

#### Identifying factors influencing timely graduation: Recent research and developments

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#### Graduation Initiative 2025 Goals CSU Fullerton

Metric	2025 Goal	Most Recent Rate
Freshman 6-Year Graduation	75%	62%
Freshman 4-Year Graduation	44%	22%
Transfer 2-Year Graduation	44%	32%
Transfer 4-Year Graduation	85%	75%
Gap - Underrepresented Minority	0	9 % points
Gap – Pell	0	4 % points





Institutional Research and Analytical Studies



# Results from 7-yrs ago...

- Predicting 6-yr graduation of Full-time FTF (cohorts from fall 1998 to 2002)
  - 13,412 FTF students entered
  - 6,589 (49.1%) graduated in six years or less
- Predictors included:
  - HS GPA, SAT, ELM
  - Gender
  - Ethnicity
  - Major in 1<sup>st</sup> term
  - Admissions area
  - Units taken in 1<sup>st</sup> Term



# Results from 7-yrs ago...

- Predicting 4-yr graduation of Transfers (cohorts from fall 1998 to 2004)
  - 33,839 transfer students entered
  - 21,778 (64%) graduated in four years or less
- Predictors included:
  - Transfer GPA
  - Major
  - Full-time status
  - Ethnicity
  - Gender



# Summary

- Less well academically prepared students are less likely to succeed.
- Men are less likely to succeed than women
- URM are less likely to graduate
- Higher # of Units in 1<sup>st</sup> term is associated with graduation
- STEM majors less likely to succeed
- Students from outside of our local admissions area are less likely to succeed
- Freshmen who complete remediation in mathematics early in their academic careers are more likely to succeed.









# Four-year graduation rates by student characteristics

- Data
  - FTF from fall 2008 to fall 2012 cohorts
    - 20621 Full-time FTFs.
    - 3823 (18.5%) graduated in 4 years or less.



# 4-yr graduation rate by HS GPA





## 4-yr graduation rate by SAT





#### 4-yr graduation rate by 1st Term Units





# 4-yr graduation rate by gender





#### 4-yr graduation rate by Major at Entry





# 4-yr graduation rate by ethnicity





#### 4-yr graduation rate by parental education





# 4-yr graduation rate by Pell Grant

21.40% 20.00% 15.00% 13.60% 10.00% 5.00% 0.00% Not received Pell Received



25.00%

### 4-yr graduation rate by admission area





# 4-yr graduation rate by college preparatory math courses





# 4-yr graduation rate by college preparatory English courses





# Method

- Logistic regression
  - Dependent variable:
    - Graduated in 4 years (yes/no)
  - Independent variables
    - Categorical:
      - Gender, ethnicity, parental education, major, admission area, Pell Grant received.
      - College preparatory courses
    - Continuous:
      - HS GPA, units taken during the first time, SAT (ACT converted if SAT missing)



# Result summary

- Nagelkerke *R*<sup>2</sup>: 21.5%
- 70% overall accuracy

	Predictor	Odds Ratio	Sig.		
	HS GPA	1.4	<0.001		
	Units taken, first term	1.4	<0.001		
	SAT	1.2	<0.001		
	Men	0.5	<0.001		
	Local	0.9	<0.001		
	Pell Received	0.7	<0.001		
	American Ind.	0.7	0.281		
	Black	0.8	0.028		
ity	Hispanic	0.6	<0.001		
juri	Asian	0.7	<0.001		
Etl	Non-Res	0.7	0.035		
	Pacific Islander	0.3	0.032		
	Unknown	0.7	<0.001		
ed	First Generation	0.7	<0.001		
rent	Some College	0.8	<0.001		
ра	Unknown	0.9	0.149		
	Arts	0.3	<0.001		
	Business	0.7	<0.001		
Z	Communications	1.3	0.001		
۱ajc	Engineering	0.1	<0.001		
2	Health and Human Dev	0.6	<0.001		
	Natural Science/Math	0.3	<0.001		
	Other	0.5	<0.001		
Ŋ	English	1.0	0.953		
ato	Math	1.1	<0.001		
par	History	1.1	0.037		
Pre	Social Science	1.0	0.337		
a B G	Foreign Language	1.1	0.001		
olle	Arts	1.0	0.722		
Ŭ	Electives	1.2	0.072		



