# JOBMATCH: SUPPORTING RECRUITERS THROUGH ALGORITHMIC HIRING

JOBMATCH CLOSING EVENT

FEBRUARY 28, 2024

- Jobmatch
  - Three-year Grand Solutions project with two post-docs (Mesut Kaya and Qiuchi Li)
  - Collaboration between University of Copenhagen, Aalborg University and Jobindex
- Goal is to support Jobindex' recruiters in their work





Mesut Kaya
Postdoc
AAU



Toine Bogers

Associate Professor

AAU/ITU



Qiuichi Li Postdoc KU



Sofie Vedtofte

Business Analyst

Jobindex



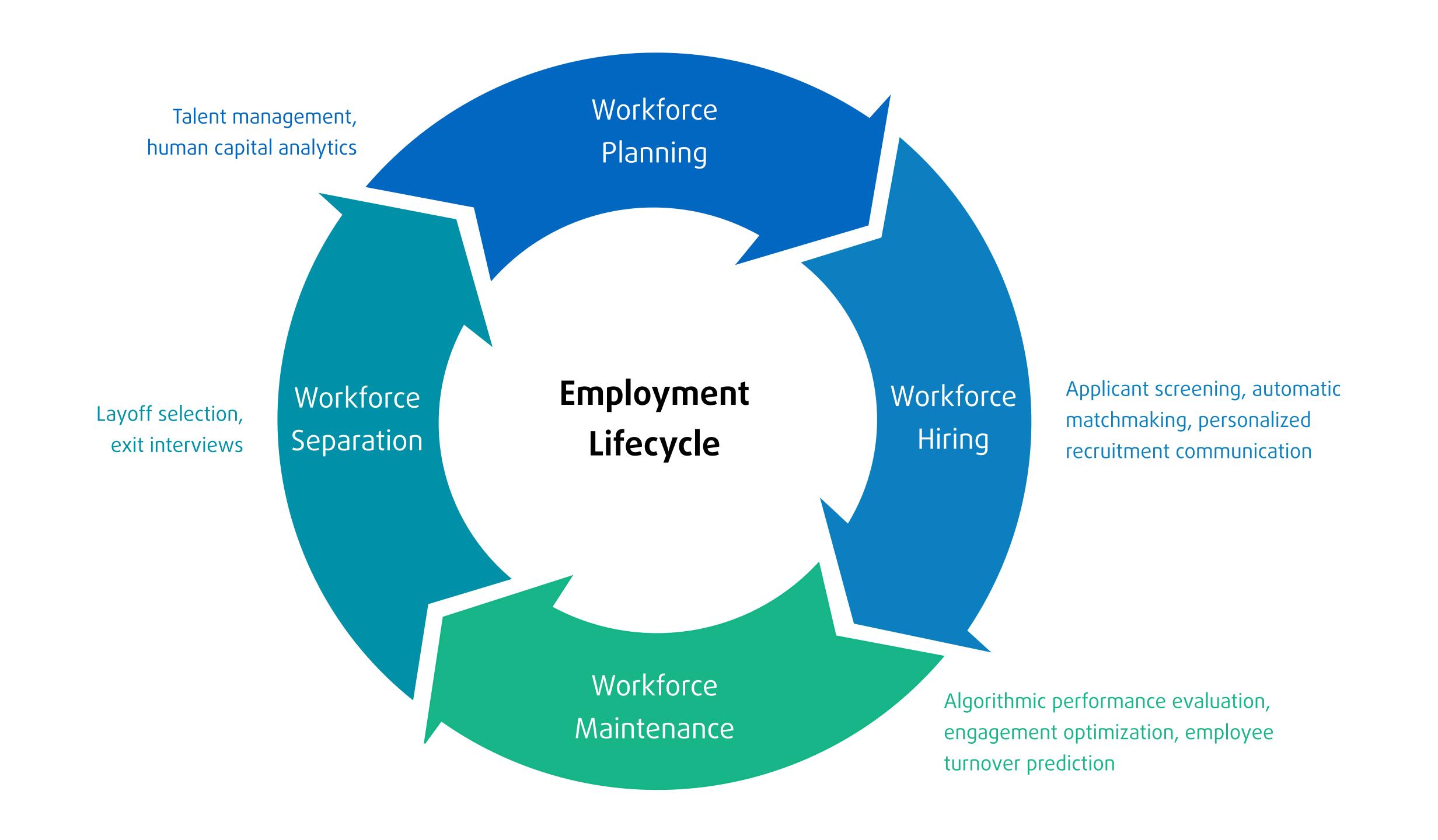
Carina Emma Sørensen
Team Leader
Jobindex

