



AgEcon SEARCH
RESEARCH IN AGRICULTURAL & APPLIED ECONOMICS

The World's Largest Open Access Agricultural & Applied Economics Digital Library

This document is discoverable and free to researchers across the globe due to the work of AgEcon Search.

Help ensure our sustainability.

Give to AgEcon Search

AgEcon Search
<http://ageconsearch.umn.edu>
aesearch@umn.edu

*Papers downloaded from **AgEcon Search** may be used for non-commercial purposes and personal study only. No other use, including posting to another Internet site, is permitted without permission from the copyright owner (not AgEcon Search), or as allowed under the provisions of Fair Use, U.S. Copyright Act, Title 17 U.S.C.*

Green Skills and Green Potential Prevalence: An Application to Appalachia

Susan Chen, Agricultural and Applied Economics, Virginia Tech (ecsusan@vt.edu)

Yang Cheng, Agricultural and Applied Economics, Virginia Tech (yangcheng@vt.edu)

Andrew Katz, 2 Andrew Katz, Engineering Education, Virginia Tech (akatz4@vt.edu)

***Selected Poster prepared for presentation at the 2022 Agricultural & Applied Economics Association
Annual Meeting, Anaheim, CA; July 31-August 2***

Copyright 2022 by Susan Chen, Yang Cheng, and Andrew Katz. All rights reserved. Readers may make verbatim copies of this document for non-commercial purposes by any means, provided that this copyright notice appears on all such copies.

Green Skills and Green Potential Prevalence: An Application to Appalachia



Susan Chen¹, Yang Cheng¹, Andrew Katz²

¹ Agricultural and Applied Economics , Virginia Tech (yangcheng@vt.edu)

² Andrew Katz, Engineering Education, Virginia Tech (akatz4@vt.edu)

Introduction

Public policy and social investment incentivize reduced carbon emissions and greater resource efficiency in a green economy. The shift toward a greener economy will induce changes in the labor market. New industries and jobs will emerge that are linked to the green economy. Energy-intensive and polluting industries will reallocate or scale down their operations while the expansion of environmentally friendly industries increases. This decreases employment in non-green jobs (Mulatu and Wossink, 2014; Kahn and Mansur, 2013) while increasing employment in environmentally sustainable industries. These structural change in the labor market will result in both a change in the **types** of occupations and the **skills** demanded within existing occupations. It is important to know which occupation, industry, and geographic regions will struggle with a green transition. This information provides policymakers with a framework of information to craft and address regional disparities and to make more informed decisions of how and where to train the future workforce.

Research Goal

- To create a measure that summarizes a region's green job skill capital.
- To create a measure of labor market vulnerability to a green labor market transition.

Data

In this project, we use Occupational data and employment data from O *NET. O *NET is database developed by the US Department of Labor. The dataset contains detailed occupation-level information on the tasks and skills involved as well as a list of tasks that are unique to green jobs. The most recently available national data containing occupational composition comes from the American Community Survey (ACS) 2019-1 year data.

Methodology

We first construct Green Potential for each occupation by applying Natural Language Processing and clustering. Then following Rutzer's (2020) methodology, we apply machine learning techniques to identify the skills needed for smooth transitioning of occupations. We then merge these occupations with occupation data from the American Community Survey (ACS) to create a green job skill matrix for each community.

Construct Green Potential

This indicator represents the ease of transitioning into a green economy. A higher value of G means that occupation has a higher ratio of green tasks over all tasks required for this job.