# Two-year graduation rates by transfer student characteristics

- Data
  - Transfer students from fall 2012 to fall 2014 cohorts (including spring)
    - 13848 Transfer students
    - 4492 (32.4%) graduated in 2 years or less



#### 2-yr graduation rate by Transfer GPA





#### 2-yr graduation rate by 1st Term Units





# 2-yr graduation rate by gender





#### 2-yr graduation rate by major at entry





#### 2-yr graduation rate by parental education





# 2-yr graduation rate by ethnicity





### 2-yr graduation rate by admission area





### 2-yr graduation rate by SB 1440 status





# Method

- Logistic regression
  - Dependent variable:
    - Graduated in 2 years (yes/no)
  - Independent variables
    - Categorical:
      - Gender, ethnicity, parental education, major, admission area,
         Pell Grant received, Fall entry, SB 1440.
    - Continuous:
      - Transfer GPA, units taken during the first time, Transfer Units



# **Results summary**

- Nagelkerke *R*<sup>2</sup>: 29.3%
- 70% overall accuracy

	Predictor	Odds Ratio	Sig.
	Transfer GPA	1.3	<0.001
	Transfer Units	1.3	<0.001
	Units taken, first term	2.4	<0.001
	Fall entry	1.2	0.066
	Men	0.7	< 0.001
	local	1.3	0.002
	SB1440 status	1.3	<0.001
	Pell Received	0.8	< 0.001
	American Ind.	1.7	0.213
	Black	1.0	0.757
itγ	Hispanic	0.9	0.104
nic	Asian	0.8	<0.001
Eth	Non-Res	0.9	0.28
	Pacific Islander	1.4	0.43
	Unknown	0.9	0.476
ed	First Generation	1.0	0.665
ent	Some College	1.1	0.031
par	Unknown	0.8	0.041
	Arts	0.1	< 0.001
	Business	0.5	< 0.001
F	Communications	1.2	0.022
1ajc	Engineering	0.03	< 0.001
2	Health and Human Dev	1.3	<0.001
	Natural Science/Math	0.1	< 0.001
	Other	0.3	0.299



# Summary

- Less well academically prepared students are less likely to succeed.
- Men are less likely to succeed than women
- URM are less likely to graduate
- Higher # of Units in 1<sup>st</sup> term is associated with graduation
  - STEM majors less likely to succeed
- Students from outside of our local admissions area are less likely to succeed
- Freshmen who complete remediation in mathematics early in their academic careers are more likely to succeed.



# More options

- Feature engineering
- Discrete-time survival analysis
- Competing risk analysis
- Survival Tree



# Feature Engineering

#### Feature engineering: create new variables from existing ones.

Example: application data Quantify institutional characteristics

Institution	Mean SAT	Mean HS GPA	First Gen %	URM %	admission rate	application count
А	1050	3.23	0.18	0.1	0.6	73
В	1092.38	2.99	0.06	0.04	0.46	96
С	1065.73	3.25	0.06	0.06	0.67	78
D	1055.37	3.09	0.18	0.18	0.67	45
E	1068.68	2.98	0.02	0.05	0.43	42

- These predictors were added to the FTF model for 4-year graduation
  - Nagelkerke *R*<sup>2</sup>: 23%
  - 71% overall accuracy



# Feature Engineering

#### Feature engineering: create new variables from existing ones.

Example: DFW

- 1. Calculate average DFW for courses taken in the first term
- 2. Calculate Max values of DFW among the courses taken
- 3. Count the number of courses with DFW >=20%

student	Course	DFW
1	DANC 101	3.0%
1	нсом 100	10.0%
1	HIST 110B	14.0%
1	MATH 115	30.0%
2	ENGL 99	12.0%
2	HIST 110B	14.0%
2	MATH 115	30.0%
2	POSC 100	22.0%



student	Avarage DFW	Max DFW	Count > 20 %
1	14.3%	30.0%	1
2	19.5%	30.0%	2



# Feature Engineering

#### Feature engineering: create new variables from existing ones.

Example: DFW.