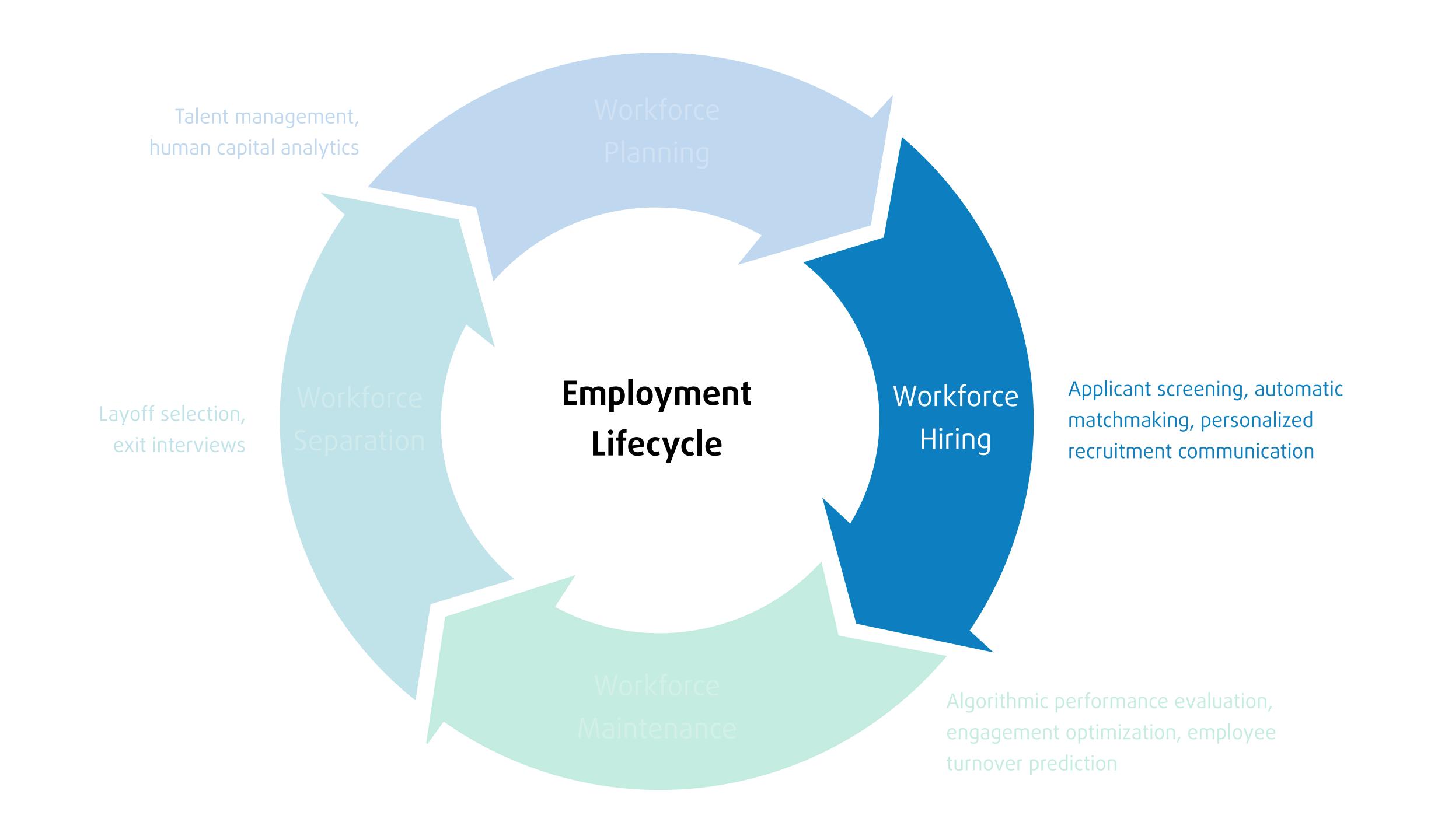


Klaus Madsen
CTO
Jobindex



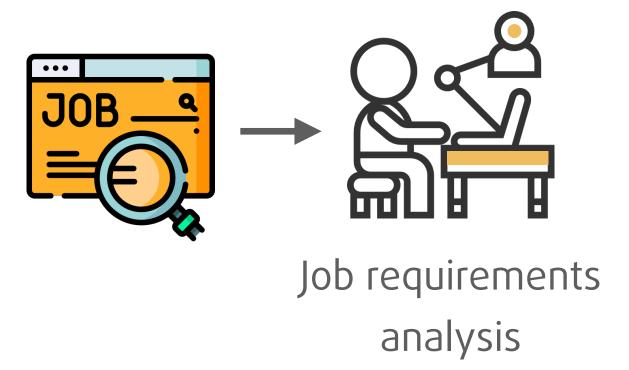
Christina Lioma
Professor
KU

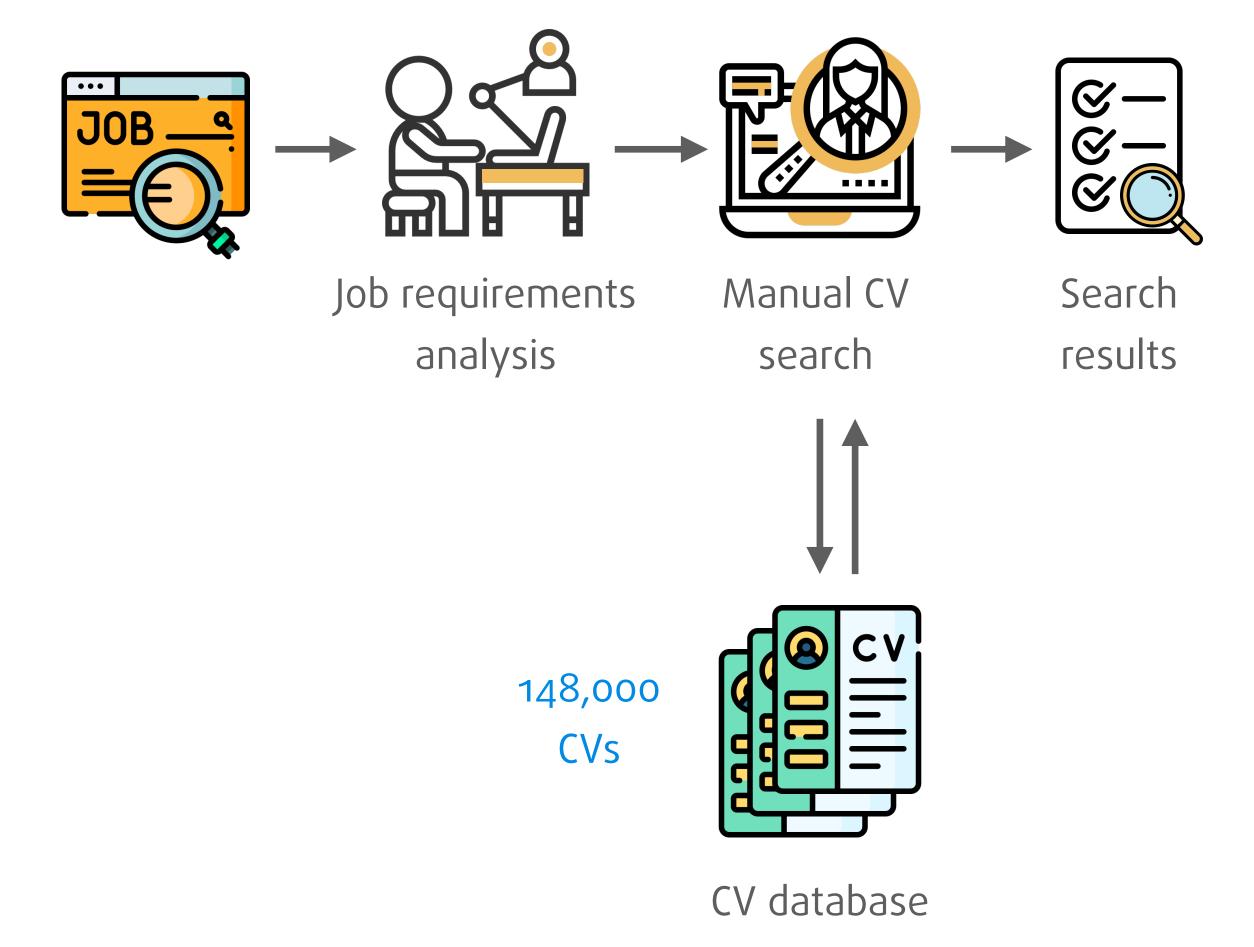


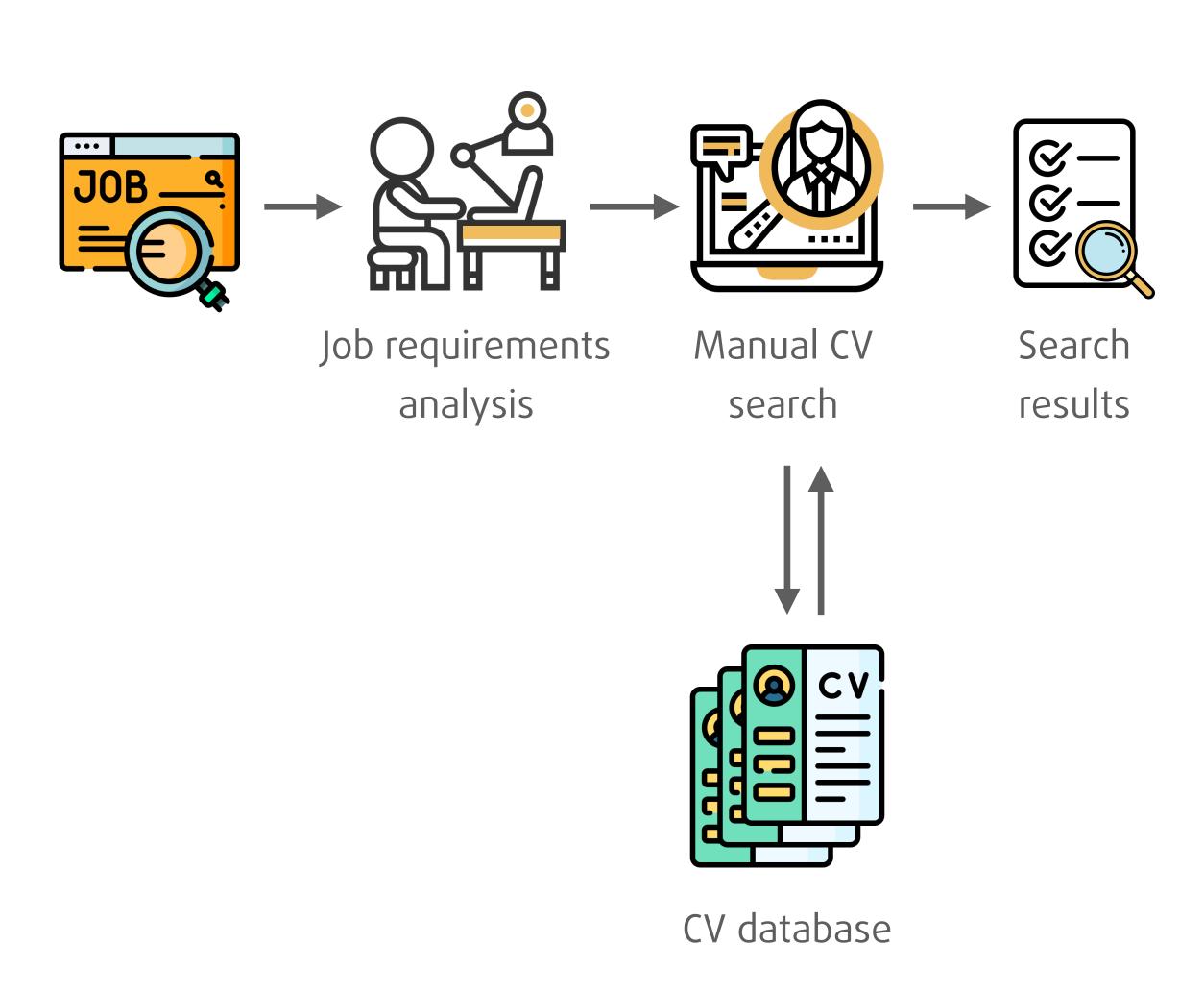


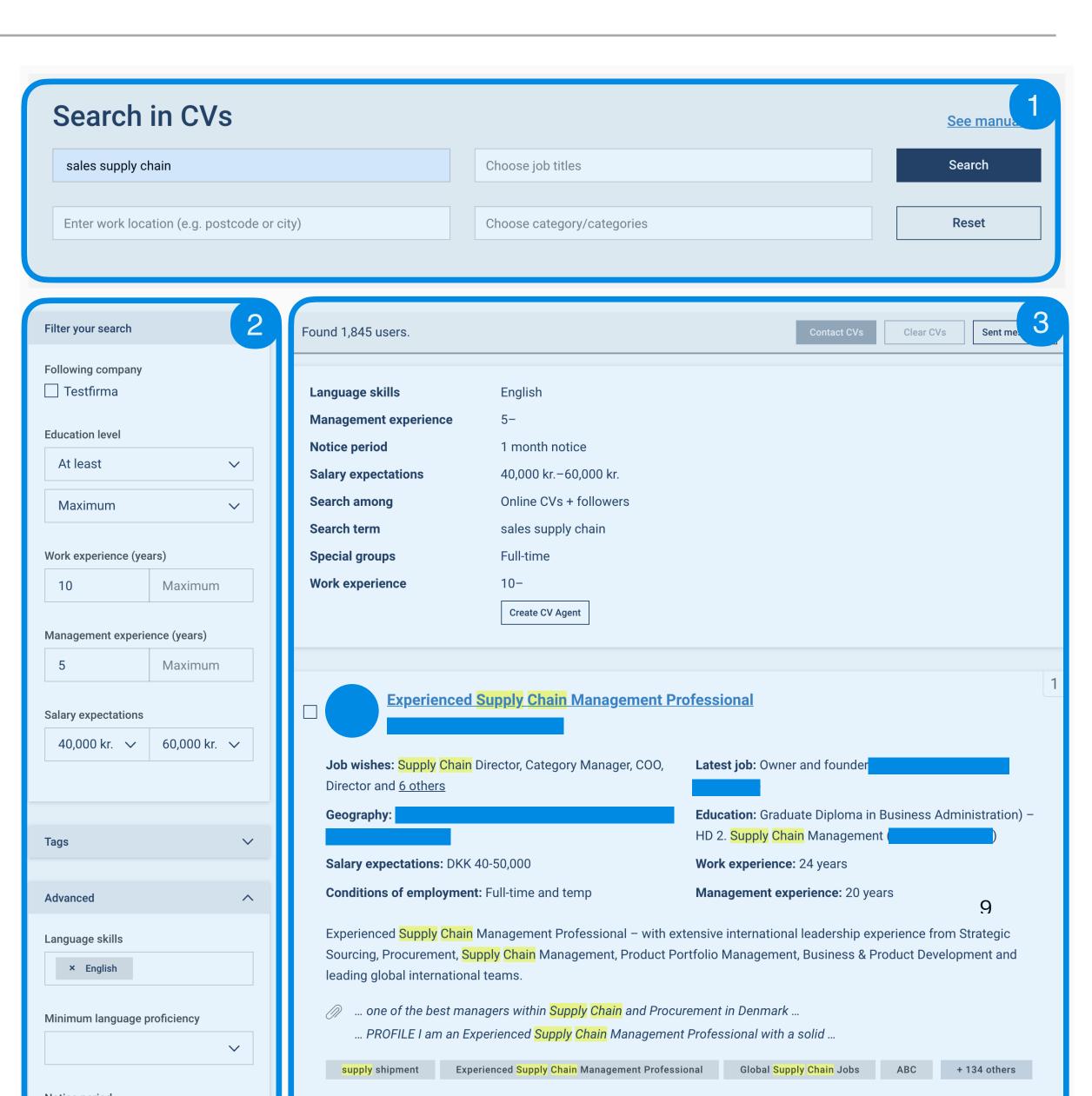
35,900 job ads

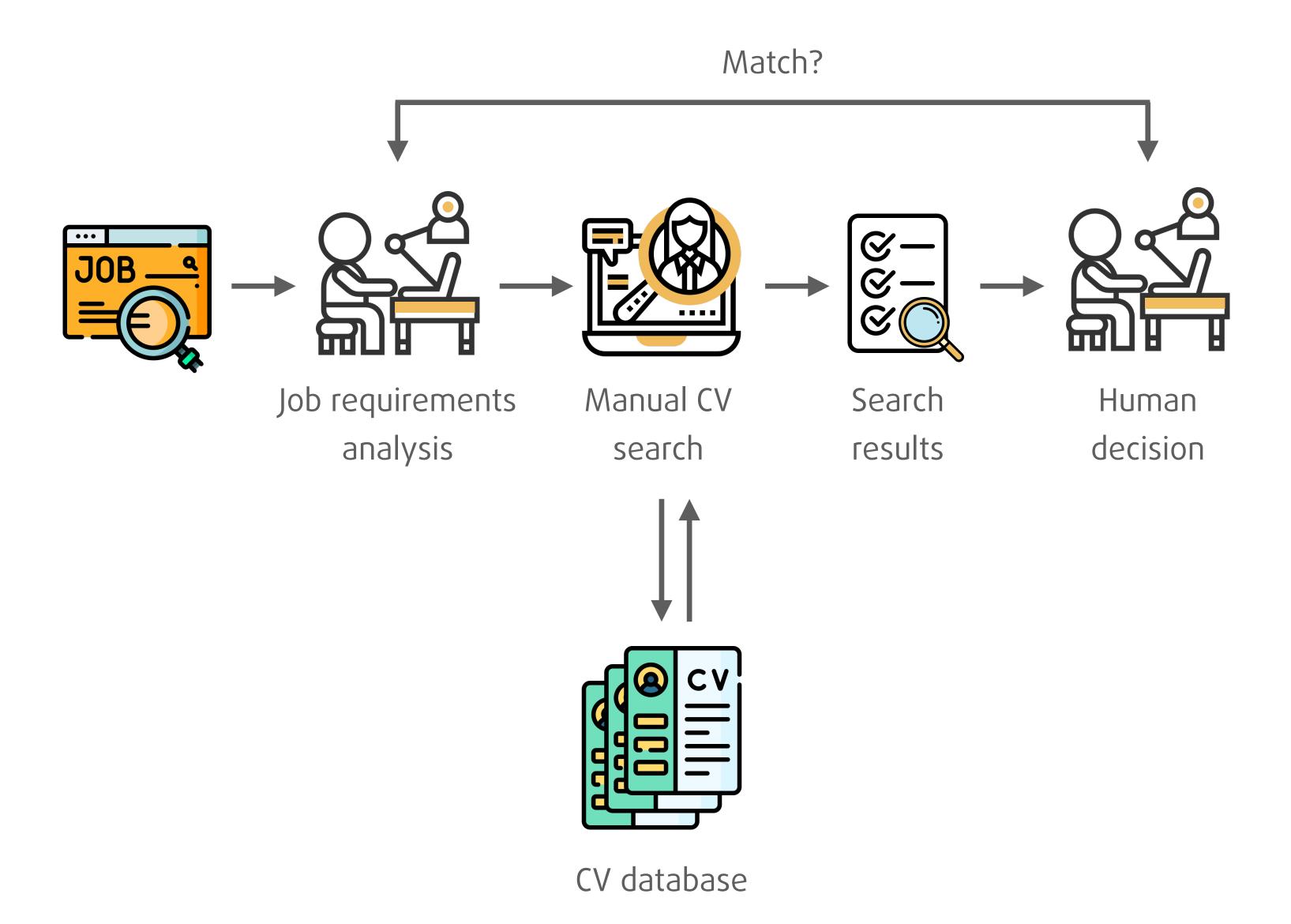


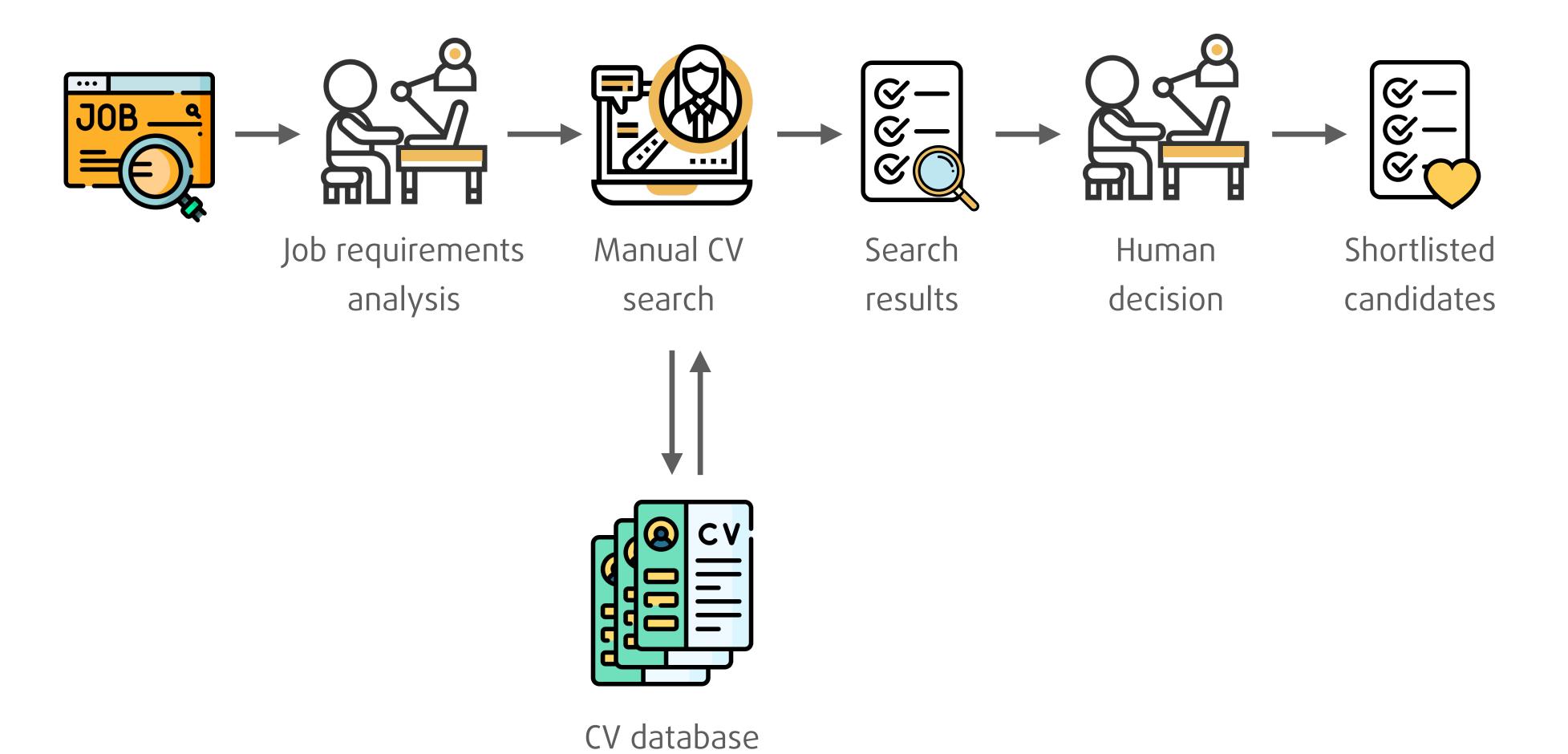


