The LEHD Infrastructure Files

and the Creation of the Quarterly Workforce Indicators

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Introduction

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  - Innovative use of wage records to constitute a frame to integrate data
  - The first statistical system to use “jobs” as a frame

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- Describe the disclosure-proofing mechanism
- Describe researcher access to infrastructure files and confidential QWI files

Input Files

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- in particular, for Minnesota, the ESTABLISHMENT is reported

Employer reports: ES202

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- Where available, more detailed data on individuals is also extracted from surveys and censuses:
  - CPS
  - SIPP
  - ACS
  - 1990 Census
  - 2000 Census

Infrastructure Files

EHF: Employment History Files

- Job-level EHF
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- Employer and establishment-level employment history
  - QCEW-based employment-activity history for every SEIN (employer) and SEINUNIT (establishment)
- Comparison of employment and activity of SEINs between UI and QCEW files is done for QA purposes, and in preparation of weighting.

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  3. unduplicated (by firm name)

Flow so far

Main page

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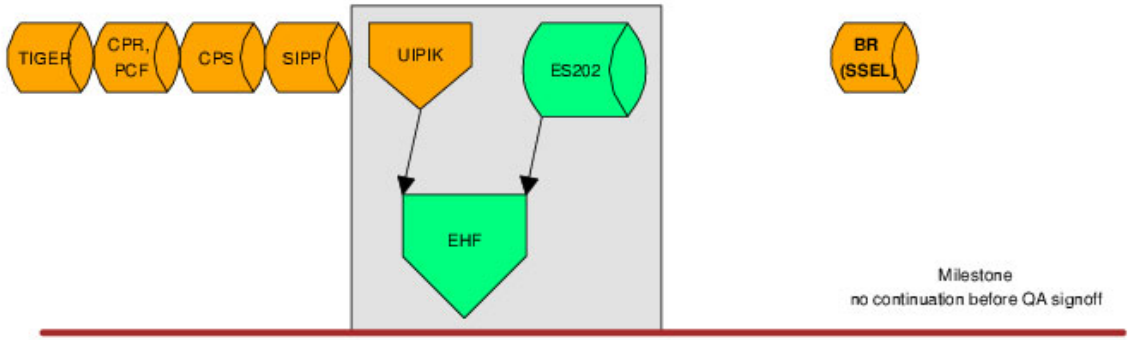

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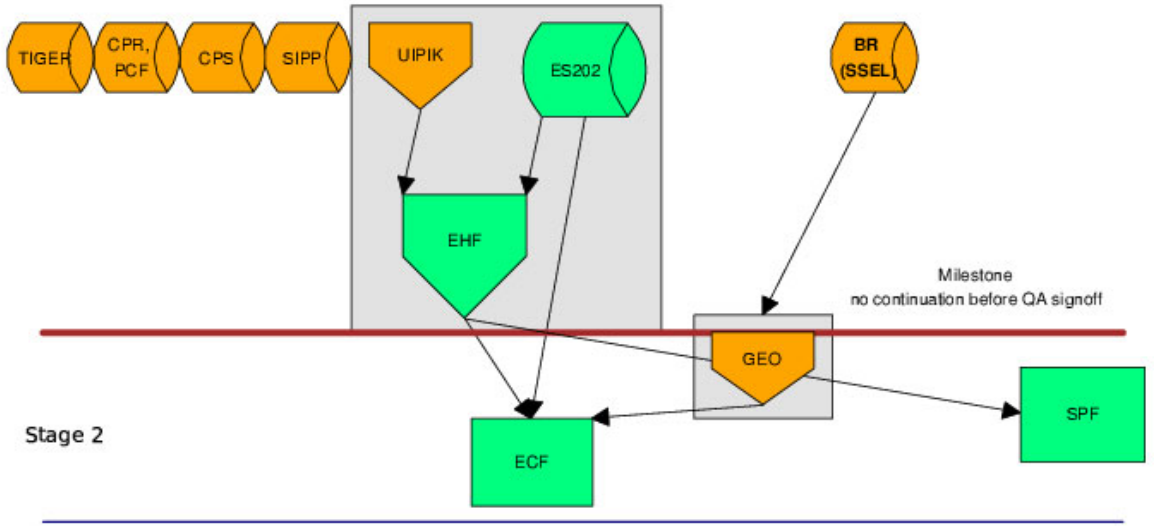

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Forming Aggregated Estimates: QWI

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- $\longrightarrow$ non-economic change in identifier creates spurious flow

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- QWI: if 80% of an SEIN's workers (the predecessor) are observed to move to a single successor, and that successor absorbs 80% of its employees from a single predecessor, then all flows between those two account numbers are filtered out, and treated as if they had never existed.