- 1. Calculate average DFW for courses taken in the first term
- 2. Calculate Max values of DFW among the courses taken
- 3. Count the number of courses with DFW >=20%
- These predictors were added to the FTF model for 4-year graduation
  - Nagelkerke R<sup>2</sup>: 22%
  - 71% overall accuracy

Predictor	Odds Ratio	Sig.
Average DFW by faculty	1.0	0.5
Average DFW by course	0.9	<0.001
Max DFW by faculty	0.9	0.18
Max DFW by course	0.9	0.01
Count of high DFW by faculty	1.0	0.35
Count of high DFW by course	1.3	<0.001



- This can be used to examine the length of time until the occurrence of some event, such as time to degree.
- Two possible outcomes:
  - event happened
  - Not happened (censored)





#### Data structure





### In spss

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In R, include -1 in the formula. For eample, glm(GRAD ~ T1 + T2 + T3... - 1, family="binomial")



- Data
  - FTF from fall 2008 to fall 2013 cohorts
    - 26059 FTFs.
- Method
  - Discrete-time survival analysis
    - DV: Graduated = 1; Not graduated = 0
    - IV: Use the same set of predictors as logistic regression analysis plus 18 binary time indicator variables.
      - Step 1: time indicators only
      - Step 2: enter all the other predictors.



#### Step 1: Time Indicator variables only

Term	В	hazard		Hazard is the probability of an event during
T1	-21.2	0.0%		the term given the event been't accured
T2	-21.2	0.0%		• the term given the event hash t occured
Т3	-10.0	0.0%		earlier. Convert B by 1/(1 + EXP(B))
T4	-10.0	0.0%		
T5	-7.5	0.1%	60.0%	
Т6	-5.1	0.6%	00.070	
Т7	-4.4	1.2%		
Т8	-1.1	24.4%	50.0%	٨٨
Т9	-1.7	15.8%		$\Lambda$ $\Lambda$
T10	0.0	49.6%	40.0%	
T11	-1.2	23.4%		
T12	0.0	49.4%	30.0% ——	
T13	-1.2	23.6%		
T14	-0.5	38.5%	20.0%	
T15	-1.1	24.4%	20.070	
T16	-0.5	38.5%		
T17	-1.3	21.9%	10.0%	
T18	-1.3	21.5%		
			0.0% —	
			1	2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18

CALIFORNIA STATE UNIVERSITY

Step 2: Add predictors.

	Predictor	Odds Ratio	Sig.
	HS GPA	1.189	<0.001
	Units taken	1.224	<0.001
	SAT	1.071	<0.001
	Men	0.69	<0.001
	Local	0.874	< 0.001
	Pell Received	0.856	<0.001
	American Ind.	1.028	0.902
	Black	0.739	<0.001
sity	Hispanic	0.663	<0.001
jurio	Asian	0.736	<0.001
Ē	Non-Res	0.662	<0.001
	Pacific Islander	0.643	0.06
	Unknown	0.827	<0.001
ed	First Generation	0.799	<0.001
ent	Some College	0.866	<0.001
par	Unknown	0.907	0.046
	Arts	0.518	<0.001
	Business	0.773	<0.001
5	Communications	1.289	<0.001
lajc	Engineering	0.366	<0.001
2	Health and Human Dev	0.794	<0.001
	Natural Science/Math	0.413	<0.001
	Other	0.65	<0.001
Ņ	English	0.917	0.006
ator	Math	1.043	0.011
bara	History	1.054	0.038
rep	Social Science	0.984	0.345
ge F	Foreign Language	1.024	0.122
olle	Arts	0.979	0.197
ŭ	Electives	0.955	0.355



# More options

- Competing risk analysis
  - Use survival data
  - More than two outcomes
    - Graduated = 1, dropped out = 2, neither = 0.
- Machine learning
  - Survival data can be examined more modern methods.
  - In R, for example, you can use DStree to build discretetime survival tree.