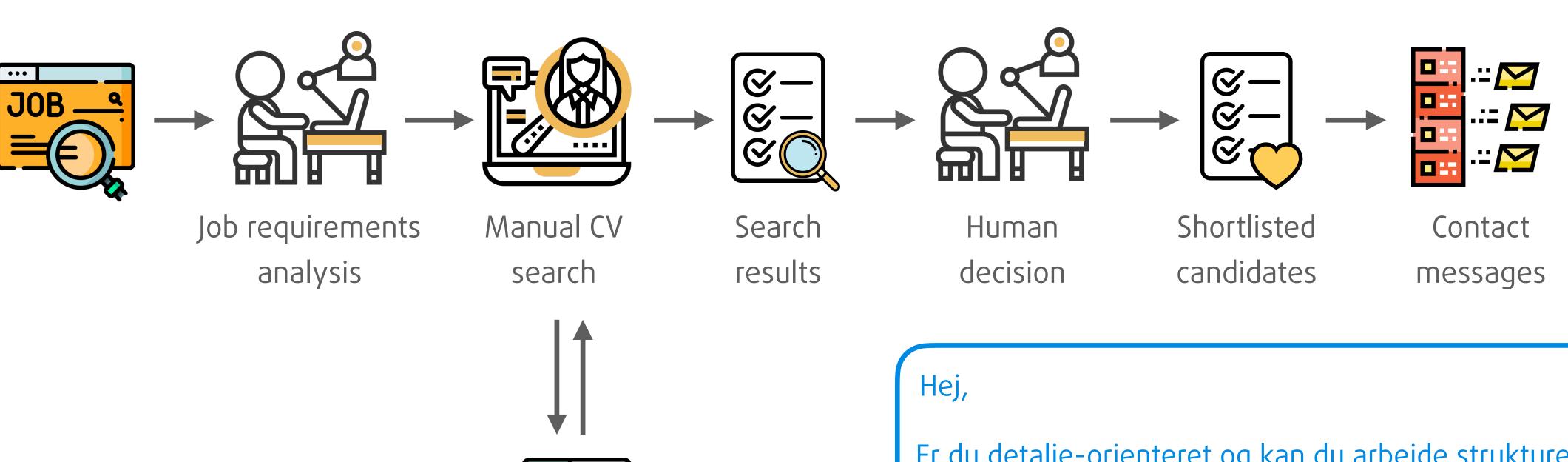






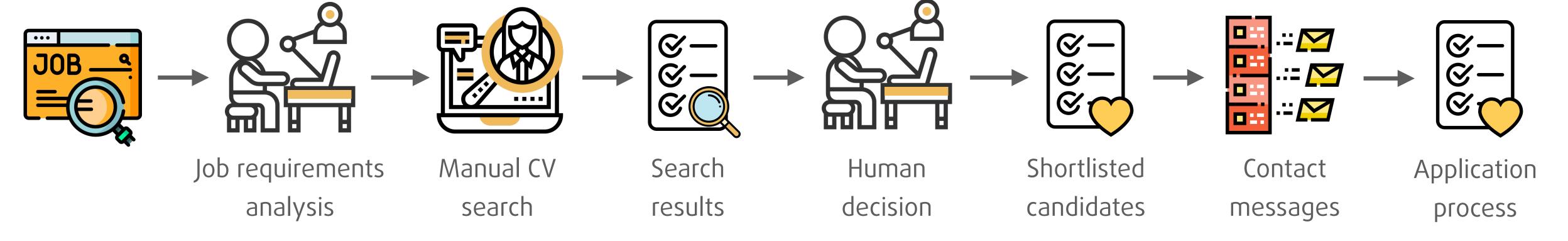


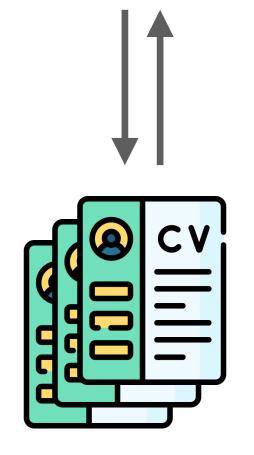




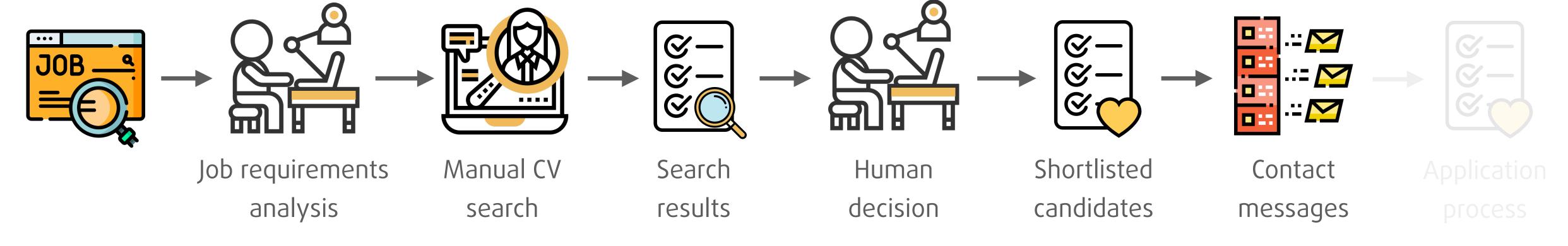
Er du detalje-orienteret og kan du arbejde struktureret og selvstændigt? Elsker du at hjælpe dine kollegaer med produktions- og supportopgaver samt øvrige interne opgaver? Er du til sund fornuft og handlekraft, så