AMERICAN SOCIOLOGICAL REVIEW

OFFICIAL JOURNAL OF THE AMERICAN SOCIOLOGICAL ASSOCIATION

ONLINE SUPPLEMENT

to article in

AMERICAN SOCIOLOGICAL REVIEW, 2018, Vol. 83

Who Are the "Illegals"? The Social Construction of Illegality in the United States

René D. Flores University of Chicago

Ariela Schachter Washington University in St. Louis

Table S1. Respondent Characteristics

-	Proportion or	
	Mean (SD)	N
Democrat or Lean Democrat	.481	1,484
Republican or Lean Republican	.454	1,484
Interested in Politics (yes $= 1$)	.461	1,504
Conservative Political Ideology (yes = 1)	.556	1,467
College Degree + (ref. = less than BA)	.554	1,507
Age	57.538 (16.326)	1,490
Woman (ref. = man)	.513	1,505

Table S2. Alternative Model Specifications, Log Odds and Standard Errors

	Logit: Suspect Illegality		Conditional Logit: Suspect Illegality		Logit: Notify Law Enforcement		Condition Notify Enforce	Law
	Log Odds	(SE)	Log Odds	(SE)	Log Odds	(SE)	Log Odds	(SE)
Country of Origin (ref. = Mexico)								
El Salvador	.005	(.096)	024	(.176)	.007	(.109)	.259	(.284)
Cuba	107	(.093)	242	(.174)	061	(.102)	438	(.276)
Argentina	141	(.094)	309	(.176)	050	(.108)	548*	(.274)
Haiti	165	(.099)	196	(.180)	.051	(.110)	224	(.283)
Nigeria	166	(.101)	236	(.188)	080	(.112)	018	(.290)
Somalia	126	(.096)	189	(.179)	.110	(.105)	.001	(.278)
Syria	027	(.095)	168	(.181)	.145	(.104)	.197	(.278)
Pakistan	216*	(.101)	111	(.183)	055	(.107)	159	(.275)
India	226*	(.099)	181	(.182)	149	(.109)	377	(.270)
China	263**	(.101)	256	(.180)	206	(.111)	584*	(.281)
Korea	276**	(.101)	464**	(.175)	058	(.107)	330	(.270)
Cambodia	198	(.102)	363*	(.174)	119	(.109)	354	(.290)
Poland	254*	(.099)	492**	(.181)	022	(.107)	156	(.269)
Italy	360***	(.101)	288	(.184)	281*	(.116)	470	(.287)
Canada	283**	(.098)	562**	(.183)	113	(.113)	442	(.288)
Receives Govt. Benefits (ref. = none)								
Welfare	033	(.061)	080	(.107)	.105	(.066)	063	(.165)
SSI	173**	(.061)	256*	(.107)	.034	(.070)	.118	(.170)
Medicaid	173**	(.060)	354***	(.106)	077	(.072)	154	(.166)
Food stamps	096	(.060)	049	(.108)	.114	(.065)	.148	(.165)
EITC	306***	(.061)	532***	(.111)	.013	(.068)	261	(.168)
Police Record (ref. = no record)								
White-collar crime	.509***	(.059)	.599***	(.099)	.580***	(.073)	.822***	(.147)
Violent crime	.790***	(.061)	.960***	(.099)	.911***	(.077)	1.824***	(.158)
Stereotypical immigrant crime	.914***	(.060)	1.054***	(.103)	.871***	(.077)	1.633***	(.160)
Minor infraction	.002	(.058)	113	(.103)	.044	(.072)	041	(.161)
Occupation (ref. = unemployed)		(/		(/		(, , ,		(- /
Low-status, informal	054	(.053)	131	(.101)	099	(.062)	337*	(.160)
Low-status, formal	370***	(.056)	684***	(.100)	234***	(.066)	501***	(.151)
High-status, informal	143*	(.056)	319**	(.100)	204***	(.061)	339*	(.156)
High-status, formal	244***	(.055)	480***	(.099)	157*	(.063)	331*	(.152)
Gender (ref. = man)		(.000)		(.0))	,	(.002)		(.102)
Woman	033	(.035)	037	(.063)	053	(.040)	111	(.099)
Age (ref. = 20)	.000	(.000)	.007	(.005)	.022	(.0.0)		(.0))
30	160**	(.055)	233*	(.101)	.022	(.062)	118	(.149)
40	221***	(.057)	317**	(.101)	131*	(.065)	370*	(.157)
50	207***	(.057)	349***	(.101)	076	(.064)	429**	(.147)
60	206***	(.055)	218*	(.100)	143*	(.063)	530***	(.157)
Years Living in the U.S. (ref. = 1 year)	.200	(.055)	.210	(.100)	.143	(.003)	.550	(.137)
5	098	(.055)	.051	(.102)	058	(.062)	061	(.157)
10	129*	(.053)	034	(.102)	078	(.062)	.155	(.157)
15	129	(.054)	034 088	(.102) $(.100)$	078 046	(.064)	.016	(.156)
20	101 190***	(.057)	088 277**	(.100)	046 145*	(.063)	187	(.150) $(.152)$
English Fluency (ref. = no English)	130	(.030)	2//	(.033)	143	(.003)	10/	(.134)
Adequate English	370***	(.050)	419***	(.091)	284***	(.055)	550***	(.137)
Good, accented English	389***		543***		268***		548***	(.137) $(.140)$
Unaccented English	389*** 392***	(.051) (.051)	545*** 549***	(.089) (.090)	268*** 245***	(.056) (.056)	634***	(.140)
	394	(1001)	J+7	(.030)	243	(.050)	034	(.137)
Education (ref. = less than HS degree)	062	(060)	120	(104)	062	(067)	421*	(160)
High school degree	063 179**	(.060)	129 482***	(.106)	062 070	(.067)	421* 202*	(.169)
Some college	178** 205***	(.060)	482*** 407***	(.112)	079	(.065)	383*	(.174)
College degree	205***	(.061)	407*** 525***	(.109)	086	(.068)	383*	(.170)
Master's degree	328*** 220***	(.060)	535*** 270***	(.109)	044	(.067)	234 240*	(.167)
PhD I continue of Interaction (nof — locking for place to rept on year block)	239***	(.061)	370***	(.108)	093	(.069)	349*	(.168)
Location of Interaction (ref. = looking for place to rent on your block)					400	(100)		
Attempting to board an airplane					438**	(.161)		
Standing outside a home improvement store					795***	(.167)		
Standing outside a welfare office					457**	(.157)		
Walking in your neighborhood					332*	(.150)		
Applying for a job at your place of work					.287*	(.144)		
Constant	.267*	(.123)			846***	(.173)		

