



# Analyzing Warehouse Productivity



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# Background and Significance

**Supplement work done by the Warehouse Work, Health, and Well-Being study conducted under Professor Erin Kelly at Sloan**

Purposes of study:

- Understand what it's like to work at a fulfillment center for an online retailer
- Identify how changes in the workplace affect workers' experiences
- Investigate how work conditions may affect workers' decisions to stay or leave job

Team aimed to build on study by finding relationships between turnover and warehouse productivity

# Objective

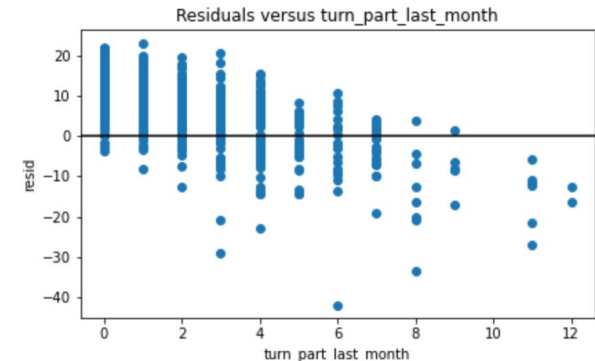
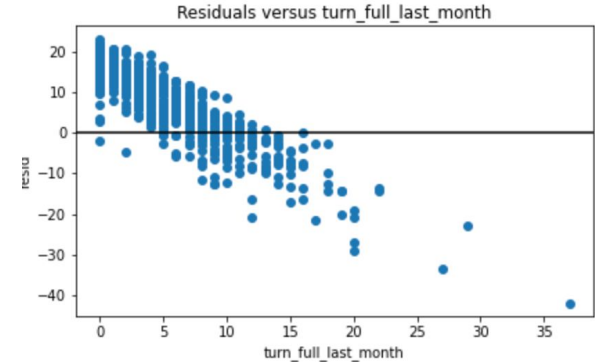
- Support the Warehouse Work, Health, and Well-Being study by studying the effect of turnover on productivity within warehouses of a national online retailer
- Finding an optimal level of turnover that maximizes productivity
  - Have data from an E-commerce retailer as well as monthly turnover data for the retailer's warehouses, which corresponds to how many workers leave each fulfillment center

# Data

- Data on labor productivity, taken weekly, for each fulfillment center (FC) between 2011 and 2020
- Monthly data on turnover for each FC between 2017 and 2020
  - Aggregated two datasets
- Final dataset has 36 features
  - **Turn\_full\_last\_month:** turnover of full-time employees from previous month
  - **Boxes\_per\_hr:** boxes moved per hour, primary measure of productivity
  - **Ot\_hrs:** total overtime hours worked
  - **Tot\_sales:** total sales

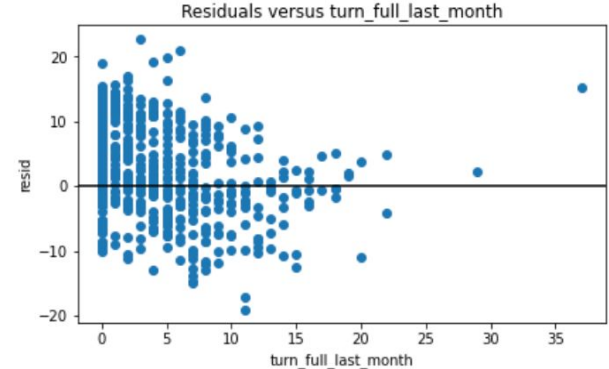
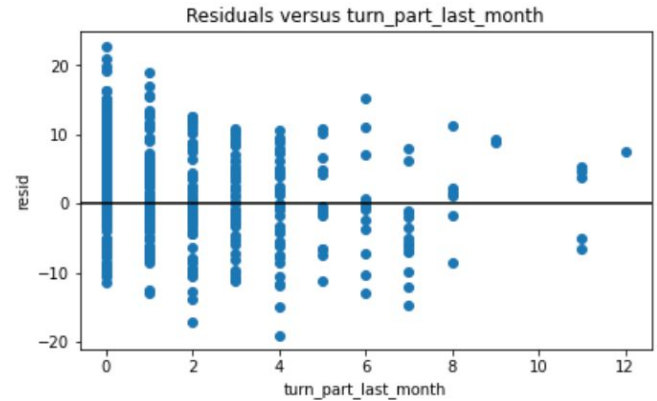
# Exploratory Data Analysis