kan du nu blive kundemedarbejder til BackOffice hos Sparekassen Danmark A/S.

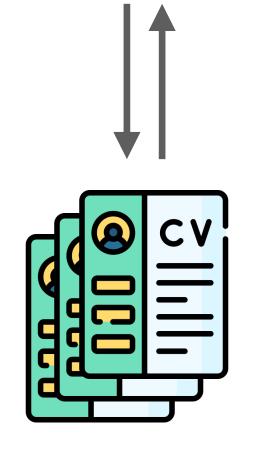
CV database



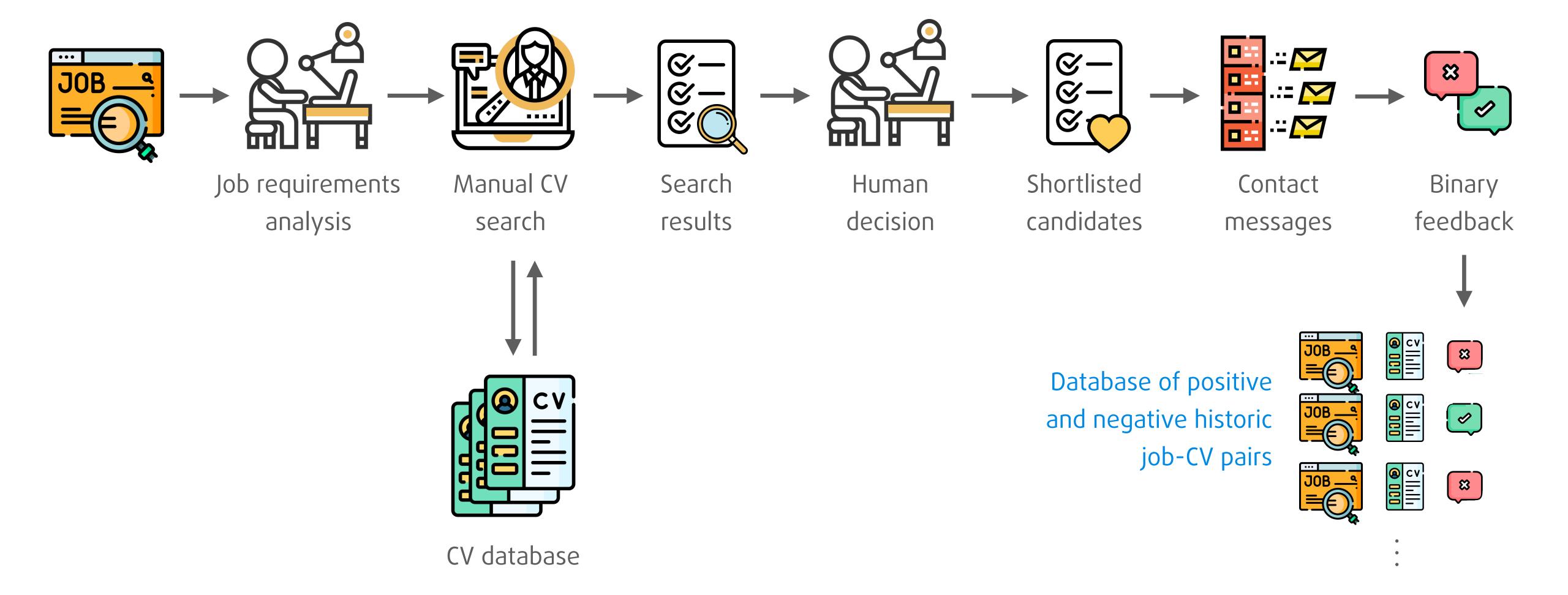


CV database



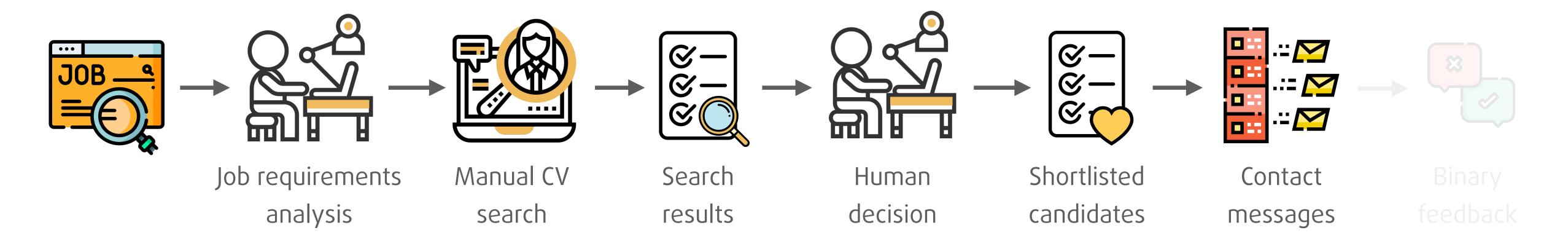


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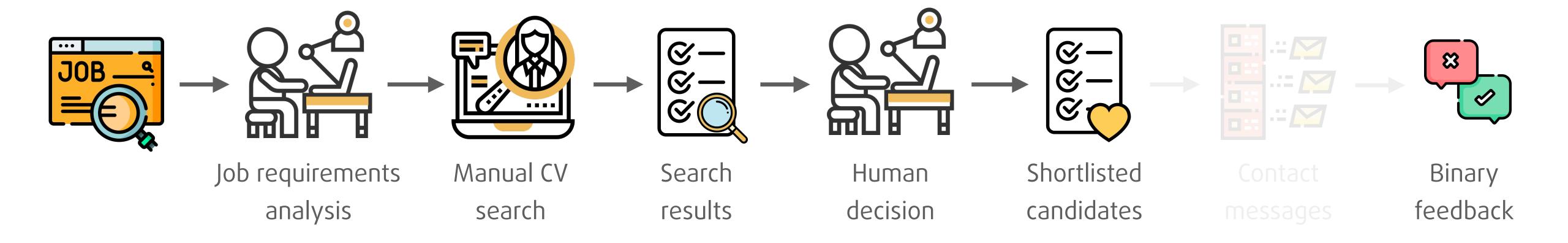
#### UNDERSTANDING RECRUITERS: CONTEXTUAL INQUIRY