### In SPSS







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40.0%         Predictor         Odds Ratio         Sig.         Odds Ratio         Sig.           30.0%	50.0%	50.0%				Graduated		Dropped out	
40.0%       HS GPA       1.2       <0.001			$\Lambda$ $\Lambda$		Predictor	Odds Ratio	Sig.	Odds Ratio	Sig.
30.0%       Units taken       1.2       <0.001	40.0%				HS GPA	1.2	<0.001	0.8	<0.001
30.0%       SAT       1.1       -0.001       1.0       0.12         20.0%       Men       0.7       -0.001       1.2       -0.001         10.0%       Men       0.7       -0.001       1.0       0.665         T1 T2 T3 T4 T5 T6 T7 T8 T9 T10T11T12T13T14T15T16       Men       0.7       -0.001       1.0       0.655         Graduated       Dropped       Men       0.7       -0.001       1.0       0.665         Menrican Ind.       1.0       0.555       1.3       0.174         Graduated       Dropped       Menres       0.7       -0.001       1.2       0.002         Menres       0.7       -0.001       1.0       0.655       1.3       0.174         Hispanic       Mon-Res       0.7       -0.001       1.2       0.002         View       Menres       0.7       -0.001       1.2       0.003         View       Menres       0.7       -0.001       1.0       0.58         Menres       0.7       -0.001       1.2       -0.001       1.0       0.48         Menres       0.7       -0.001       1.2       -0.001       1.1       0.001         Matori       0.5	20.00/				Units taken	1.2	<0.001	1.0	0.848
20.0%       Men       0.7       <0.001	30.0%	30.0%			SAT	1.1	<0.001	1.0	0.172
Interview         Iocal         0.9         <0.001         0.7         <0.001           0.0%         I 12 T3 T4 T5 T6 T7 T8 T9 T10T11T12T13T14T15T16         American Ind.         1.0         0.855         1.3         0.070           0.0%         I 12 T3 T4 T5 T6 T7 T8 T9 T10T11T12T13T14T15T16         Graduated         Dropped         American Ind.         0.7         <0.001	20.0%	0.0%			Men	0.7	<0.001	1.2	<0.001
10.0%         Pell Received         0.9         <0.001         1.0         0.665           0.0%         11 T2 T3 T4 T5 T6 T7 T8 T9 T10T11T12T13T14T15T16         American Ind.         1.0         0.885         1.3         0.174           Graduated         Dropped         Dropped         Asian         0.7         <0.001	2010/0				Local	0.9	<0.001	0.7	<0.001
0.0%       I T2 T3 T4 T5 T6 T7 T8 T9 T10T11T12T13T14T15T16       American Ind.       1.0       0.855       1.3       0.174         Black       0.7       <0.001	10.0%				Pell Received	0.9	<0.001	1.0	0.665
0.0%       11 T2 T3 T4 T5 T6 T7 T8 T9 T10T11T12T13T14T15T16       Black       0.7       <0.001					American Ind.	1.0	0.855	1.3	0.174
In 12 13 14 15 16 17 18 19 11011112113114115116       Asian       0.7       <0.001	0.0%	T1 T2 T3 T4 T5 T6 T7 T8 T9 T10T11T12T13T14T15T16		hnicity	Black	0.7	<0.001	1.2	0.002
Graduated         Dropped           Graduated         Dropped           Asian         0.7         <0.001					Hispanic	0.7	<0.001	1.0	0.5
Image: second					Asian	0.7	<0.001	0.8	< 0.001
Black       47.00%         Hispanic       54.90%         Asian       64.50%         White       64.60%         Non Local       59.00%         Local       60.20%				LT L	Non-Res	0.7	<0.001	1.2	0.083
Image: bit of the second se					Pacific Islander	0.6	0.054	0.7	0.256
