OEWS Autocoder Overview

Sam Fincher

Branch Chief, Occupational Analytics and Classification Branch

Occupational Employment and Wage Statistics

OEWS Autocoder

- OEWS Data Collection
- Autocoder Background and Overview
- Model History
- Thresholds
- Accuracy Monitoring
- Future Model

OEWS Data Collection

- Occupational Employment and Wage Statistics program
- Federal-State partnership
- Collects and publishes employment and wage data for about 830 occupations at the national, state, and metropolitan area as well as industry level data (NAICS)
- Occupations are coded to the Standard Occupational Classification (SOC) system

OEWS Data Collection

- OEWS fields collected:
  - Job Title (required)
  - Annual or Hourly Wage Rate (required)
  - Description of Duties (optional)
  - Department (optional)
  - Worksite Location (optional)
- Additional establishment level fields available

OEWS Autocoder Background

- Assigns Standard Occupational Classification (SOC) codes to respondent data
- Assists states and regions with the labor-intensive process of occupational coding
- Offers two potential SOC codes for job titles meeting the minimum confidence thresholds
- First piloted in 2016 and fully implemented in OEWS production systems in 2021
- Regularly retrained and tested to maintain currency of new job titles and occupations from latest closed panels
- Accuracy continues to improve due to growing training dataset and technological advances

Example of Autocoded Establishment

Reported job title from respondent

Autocode 1 and confidence score

Autocode 2 and confidence score

Use SOC Fr | Copy to Final SOC | Apply Crosswalk | Save In Progress | Export | RTE 13 | Calc Total 13

Page S 00 Total Rows: 8 | First Previous 1 / 1 Next

Job Title	Duties	Dept	Hourly	Annual	Wage Code	Emp Nur	AC1	AC1 Title	AC1 Score	AC2	AC2 Title	AC2 Score	Final SOC
Waitress	Take food orders and serve customers	Front of House	$10.00	$0.00		4	35-3031	Waiters and Waitresses	.79	35-9031	Hosts and Hostesses, Re	.02	
Prep/Line Cook	Prepare food to cook, and make plates for orders	Back of House	$14.00			1	35-2014	Cooks, Restaurant	.88	35-2021	Food Preparation Worker	.05	
Prep/Line Cook	Prepare food to cook and make plates for orders	Back of House	$12.50			1	35-2014	Cooks, Restaurant	.88	35-2021	Food Preparation Worker	.05	
Line Cook	Cook food to order	Back of House	$12.00			1	35-2014	Cooks, Restaurant	.92	25-9021	Farm and Home Manager	0	
Dishwasher	Cleans, bus, and sanitizes dishes according to hea	Back of House	$10.00			2	35-9021	Dishwashers	.89	35-2014	Cooks, Restaurant	.01	
Cashier	Receives cash and credit card payments from cus	Front of House	$10.00			2	41-2011	Cashiers	.93	35-3031	Waiters and Waitresses	.01	
Cashier	Receives cash and credit card payments from cus	Front of House	$9.50			1	41-2011	Cashiers	.93	35-3031	Waiters and Waitresses	.01	
Supervisor	Supervises front/back of house workers to mainta		$12.00			1	35-1012	First-Line Supervisors of	.64	51-1011	First-Line Supervisors of	.05	

Model History

- Supervised machine-learning, logistic regression model
- Trained on previously coded, labeled job titles
- Three model inputs:
  - Job Title
  - NAICS Code (Industry)
  - EIN (Employer Identification Number)
- Pilot model used approximately 1 million records of training data and 491,000 records of validation data
- Current production model uses 8.1 million records of training data and 4.6 million records of validation data

Autocoder Thresholds

- Thresholds were revamped in 2022 to maximize output to states
  - Autocoder is better at some occupations than others
  - Varying thresholds by SOC Major group
  - Thresholds range from 0.43 to 0.59, and represent the level at which each major group achieves its peak overall accuracy when combined with the accuracy of state coders
- Threshold confidence score must meet one of the following:
  - Autocode1 is at least as high as assigned major group threshold
  - Combined score of Autocode 1 and Autocode 2 achieves 0.65