Note: Clustered standard errors are in parentheses (except for conditional logit models). Conditional logit models only include profile pairs where one profile is suspected and one is not.

^{*}p < .05; **p < .01 ***p < .001 (two-tailed tests).

Table S3. Global Interaction Models by Partisanship and Education Predicting Suspicions of Illegality, Coefficients and SEs

(Including Leaners) (Including Leaners)	Degree	Less Than College			
Country of Origin (ref. = Mexico)	Degree		College De	5.00	
Country of Origin (ref. = Mexico) El Salvador Cuba	Coef.	(SE)	Coef.	(SE)	
El Salvador Cuba -0.042 (.034) .037 (.031) Cuba -0.019 (.033) .021 (.029) Argentina -0.40 (.032) .017 (.030) Haiti -0.669* (.033) .031 (.031) Nigeria -0.668* (.034) .039 (.032) Somalia -0.500 (.032) .035 (.031) Syria -0.26 (.032) .006 (.031) Pakistan -0.84* (.035) .012 (.031) India -0.94** (.034) .019 (.032) China -0.94** (.034) .019 (.032) Korea -0.96** (.034) .019 (.032) Reabodia -0.65 (.035) .023 (.032) Poland -1.35*** (.035) .001 (.031) Italy -1.30*** (.035) .064 (.030) Receives Govt. Benefits (ref. = none) Welfare SSI -0.01 (.021) .093*** (.019) SSI -0.03 (.020) .051** (.019) Food stamps -0.14 (.021) .093*** (.019) Food stamps -0.14 (.021) .046* (.019) Food stamps -0.04 (.012) .009*** White-collar crime -0.05** (.020) .185*** (.019) Ninor infraction -0.05 (.019) .010** Violent crime -0.063** (.020) .051** Violent crime -0.07** (.019) .046*** (.017) Food on .07** (.019) .046** (.017) Food on .07** (.019) .046** (.017) Food on .07** (.019) .046** (.017) Food on .07** (.019) .040** (.017) Food .000** (.000**					
Argentina	018	(.034)	.022	(.029)	
Haiti	008	(.033)	027	(.028)	
Nigeria Somalia050 (.032)035 (.031) Syria026 (.032)006 (.031) Pakistan084* (.035)012 (.031) India094** (.035)012 (.031) India094** (.033)026 (.032) China102** (.033)026 (.032) China102** (.033)026 (.032) Korea096** (.034)049 (.030) Cambodia065 (.035)023 (.032) Poland135*** (.035)023 (.032) Poland135*** (.035)001 (.031) Italy130*** (.034)037 (.031) Canada089** (.033)061* (.030) Receives Govt. Benefits (ref. = none) Welfare SSI001 (.021)093*** (.019) Medicaid030 (.020)051** (.019) Medicaid030 (.020)051** (.019) Food stamps014 (.021)046* (.019) FITC059** (.021)089** (.019) Police Record (ref. = no record) White-collar crime120*** (.020) .103*** (.017) Violent crime216*** (.020) .103*** (.017) Violent crime216*** (.020) .185*** (.019) Minor infraction005 (.019)010 (.015) Occupation (ref. = unemployed) Low-status, informal063** (.020)064*** (.017) High-status, formal063** (.020)004 (.017) High-status, formal087*** (.019)064** (.017) High-status, formal087*** (.019)026 (.018) Gender (ref. = man) Woman004 (.012)020 (.011) Age (ref. = 20) 30049** (.018)004 (.018) 50037* (.019)041* (.017) Years Living in the U.S. (ref. = 1 year) 5025 (.019)037* (.017) 10024 (.019)042* (.017) 15032 (.020)021 (.017) English Fluency (ref. = no English) Adequate English122*** (.018)056*** (.016) Good, accented English122*** (.018)056*** (.016)	051	(.033)	014	(.028)	
Somalia	046	(.034)	030	(.030)	