- Supporting the recruiters means understanding how they work
  - Qualitative study of recruiters' work processes
    - 12 contextual inquiry sessions (≈ mix of observation and interview)
    - 1 focus group
  - Quantitative study of recruiters' search logs
    - Analysis of 157,046 search sessions (queries, contacted candidates, responses)



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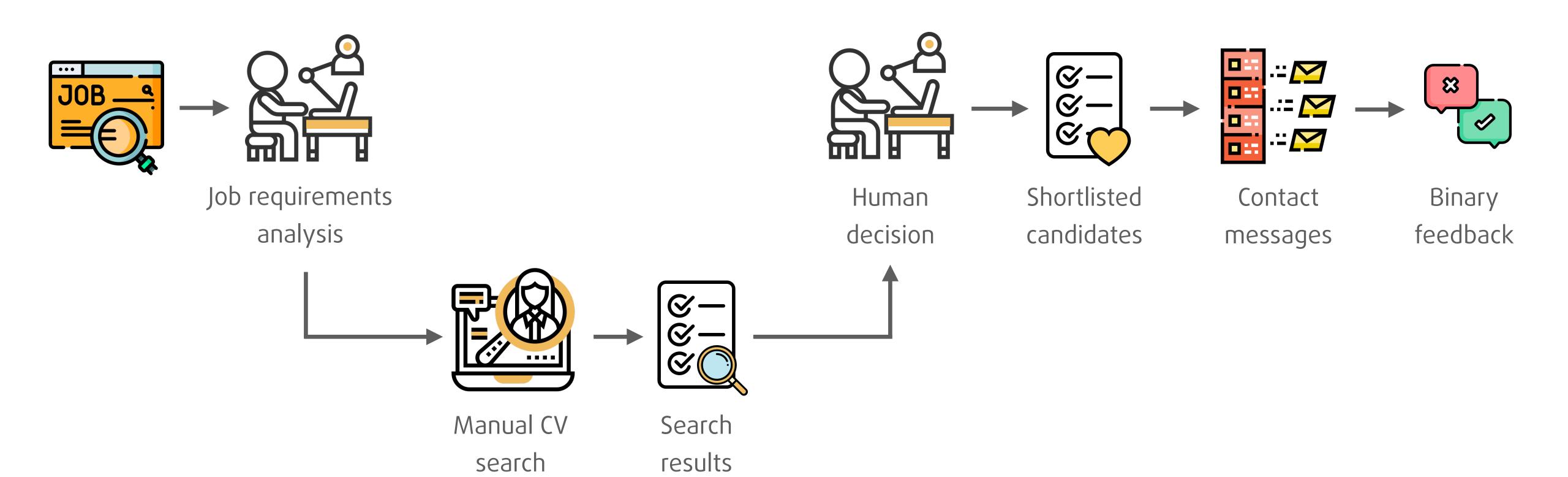


