



# Evaluation of a Multiple Measures Placement System Using Data Analytics

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## Agenda

- Why use multiple measures assessment for placement
- The national picture
- Multiple measures options in the current moment
- What we learned from research

# Multiple Measures

## **Definition of Multiple Measures Assessment**

....a system that **combines** two or more measures to place students into appropriate courses and/or supports.

(Barnett and Reddy, 2017)



CAPR \ 2019



## Percent of Colleges Using Measures Other than Standardized Tests for Assessment



SOURCES: 2011 data from Fields and Parsad (2012); 2016 data from CAPR's institutional survey.

NOTE: The Fields and Parsad (2012) reading statistics are for reading placement only, whereas the CAPR survey data are for both reading and writing.

## Processes Used to Determine College Readiness in Community Colleges



## Students Needing 1+ Developmental Education Course (NCES, 2013)





## Under-placement and Over-placement

		<b>Placement According to Exam</b>	
		Developmental	College Level
Student Ability	Developmental		<b>Over-placed</b> (English – 5%) (Math – 6%)
	College Level	<b>Under-placed</b> (English – 29%) (Math – 18%)	



## What measures to consider

- Tests
- HS GPAs
- Both together
- Both plus other data points



#### SUNY COLLEGE 2: ENGLISH

#### SUNY COLLEGE 2: MATH





## Some things to consider.....

- Better assessment systems are needed.
- Tests don't do a good job.
- HS GPA is the best predictor.
- None of these is a *great* predictor.

## **Multiple Measures Options**

MEASURES	SYSTEMS OR APPROACHES	PLACEMENTS
<ul> <li><u>Administered by college</u>:</li> <li>1. Traditional or alternative placement tests</li> <li>2. Non-cognitive assessments</li> <li>3. Computer skills or career inventory</li> <li>4. Writing assessments</li> <li>5. Questionnaire items</li> </ul>	<ul> <li>Waiver system</li> <li>Decision rules or bands</li> <li>Placement formula (algorithm)</li> <li>Directed self- placement</li> </ul>	<ul> <li>Placement into traditional courses</li> <li>Placement into alternative coursework</li> <li>Placement into support services</li> </ul>
<ul> <li><u>Obtained from elsewhere</u>:</li> <li>1. High school GPA</li> <li>2. Other HS transcript information</li> <li>3. Standardized test results (e.g., ACT, SAT, AP)</li> </ul>		

# Digging in on the HS GPA

(with thanks to John Hetts and Brad Bostian)

- How are we going to get the HS GPA?
- *Our* test is different/better/more awesome.
- High school GPA is only predictive for recent graduates.
- Different high schools grade differently.

#### Sources of HS transcript data

#### Self-report research

- The students bring a transcript
- The high school sends
- Obtained from state data files
- Self report

Note: Consider using the 11<sup>th</sup> grade GPA

- UC admissions uses self-report but verifies after admission. In 2008, at 9 campuses, 60,000 students. No campus had >5 discrepancies b/w reported grades and student transcripts (Hetts, 2016)
- College Board: Shawn & Mattern, 2009: "Students are quite accurate in reporting their HSGPA", r = .73.
- ACT research often uses self-reported GPA and generally find it to highly correlate with students' actual GPAs: ACT, 2013: *r* = .84.



#### None of the tests are that good for placement.

North Carolina ENGLISH

North Carolina MATH



From Bostian (2016), North Carolina Waves GPA Wand, Students Magically College Ready; adapted from research of (Belfield & Crosta, 2012)

# HS GPA is a better predictor than test results for a long time (from Hetts, 2016)



MMAP (in preparation): correlations b/w predictor and success (C or better) in transfer-level course by # of semesters since HS

### For the most part, college grades stay parallel with feeder high school grades (Bostian, 2016)





## Non-cognitive assessments

Development of non-cognitive skills promotes students' ability to think cogently about information, manage their time, get along with peers and instructors, persist through difficulties, and navigate the landscape of college...(Conley, 2010).

Non-cognitive assessments may be of particular value for:

- Nontraditional (older) students.
- Students without a high school record.
- Students close to the cut-off on a test.

## Ways to Combine Measures

- Algorithms/predictive analytics
- Decision rules or bands
- Directed self-placement













## Directed Self Placement in Math (Kosiewicz and Ngo, 2019)

- More students chose to enroll in college- and transfer-level math courses
  - More female, Black, and Hispanic students enrolled in the lowest levels of math.
- There was decreased withdrawal from courses.
- More students completed the math required for Associates degree.
  - Especially White, Asian, and male students.

# Research on a Multiple Measures, Data Analytics Placement System

#### CAPR

# **Organization of CAPR**



**CCRC** 

Descriptive Study of Developmental Education

Evaluation of The New Mathways Project (RCT in TX) Evaluation of New Assessment Practices (RCT in NY)

#### **Supplemental Studies**

## **CAPR Assessment Research**

- 1. 7 State University of New York (SUNY) community colleges.
- 2. Each worked with CAPR team to develop an alternative placement method using an algorithm.
- 3. Students were randomly assigned to be placed using either the existing placement method or the algorithm.
- 4. We looked for differences in student outcomes based on placement method.

## Creating the algorithm



- Historical data from 3 cohorts of students
- Select students who:
  - Took a placement test
  - Took a college-level course first
- Use their outcome in the initial college-level course to gauge how well certain factors predict success (Passing the course with a C or better)
  - HS GPA
  - ACCUPLACER
  - Other HS information (time from graduation, GED, Regents exams, etc.)
- Establish minimum acceptable probability for success in college-level course





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# **Overall Findings**

**Full Analytic Sample** 

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### Final Analysis Sample

Sample = 12,971 first year students across 7 colleges and 3 cohorts

- 51% of students assigned to program group (n=6,589)
- 49% of students assigned to business-as-usual group (n=6,382)
- 86% of students enroll into at least one course in 2016 (n=11,102)



### Differences in Placement among Program Students



### Treatment Effects: College-Level English



\*\*\*p < .01, \*\*p < .05, \*p < .10.



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### Treatment Effects: College-Level Math



\*\*\*p < .01, \*\*p < .05, \*p < .10.

# Subgroup Analyses

Full Analytic Sample

#### Treatment Effects: CL English Placement





## Treatment Effects: CL English Completion by Gender



## Treatment Effects: CL English Completion by Pell Status



\*\*\**p* <. .01, \*\**p* < .05, \**p* < .10

## Treatment Effects: CL Eng Completion by Race/Ethnicity







#### Treatment Effects: CL Math Placement



\*\*\**p* <. .01, \*\**p* < .05, \**p* < .10



### Treatment Effects: CL Math Completion by Gender



\*\*\**p* <. .01, \*\**p* < .05, \**p* < .10



### Treatment Effects: CL Math Completion by Pell Status



\*\*\**p* <. .01, \*\**p* < .05, \**p* < .10



## Treatment Effects: CL Math Completion by Race/Ethnicity



<sup>\*\*\*</sup>*p* <. .01, \*\**p* < .05, \**p* < .10



# Summary

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## Summary of Findings

- Most program group students whose placement changed received a higher placement than they would have received under the status quo system
  - Placement gaps narrowed in favor of women and traditionally underrepresented groups in English
  - Placement gaps between White students and Black and Hispanic students widened in math
- Program group students were more likely to enroll in and complete (with a grade of C or higher) a college-level English course within 3 terms of testing
- Program group students were more likely to enroll in and complete (with a grade of C or higher) a college-level math course within 1 term of testing

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