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  - Several million imputations every quarter

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α_{jrt} establishment- and quarter-specific effect

x_{ijrt} time-varying vector, worker and establishment

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Disclosure-proofing the QWI

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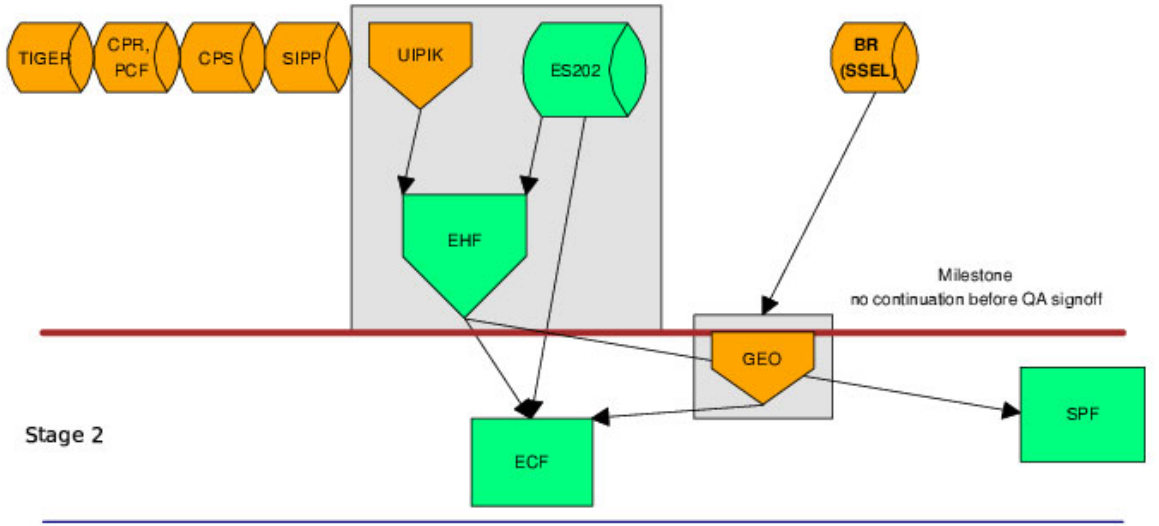
Conclusion

Flow so far



Main page

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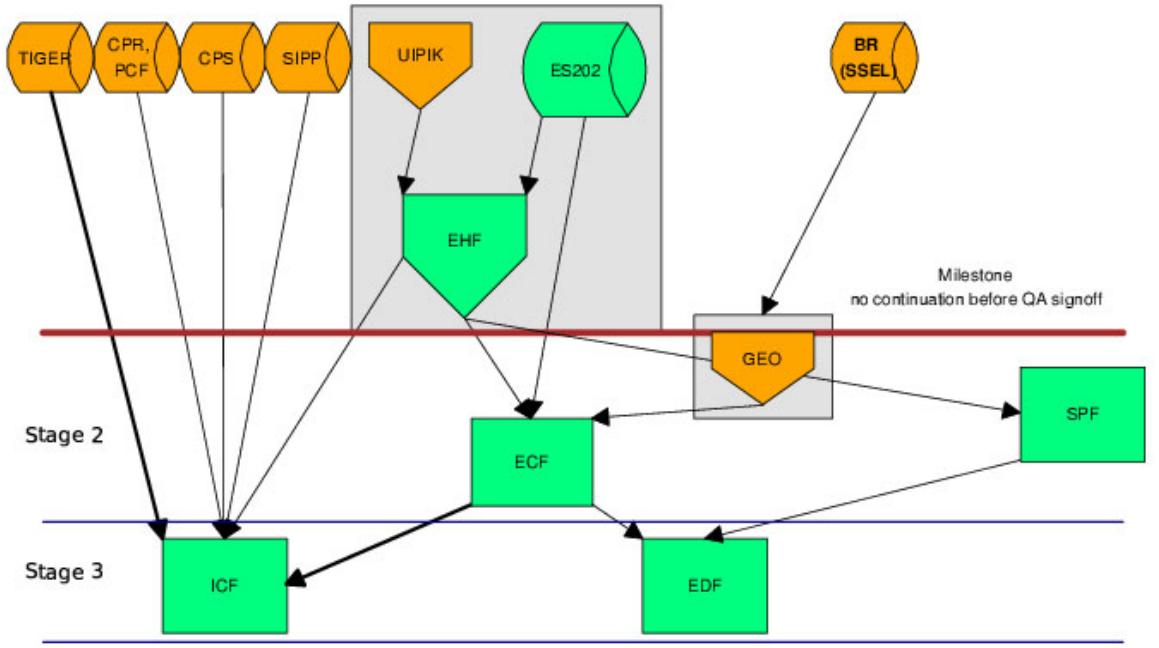