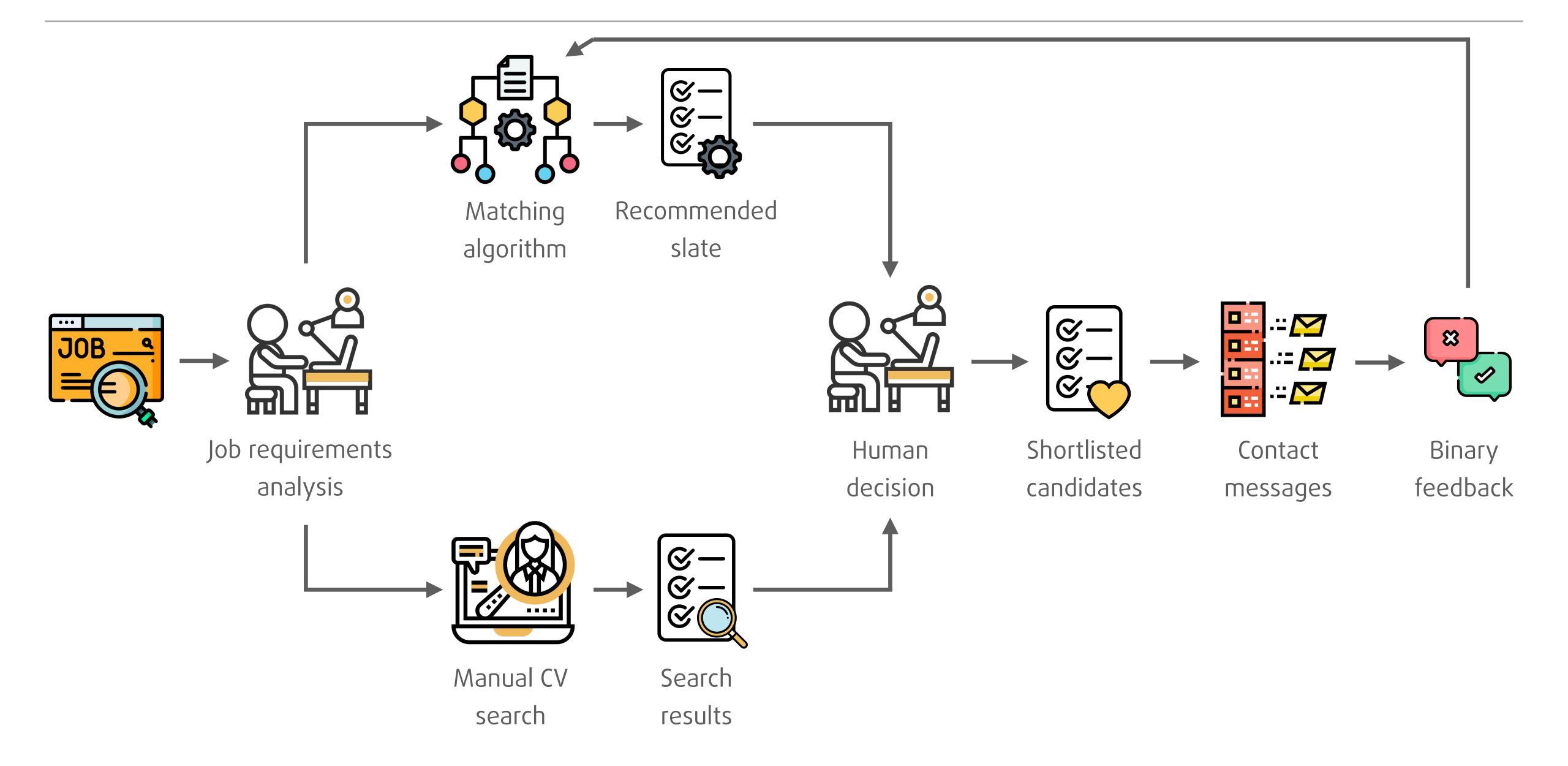
#### UNDERSTANDING RECRUITERS: FINDINGS

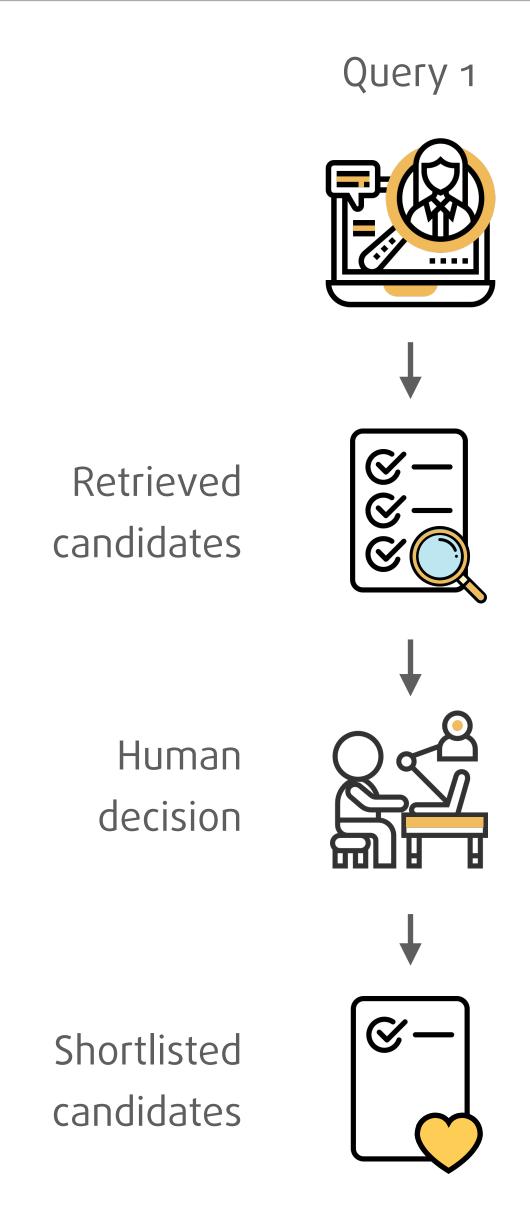
#### Findings

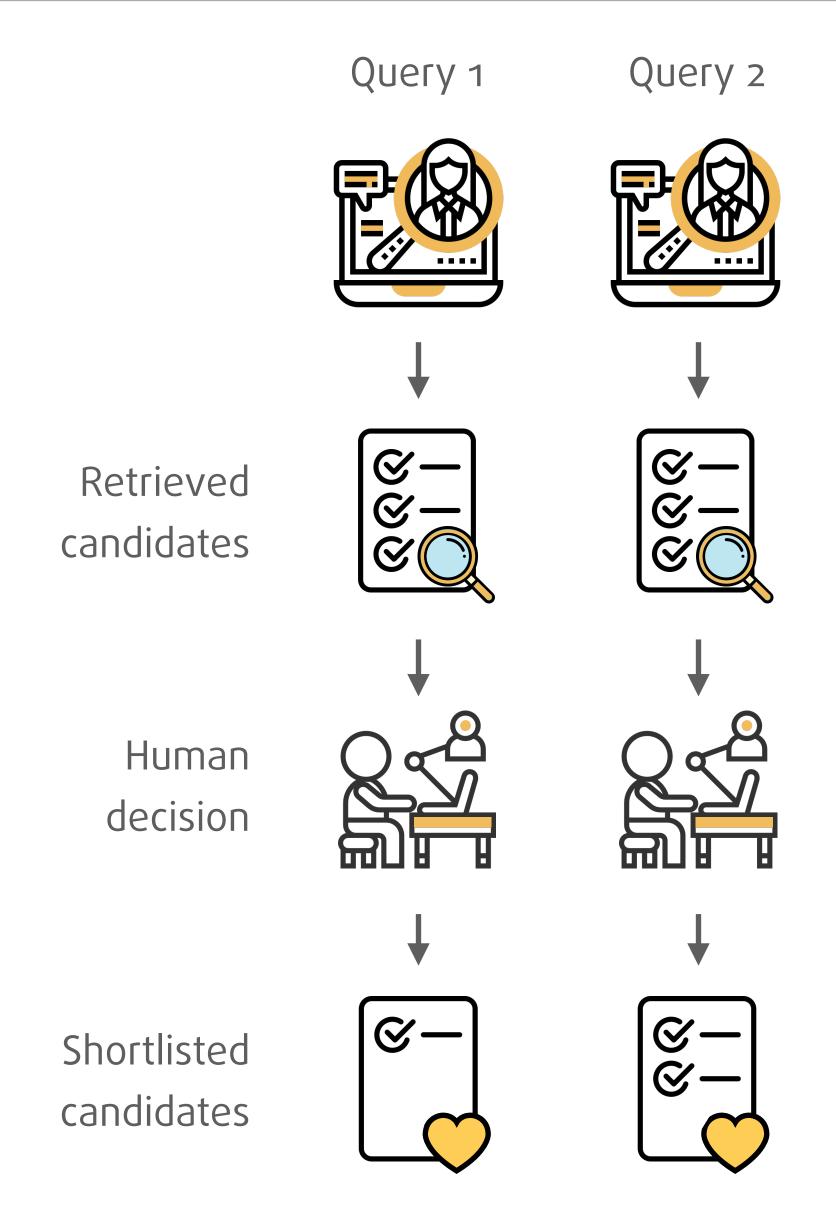
- Recruiters focus on primary qualifications (= skills & competences) and past/current job titles
- Search results can serve as source of new search terms
- Commonly used filters include salary and years of work/management experience
- Relevance assessed using 1-3 most recent work experiences and matches on qualifications
- Strong reliance on highlighting in the results list
- First 10 candidates are low-hanging fruit, next 10 candidates require more time and effort
- Time per query, use of filters and query complexity all increase throughout the sessions
- Caveat: no clickstream data  $\rightarrow$  we are collecting a richer search log for future work