- Plotted turnover (FT (full time) and PT (part time)) versus productivity (boxes\_per\_hr).
  - No clear correlation
- Simple regression of productivity on FT and PT turnover.
  - $R^2$  value of 0.394, probably due to outliers
- Removed outliers and ran regression again.
  - The  $R^2$  was still quite low.
  - Noted non-linear relationship with FT/PT turnover since the residuals not randomly distributed
- Introduced features  $\text{turnoverFT}^2$  and  $\text{turnoverPT}^2$ 
  - Residual plots looked better, decided to introduce more features



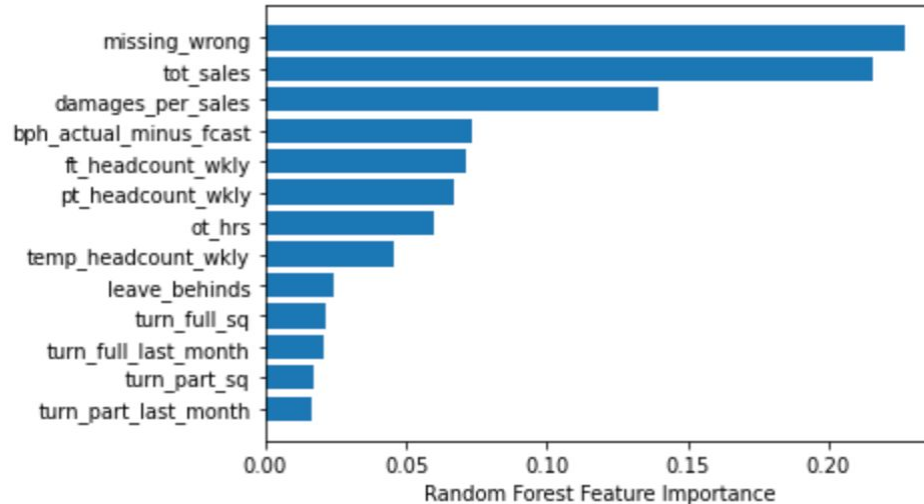
# Model 1: Linear Regression

- **Response variable:** productivity (boxes\_per\_hr)
- **Explanatory variables:** full time turnover, part time turnover, turnover\_FT^2 and turnover\_PT^2.
- **70%** of the data in training and **30%** of the data in testing and build a model on **training**
- Features with the highest significance are **turnover variables, bph\_actual\_minus\_fcast, ft\_headcount\_wkly, ot\_hrs**
- Residuals are much more randomly distributed in this model



## Model 2: Random Forest

- Random forest regression using 30% train/test split
  - $R^2$  of 0.953
  - MSE of 1.935
- Feature importances rank turnover variables as least important



# OLS vs. RFR

- OLS has more interpretable results, but it is not as accurate
- RFR doesn't know how to extrapolate outside of the data set, which could be problematic
- RFR probably performs better since there isn't a linear relationship between turnover and productivity and it is able to identify nonlinear relationships
  - Risk that it could be overfitting
- If we wanted a more accurate model that is able to extrapolate, we could train a neural network
  - Ideally done with more data



# Model 3: Optimization

- Aimed to maximize the productivity for a given FC for a given month. We used coefficients from the train-test-split model

Objective Function:

*maximize productivity for a given FC for a given month*

$$\mathbf{max\ 1.163F + 1.570P - 0.037F^2 - 0.145P^2}$$

*F and P are the levels of full time and part time turnover per month*

Constraints:

Turnover (positive or negative integer) for next month cannot exceed the headcount that the FC starts out with (equal to headcount for the last week of the previous month).

**Results for FC 6710 2020-06: a turnover of 16 FT and 5 PT maximizes productivity**

# Results and Insight

- Having some amount of turnover, usually associated with **negative worker well-being**, actually **improves productivity**
- The social science aspect of this study would look at “**which workers are the ones that should go?**” “**Is it ethical to lay off workers to increase productivity?**”
- It would be great to have **more turnover data for a longer time period**.
  - Focus on other features affecting productivity
  - Seasonality component
  - Separate data on those who left due to work conditions and those who left for other reasons
- Ultimately, the goal of this project is to **improve worker well-being**, not necessarily productivity
  - Turnover is usually associated with negative well-being, but it improves productivity, which is usually associated with positive well-being

Thank you!

Questions?