image: second					Unknown	0.8	<0.001	1.1	0.089
Some College         0.9         <0.001         1.1         <0.001           in the second s				ted	First Generation	0.8	<0.001	1.2	<0.001
$ \begin{array}{c c c c c c c c c c c c c c c c c c c $				ren	Some College	0.9	<0.001	1.1	<0.001
6yr-grad rate       Arts       0.5       <0.001				ba	Unknown	0.9	0.055	1.1	0.13
6yr-grad rate       Business       0.8       <0.001					Arts	0.5	<0.001	1.3	<0.001
Black       47.00%         Hispanic       54.90%         Asian       64.50%         White       64.60%         Non Local       59.00%         Local       60.20%             English       0.9         0.123       0.001         0.4       <0.001			6yr-grad rate		Business	0.8	<0.001	0.9	0.164
Hispanic       54.90%         Asian       64.50%         White       64.60%         Non Local       59.00%         Local       60.20%         English       0.9         Math       1.0         0.025       0.9         0.027       1.1         0.028       0.001         0.008       1.1         0.011       0.025         0.020%       0.025         0.021       0.025         0.021       0.025         0.021       0.025         0.022       0.021         0.025       0.9         0.021       0.025         0.021       0.025         0.022       0.021         0.023       0.024         0.024       0.9         0.025       0.9         0.024       0.9         0.025       0.9         0.024       0.9         0.025       0.9         0.024       0.9         0.9       0.025         0.9       0.021         0.9       0.021         0.001       0.251         0.01		Black	47.00%	o	Communications	1.3	<0.001	0.9	0.123
Asian       64.50%       Natural Science/Math       0.4       <0.001		Hispanic	E4.00%	Maj	Engineering	0.4	<0.001	1.2	< 0.001
Asian       64.50%       Natural Science/Math       0.4       <0.001		i iispanic A sism	54.50%		Health and Human Dev	0.8	<0.001	1.0	0.466
White         64.60%         Other         0.6         <0.001         1.0         0.448           Non Local         59.00%         English         0.9         0.008         1.1         0.032           Local         60.20%         Math         1.0         0.025         0.9         <0.001		Asian	64.50%		Natural Science/Math	0.4	<0.001	1.1	0.011
Non Local       59.00%       English       0.9       0.008       1.1       0.032         Local       60.20%       Math       1.0       0.025       0.9       <0.001		White	64.60%		Other	0.6	< 0.001	1.0	0.448
Local       60.20%       History       1.1       0.025       0.9       <0.011         History       1.1       0.027       1.1       0.005       0.9       <0.001		Non Local	59.00%	ory		0.9	0.008	1.1	0.032
Image       Image <td< td=""><td></td><td>Local</td><td>60.20%</td><td>arat</td><td>History</td><td>1.0</td><td>0.025</td><td>0.9</td><td>&lt;0.001</td></td<>		Local	60.20%	arat	History	1.0	0.025	0.9	<0.001
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$				ebe	Social Science	1.1	0.027	1.1	0.005
$\begin{array}{c} & 1.0 & 0.103 & 0.001 \\ \\ \hline \\ \\ \\ \\ \\ \\ \\ \\ \\ \\ \\ \\ \\ \\ \\ \\ $				e Pr	Foreign Language	1.0	0.240	0.9	0.001
$\frac{1}{0}$ Electives 1.0 0.251 1.1 0.005				le g	Arts	1.0	0.103	1 1	0.001
				Co	Electives	1.0	0 351	1.1	0.836

CALIFORNIA STATE UNIVERSITY FULLERTON"

#### Result of Discrete-time survival tree





## End

## Question?