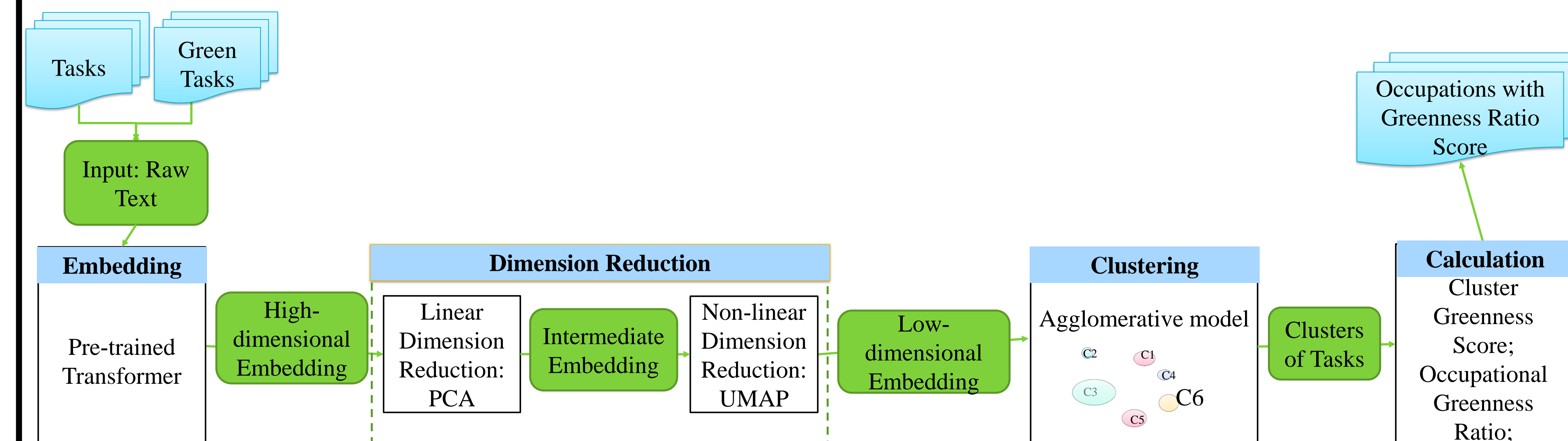


Figure 1: Work Flow of Constructing Occupational Green Potential

Table 1: Constructing a Cluster Greenness

Cluster Score	Vector of Tasks	Total Tasks in Cluster	Number of Green Tasks in cluster	Cluster Greenness score
1	Task 1, Task 2, Task 20, Task 51, Task 52, Task 1001, Task 12,298, Task 12,999	8	2	0.25
2	Task 6, Task 7, Task 8, Task 9, Task 8000	5	3	0.6
3	Task 14, Task 15, Task 16, Task 17	3	0	0
4	Task 33, Task 34	2	1	0.5

Table 2 Green Task Potential of Occupations

Occupation	# of necessary tasks	Tasks	Cluster	Cluster Greenness score	Occupation Green Task Potential Score
civil engineer	2	Task 1	1	0.25	0.85
civil engineer	2	Task 6	2	0.6	0.85
economist	4	Task 1	1	0.25	1.35
economist	4	Task 6	2	0.6	1.35
economist	4	Task 14	3	0	1.35
economist	4	Task 33	4	0.5	1.35

Results

By linking our Green Potential Index driven by the O*NET data and the occupation data from ACS, we can conduct our analysis at the PUMAs level, the lowest level of geographic identifier available in the ACS, and map the Green Potential geographically.