Autocoder Thresholds

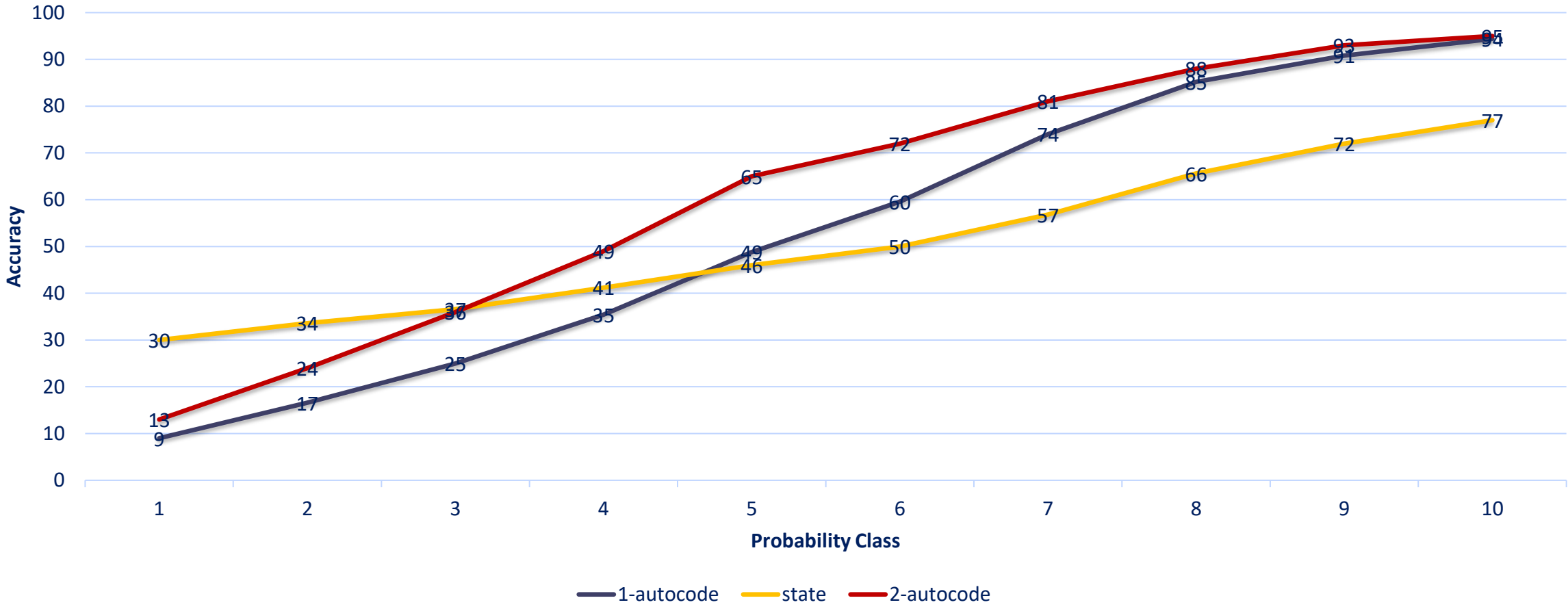
- A lower threshold indicates a higher overall accuracy for a SOC Major Group
- Least confident in Major Group 25, Educational Instruction and Library occupations
- Most confident in Major Group 45, Farming, Fishing, and Forestry occupations

SOC Major Group	Threshold
11 - Management	0.52
13 - Business and Financial Operations	0.55
15 - Computer and Mathematical	0.47
17 - Architecture and Engineering	0.48
19 - Life, Physical, and Social Science	0.50
21 - Community and Social Service	0.57
23 - Legal	0.54
25 - Educational Instruction and Library	0.59
27 - Arts, Design, Entertainment, Sports, and Media	0.49
29 - Healthcare Practitioners and Technical	0.46
31 - Healthcare Support	0.53
33 - Protective Service	0.56
35 - Food Preparation and Serving	0.51
37 - Building and Grounds Cleaning and Maint.	0.52
39 - Personal Care and Service	0.55
41 - Sales and Related	0.53
43 - Office and Administrative Support	0.52
45 - Farming, Fishing, and Forestry	0.43
47 - Construction and Extraction	0.49
49 - Installation, Maintenance, and Repair	0.47
51 - Production	0.55
53 - Transportation and Material Moving	0.48

Accuracy Monitoring

- "Gold-Code" dataset used to separately test the accuracy of the Autocoder and determine thresholds
- The Gold-Code dataset contains job titles that are expertly coded and agreed upon by two state coders without assistance from the autocoder
- This process brings an impartial review of overall coding accuracy of both the autocoder and human coders
- Current Gold-Code file contains 60,000 titles and we are currently collecting an additional 16,000

Probability Class Accuracies



Monitoring Autocode Use

- Autocoder State Report provided to the field each panel
- Tracking percentage of autocodes being used as final production code
- Encourage use of Autocoder, but careful coders are not “blind coding”
- All autocodes need to be reviewed by humans before applying
- November 2023 Panel
  - Out of 3,548,635 job titles, 2,447,686 (69%) above thresholds and provided an autocode
- The final chosen SOC code matches an autocode **87.7%** of the time
  - When provided a code, Autocode 1 matches the final SOC code 78.9% of the time
  - Autocode 2 matches the final SOC code 8.8% of the time

Future Model

- In the process of transitioning to new model
- Convolution Neural Network (CNN)
  - Job titles are vectorized and used to create embedding matrixes
- Early results are promising, with 4% increase in overall accuracy from current production model
- Currently conducting model comparisons
  - New model has higher accuracy in lower probability classes

Contact Information

Sam Fincher

fincher.samuel@bls.gov

Occupational Analytics and Classification

OEWS_OAC@bls.gov