Syria 026 (.032) 006 (.031) Pakistan 084* (.035) 012 (.031) India 094** (.034) 019 (.032) China 102** (.033) 026 (.032) Korea 096** (.034) 049 (.030) Cambodia 065 (.035) 023 (.032) Poland 135*** (.035) 001 (.031) Italy 130*** (.034) 007 (.031) Canada 089** (.033) 061* (.030) Receives Govt. Benefits (ref. = none) Welfare .054* (.023) 064*** (.019) SSI .001 (.021) 093*** (.019) SSI .001 (.021) 093*** (.019) Food stamps .014 (.021) 089*** (.019 FITC 059*** (.021) 089*** (.019 Violent crime	059	(.035)	018	(.031)	
Pakistan 084* (.035) 012 (.031) India 094** (.034) 019 (.032) China 102** (.033) 026 (.032) Korea 096** (.034) 049 (.030) Cambodia 065 (.035) 023 (.032) Poland 135*** (.035) 001 (.031) Italy 130*** (.034) 037 (.031) Canada 089** (.033) 061* (.030) Receives Govt. Benefits (ref. = none) Welfare (.054* (.023) 064*** (.019) SSI .001 (.021) 093*** (.019) SSI .001 (.021) 093*** (.019) Food stamps .014 (.021) 046** (.019) Fold stamps .014 (.021) 049*** (.019) Police Record (ref. = no record) White-collar crime .120**** (.020) .103****	036	(.033)	022	(.029)	
India	.033	(.033)	030	(.029)	
China 102** (.033) 026 (.032) Korea 096** (.034) 049 (.030) Cambodia 065 (.035) 023 (.032) Poland 135*** (.034) 037 (.031) Italy 130*** (.034) 037 (.031) Canada 089** (.033) 061* (.030) Receives Govt. Benefits (ref. = none) Welfare .054* (.023) 064**** (.019) SSI .001 (.021) 093*** (.019) Medicaid 030 (.020) 051** (.019) Food stamps .014 (.021) 046* (.019) FOR eccord (ref. = no record) White-collar crime .120*** (.020) .103*** (.019) White-collar crime .120*** (.020) .103*** (.017) Violent crime .216*** (.020) .185*** (.019) Stereotypical immigrant crime .216***	049	(.034)	045	(.030)	
Korea	057	(.034)	043	(.029)	
Cambodia 065 (.035) 023 (.032) Poland 155*** (.035) 001 (.031) Italy 130*** (.034) 037 (.031) Canada 089** (.033) 061* (.030) Receives Govt. Benefits (ref. = none) Welfare .054* (.023) 064*** (.019) SSI .001 (.021) 093*** (.019) Medicaid 030 (.020) 051** (.019) Frod stamps .014 (.021) 046* (.019) EITC 059** (.021) 089*** (.019) Police Record (ref. = no record) White-collar crime .120**** (.020) .103*** (.019) White-collar crime .120**** (.020) .185*** (.019) Stereotypical immigrant crime .216*** (.020) .185*** (.019) Stereotypical immigrant crime .216*** (.020) .185*** (.019) Low-status, informal <td>053</td> <td>(.036)</td> <td>062*</td> <td>(.029)</td>	053	(.036)	062*	(.029)	
Poland	063	(.033)	057	(.031)	
Italy	052	(.035)	037	(.031)	
Canada 089** (.033) 061* (.030) Receives Govt. Benefits (ref. = none) Welfare .054* (.023) 064*** (.019) SSI .001 (.021) 093*** (.019) Medicaid 030 (.020) 051** (.019) Food stamps .014 (.021) 046* (.019) EITC 059** (.021) 089*** (.019) Police Record (ref. = no record) White-collar crime .120*** (.020) .103*** (.017) Violent crime .201*** (.020) .185*** (.019) Stereotypical immigrant crime .216*** (.020) .185*** (.019) Minor infraction .005 (.019) 010 (.015) Occupation (ref. = unemployed) Low-status, informal 024 (.018) 001 (.017) Ligh-status, informal 063** (.020) 004** (.017) High-status, informal 087*** (.019) 026	049	(.033)	064*	(.030)	