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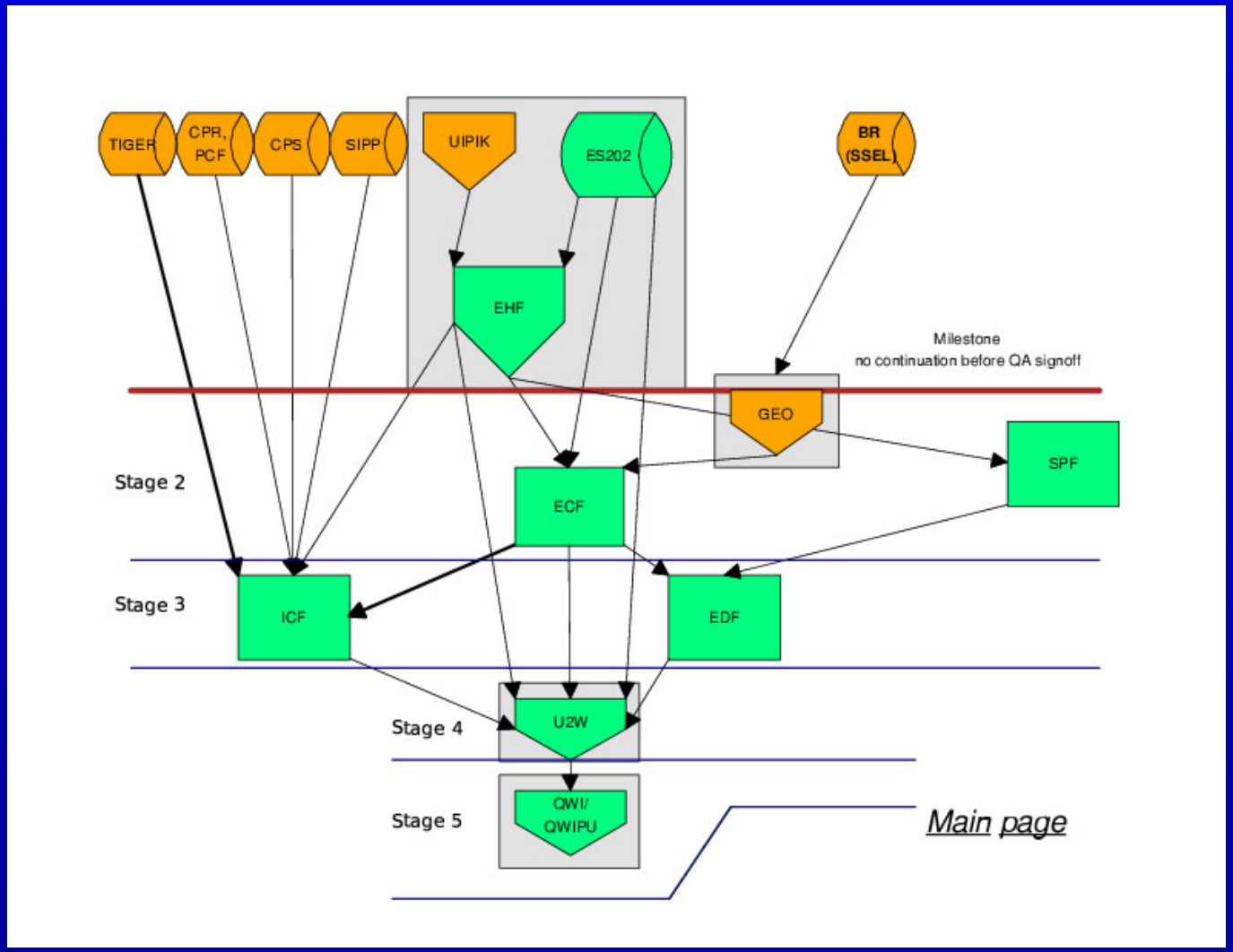

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