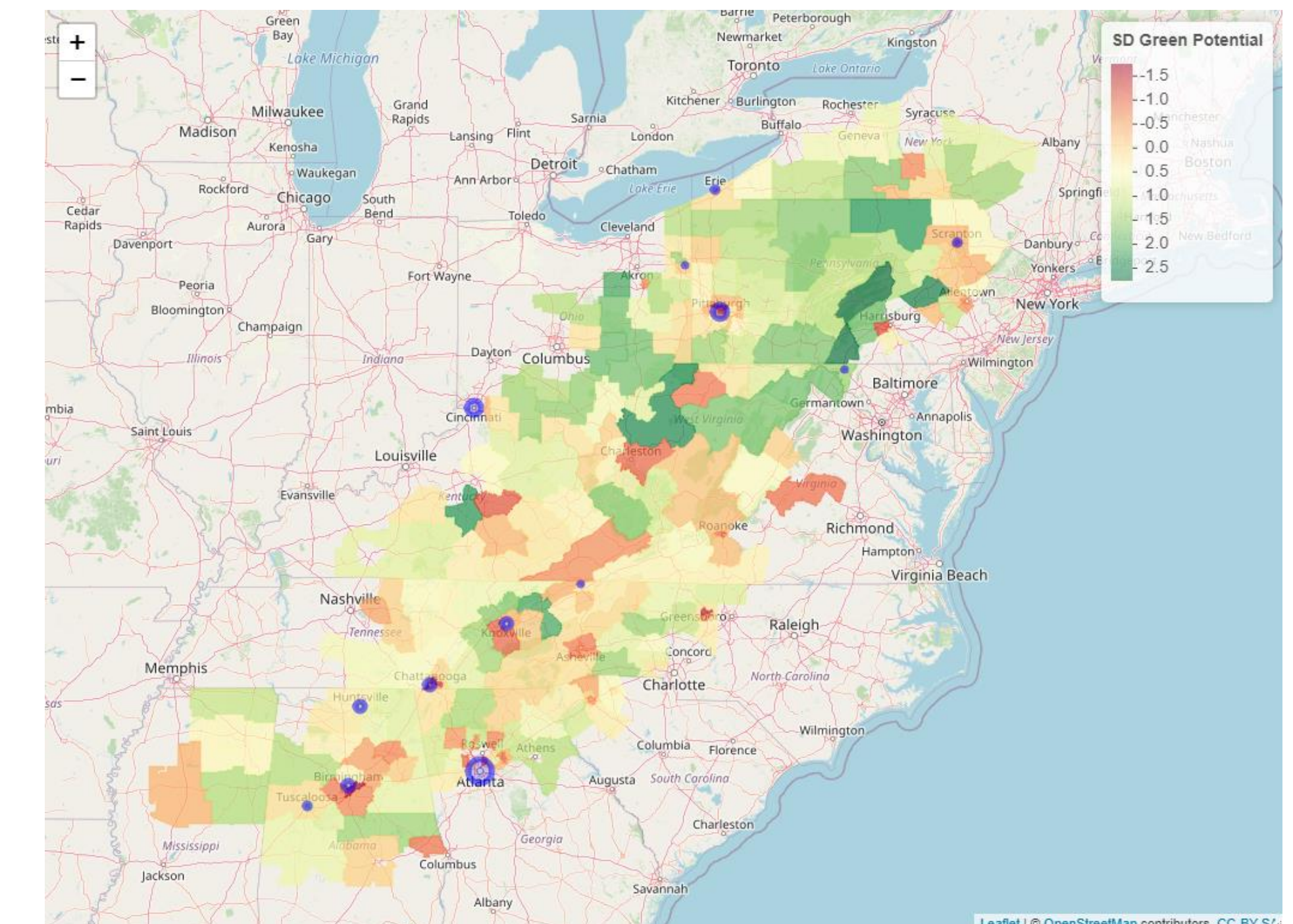


Figure 2: Green Potential by PUMA, Appalachia

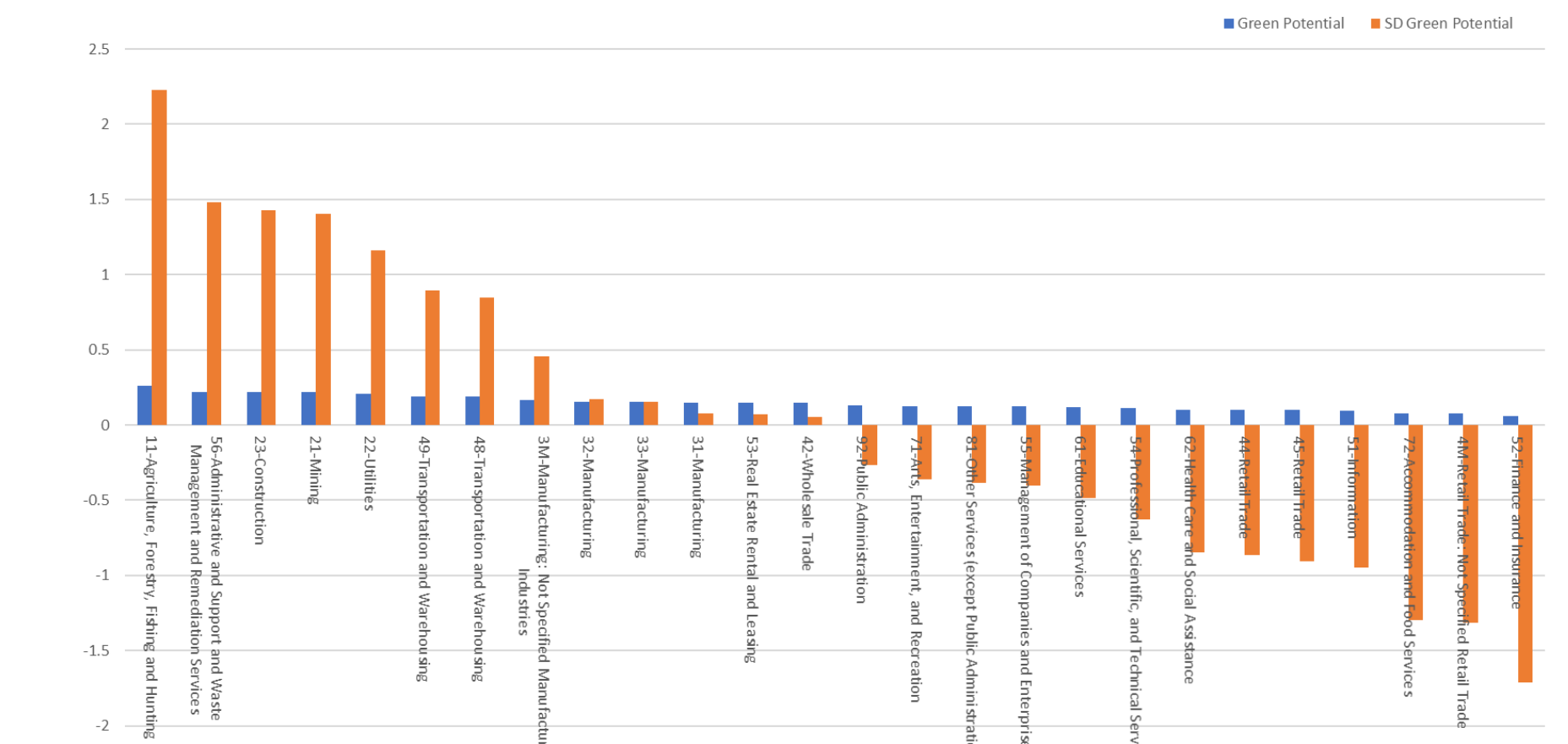


Figure 3: Green Potential by NAICS, Appalachia

Future Plan

Future Extensions will build on this measure to construct a skill distance index. A skill distance measure will estimate how far a regions skill capital is from a hypothetical amount of green skill capital. It is a measure of labor market vulnerability to transitioning to a green economy.