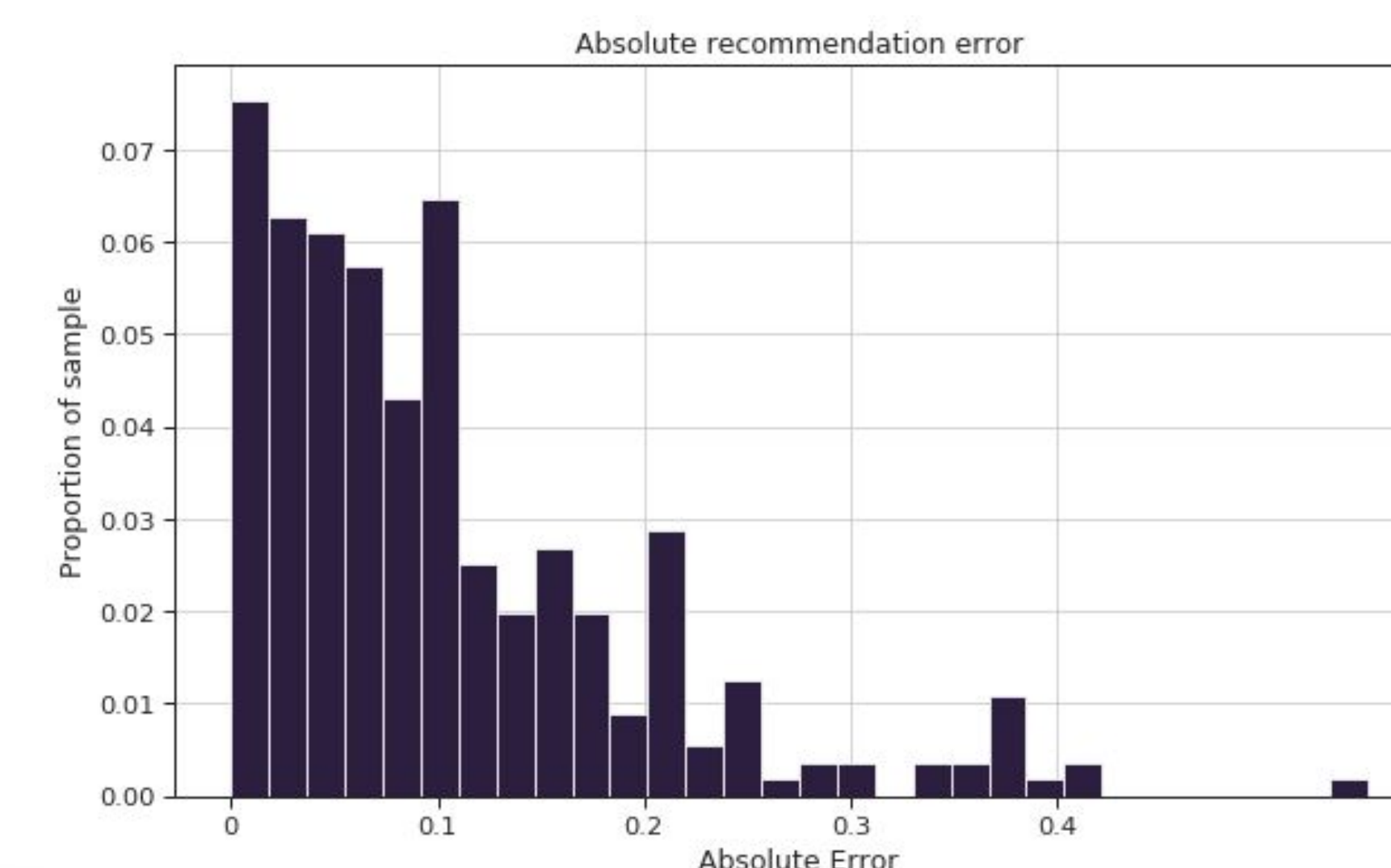
Evaluate recommendations by comparing with ground truth of reference groups

Test data (2019) → **predict & rank** → Top-k interventions → Predicted success with taken interventions

Find reference group that took interventions → Calculate mean success rate of reference group

↓ **Error**

Results



Recommendations generated by the **Random Forest** model performed the best in the ground truth evaluation process with a mean error rate of 0.096.

Output

Below is an **example of a mock-up** that shows the output of the recommender system:

Intervention Recommendations

UTE ID: 235245

	Code	Intervention Description	Predicted time to employment
Option 1	134	MEDIA TRAINING COURSE	6 months
	7	CV WORKSHOP (CREATIVE)	
	57	JOB SEARCH TECHNIQUES	
Option 2	134	MEDIA TRAINING COURSE	11 months
	13	INTERVIEW WORKSHOP (CREATIVE)	
	156	ADVERTISING COURSE	

Change priority | More options

Social impact

- Providing action plans that are optimal for reducing the length of unemployment helps to break the **"unemployment spell"**.
- Providing a platform for **data-driven** approaches to understanding the pathways to employment and the impact of interventions.
- Enabling **smarter decisions** about which interventions to divert funding towards.

Next steps

- Implement **job-skills mappings** to intelligently guide people towards desired and feasible employment.
- Analyse current and desired employment **trends**.
- Enable a **human feedback** loop to optimise ai-empowered human decision making.