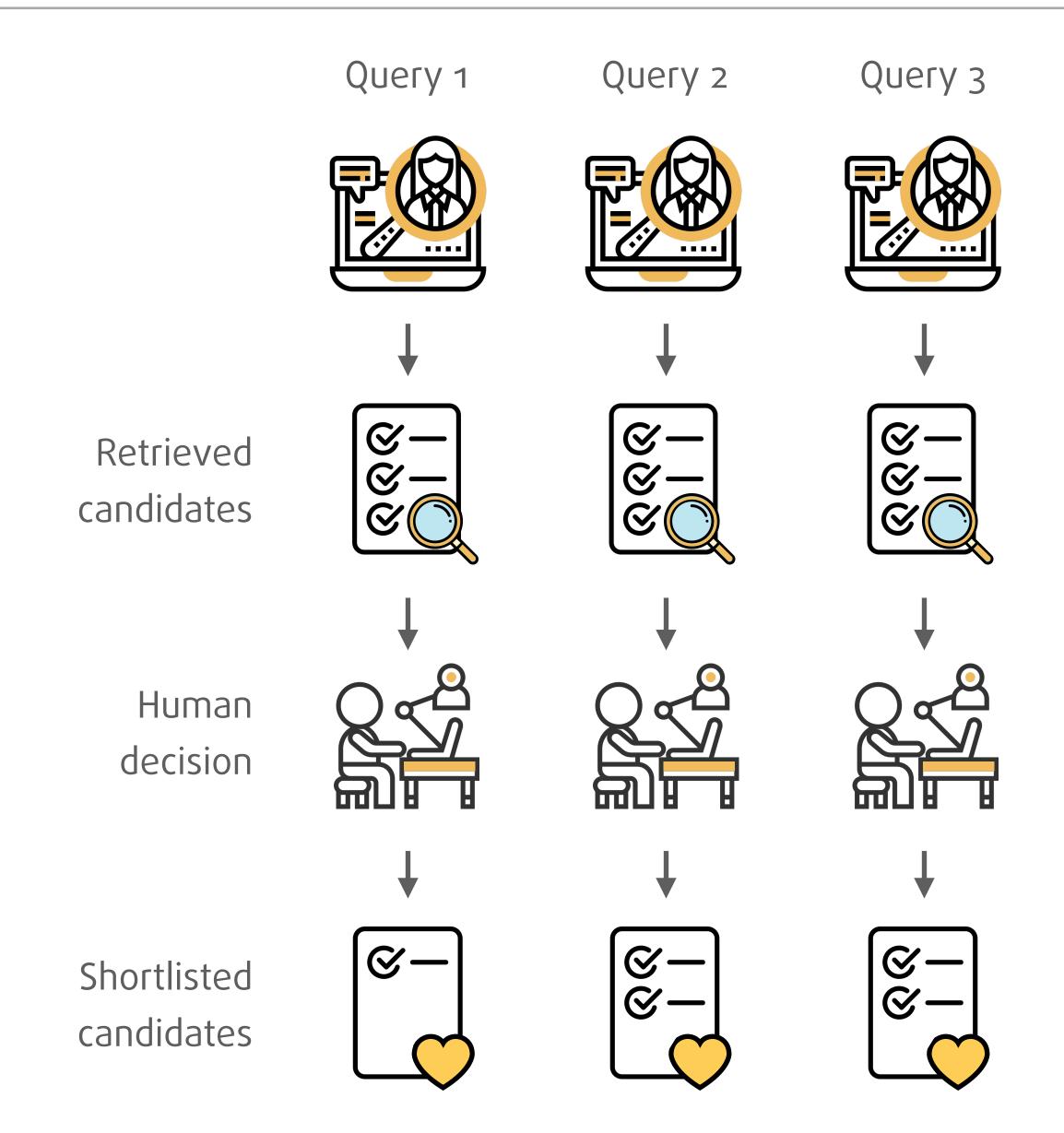
### Part 2

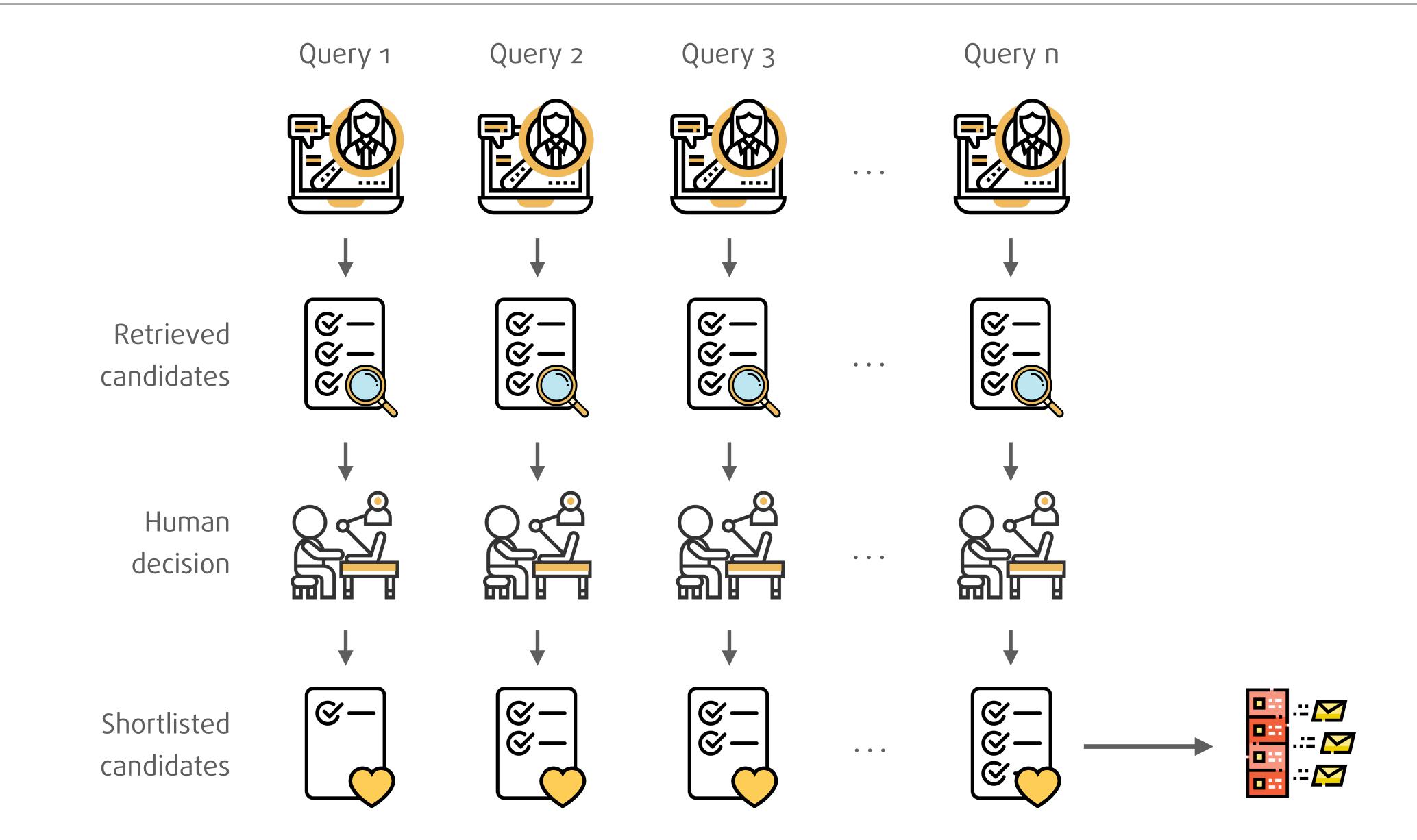








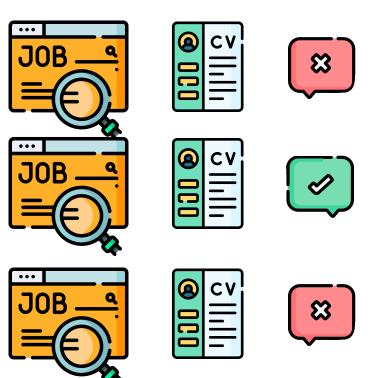




Query o Query 1 Query 2 Query 3 Query n Recommended/ retrieved candidates Human decision Shortlisted • • • candidates

#### CANDIDATE RECOMMENDATION :: EXPERIMENTS

- Goal is to support recruiters with recommendations instead of replacing them
  - Relevant recommendations should result in **shorter sessions** and more efficient recruiters
- Algorithm development
  - **Collaborative filtering** baselines  $\rightarrow$  using past interactions between job seekers and job ads ("People who applied for job X also applied for this job")
    - Item-based k-Nearest Neighbor
    - Bayesian Personalized Ranking
    - Problems: cold start (each job is new!) and data sparsity (few interactions!)



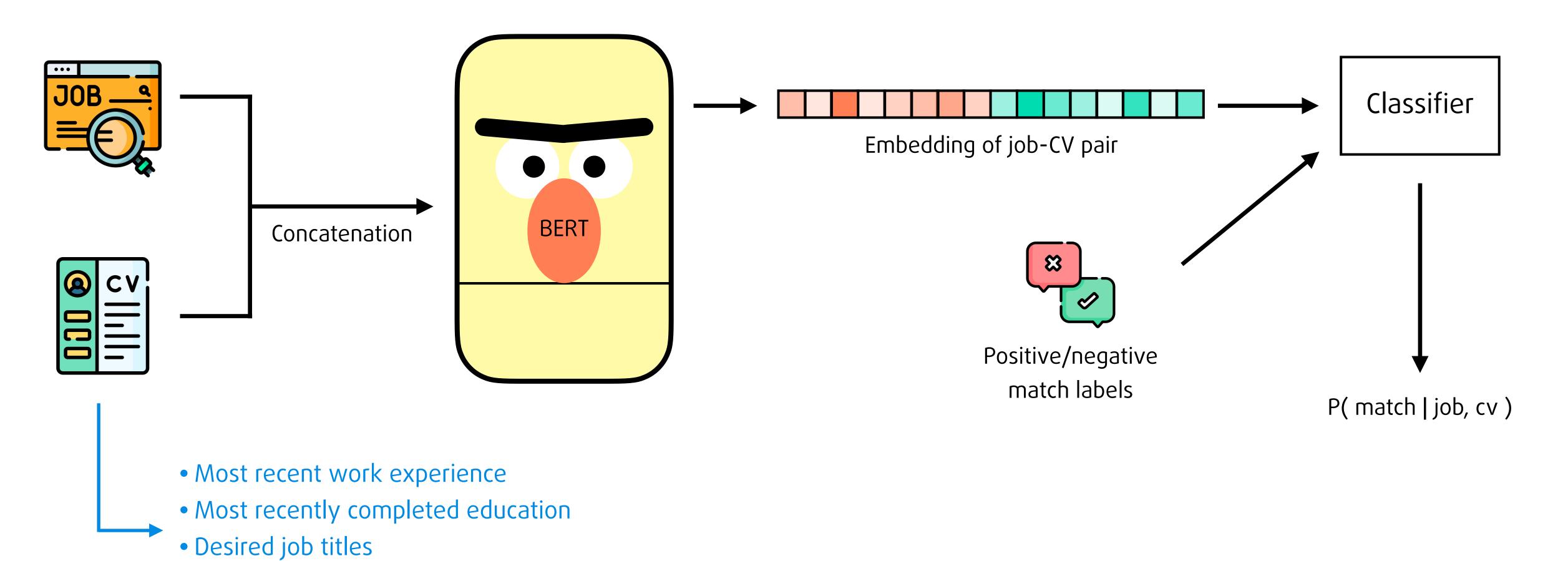
#### CANDIDATE RECOMMENDATION :: EXPERIMENTS

- ightharpoonup Content-based filtering ightharpoonup using semantic representations of jobs and CVs to match them
  - Generated embeddings of job ads and CVs using wordzvec, doczvec, and (Danish) BERT
    - Pre-trained as well as fine-tuned on ~260,000 CVs and ~426,000 job ads
    - Standard matching between job ad and CV embeddings using FAISS library
  - Development iterations
    - Iteration 1: Full-text embeddings vs. embedding only job titles
    - Iteration 2: Hybrid algorithm that incorporates the interaction history embedding of job titles only
      - \* A CV is represented by the average of all job (embeddings) they interacted with
      - \* A job is represented by the average of all CV (embeddings) that interacted with it (cold start issue)

#### **CANDIDATE RECOMMENDATION:: EXPERIMENTS**

Keywords (may include skills)

• Iteration 3: Matching CVs and jobs using sentence-pair classification

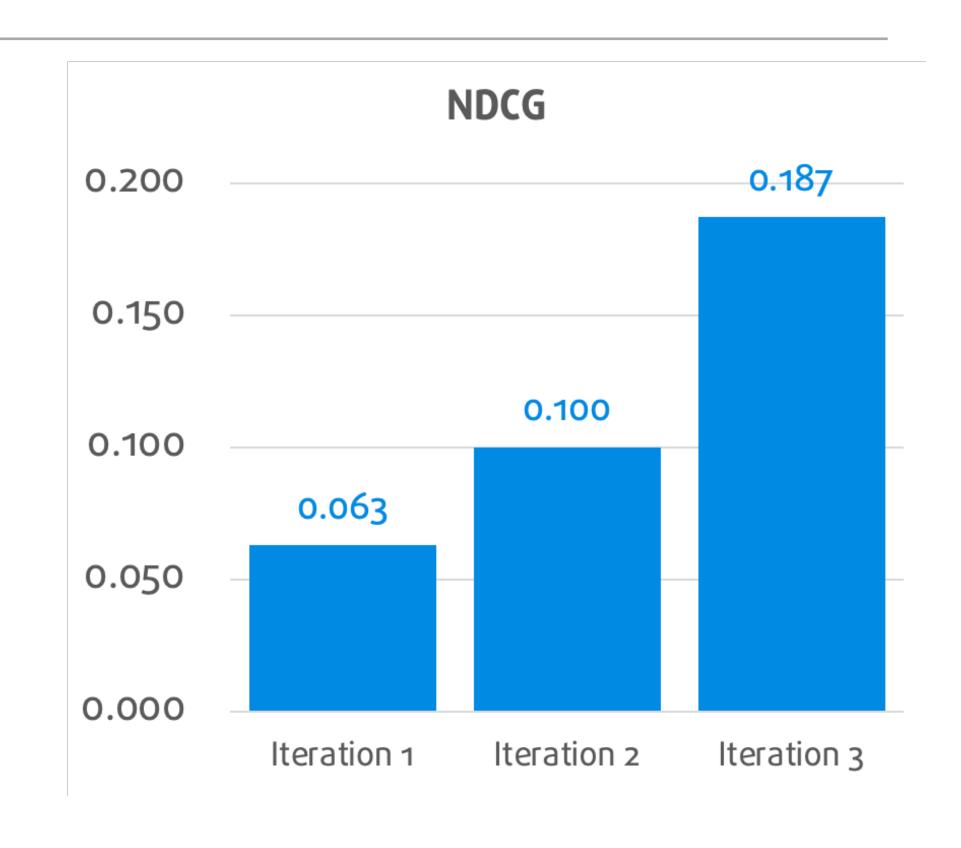


#### **CANDIDATE RECOMMENDATION:: FINDINGS**

- Offline evaluation
  - Iteration 3 outperforms the existing prototypes by a wide margin
- Online evaluation (of iteration 3, April-Nov 2023)
  - 11.5% higher response rate when viewing the list of recommendations



- Around 8 out of 23 candidates to be contacted are in the list of recommendations
- Lots of effort went into implementing A/B testing infrastructure at Jobindex

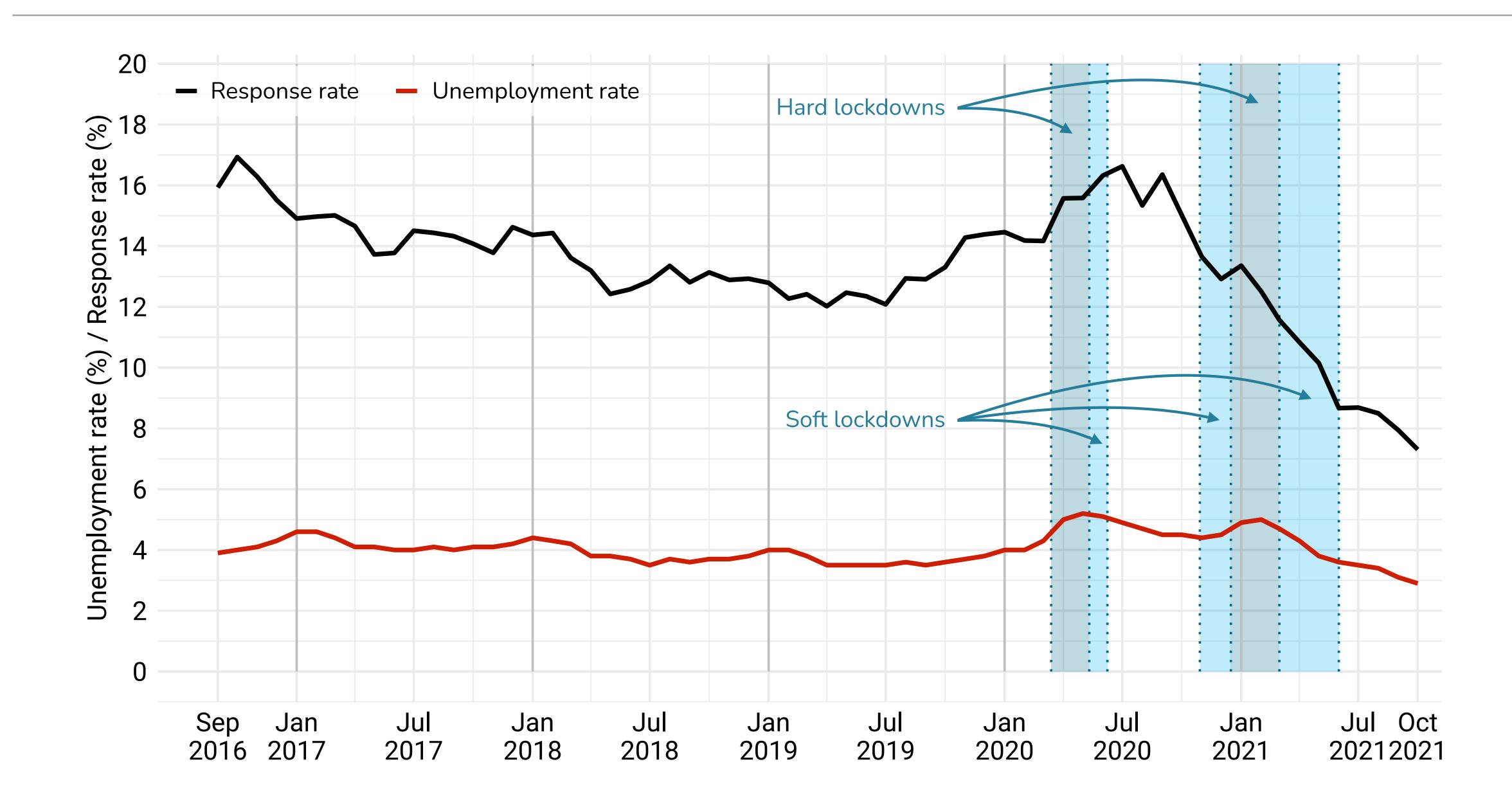


#### **CANDIDATE RECOMMENDATION:: FINDINGS**

#### Lessons learned

- Habits can be hard to break (~40% ignore the recommendations)
- Communication and involvement is key! (fear of being replaced vs. lack of trust)
- Check Double-check Triple-check for bugs!
- Pre-filtering the candidate set can be used to offset costly sentence-pair classification
- Important to coordinate A/B tests with other departments
- Success partially depends on external factors

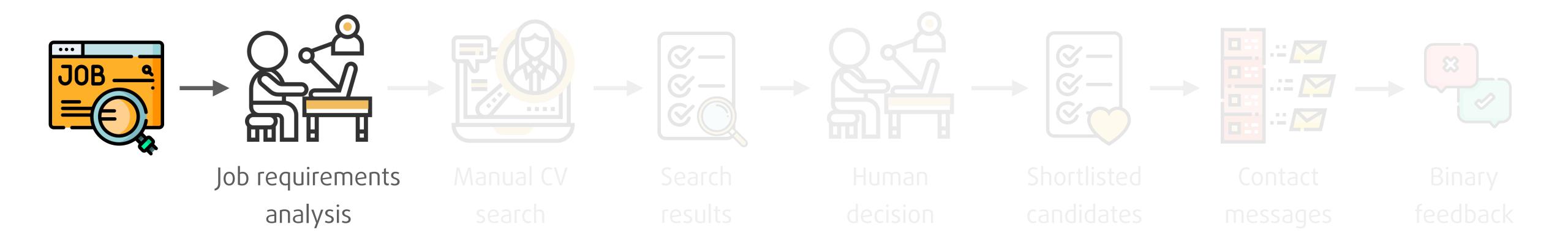
#### **CANDIDATE RECOMMENDATION:: FINDINGS**



## Part 3

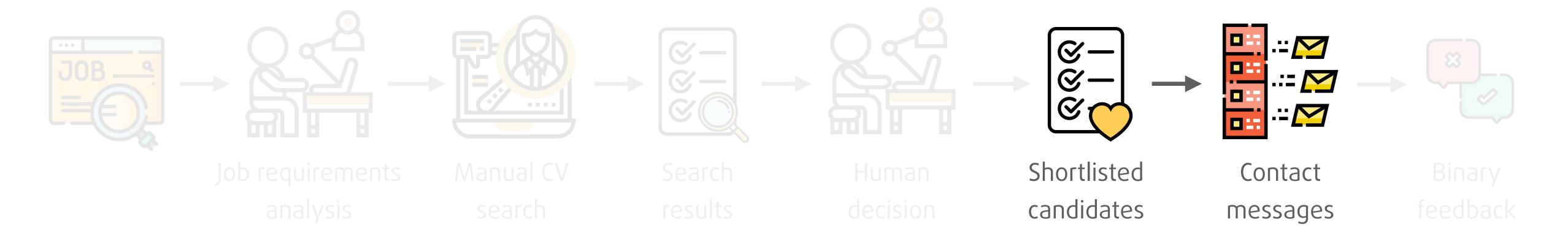
#### SKILL EXTRACTION

- Work by Qiuchi Li on skill extraction from job ads and CVs
  - Li, Q., & Lioma, C. (2023). **Joint Extraction and Classification of Danish Competences for Job Matching**. In European Conference on Information Retrieval (pp. 475-483). Springer.
- Future work
  - Distinguishing between primary and secondary skills
  - Renewed effort on integrating this in the recommendation algorithm



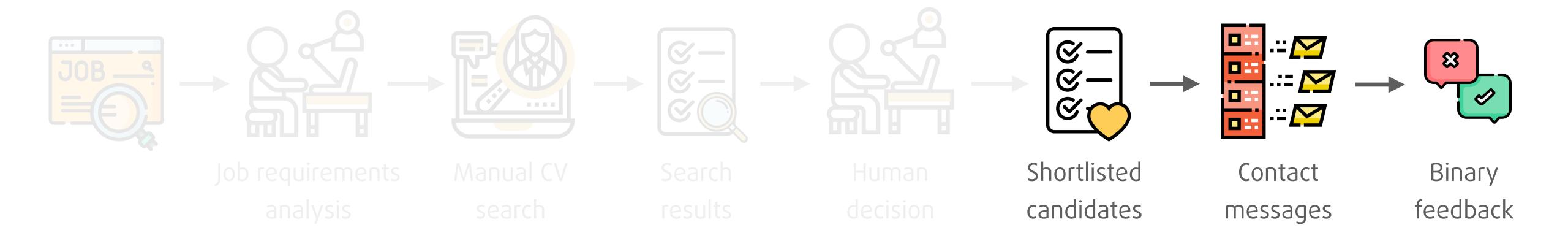
#### NEURAL MESSAGE GENERATION

- ▶ Recruiters need to send contact emails to matched candidates asking them to apply for the job
  - → currently, all candidates receive the same message
  - Goal is to automatically generate personalized contact emails!
  - Recruiters typically follow template-filling practices when writing contact emails, so the challenge is
    to automatically generate candidate-specific text fields
  - Core challenge is the generation of a motivational sentence, which should convince the candidate to apply



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- Incorporating other features
  - Full-time vs. part-time, renewed look at integration of extracted skills
- Further integration between recommendation and message generation
- Multi-sided fairness for algorithmic hiring
  - Multi-stakeholder approach to fairness in algorithmic hiring
  - Development of fairness dashboard for Jobindex
  - Fairness-aware candidate and job recommendation algorithms
- Career path prediction
- Generative Al support for job seekers



# THANKS FOR YOUR ATTENTION!