Receives Govt. Benefits (ref. = none) Welfare .054* (.023) 064*** (.019) SSI .001 (.021) 093*** (.019) Medicaid 030 (.020) 051*** (.019) Food stamps .014 (.021) 089*** (.019) ETTC 059** (.021) 089*** (.019) Police Record (ref. = no record) White-collar crime .120*** (.020) .103**** (.017) Violent crime .201*** (.021) .147*** (.019) Stereotypical immigrant crime .216*** (.020) .185*** (.019) Minor infraction .005 (.019) 010 (.015) Occupation (ref. = unemployed)	104**	(.034)	060*	(.029)	
Welfare .054* (.023) 064*** (.019) SSI .001 (.021) 093*** (.019) Medicaid 030 (.020) 051*** (.019) Food stamps .014 (.021) 089*** (.019) ETTC 059** (.021) 089*** (.019) Police Record (ref. = no record) White-collar crime .120*** (.020) .103**** (.017) Violent crime .201*** (.021) .147**** (.019) Stereotypical immigrant crime .216*** (.020) .185**** (.019) Minor infraction .005 (.019) 010 (.015) Occupation (ref. = unemployed) .005 (.019) 010 (.015) Low-status, informal 024 (.018) 001 (.017) Low-status, formal 111**** (.019) 064*** (.017) High-status, formal 063** (.020) 004 (.017) High-status, formal	069*	(.032)	070*	(.030)	
SSI .001 (.021) 093*** (.019) Medicaid 030 (.020) 051** (.019) Food stamps .014 (.021) 046* (.019) EITC 059** (.021) 089*** (.019) Police Record (ref. = no record) White-collar crime .120*** (.020) .103*** (.017) White-collar crime .201*** (.020) .185*** (.019) Stereotypical immigrant crime .216*** (.020) .185*** (.019) Minor infraction .005 (.019) 010 (.015) Occupation (ref. = unemployed) Low-status, informal 024 (.018) 001 (.015) Low-status, formal 111*** (.019) 064*** (.017) High-status, formal 063** (.020) 004 (.017) High-status, formal 087*** (.019) 026 (.018) Gender (ref. = man) Woman 004 (.012) 020 (.011) <td></td> <td></td> <td></td> <td></td>					
Medicaid 030 (.020) 051** (.019) Food stamps .014 (.021) 046* (.019) EITC 059** (.021) 089*** (.019) Police Record (ref. = no record) White-collar crime .120*** (.020) .103*** (.017) White-collar crime .201*** (.020) .185*** (.019) Stereotypical immigrant crime .216*** (.020) .185*** (.019) Minor infraction .005 (.019) 010 (.015) Occupation (ref. = unemployed) Low-status, informal 024 (.018) 001 (.017) Low-status, formal 111*** (.019) 064*** (.017) High-status, formal 063** (.020) 004 (.017) High-status, formal 087*** (.019) 026 (.018) Gender (ref. = man) Woman 004 (.012) 020 (.011) Age (ref. = 20) 30 049** (.018)	.020	(.022)	029	(.018)	
Food stamps EITC 059** (.021) 046** (.019) Police Record (ref. = no record) White-collar crime .120*** (.020) Stereotypical immigrant crime .216*** (.020) Stereotypical immigrant crime .216*** (.020) Stereotypical immigrant crime .216*** (.020) .185*** (.019) Stereotypical immigrant crime .005 (.019) 010 (.015) Occupation (ref. = unemployed) Low-status, informal 024 (.018) 001 (.017) Low-status, formal 111*** (.019) 064*** (.017) High-status, informal 063** (.020) 004 (.017) High-status, formal 087*** (.019) 026 (.018) Gender (ref. = man) Woman 004 004 (.012) 020 (.011) Age (ref. =20) 30 049** 30 049** (.018) 033 (.019) 041* (.018) 50 037* (.019) 041* (.018) 50 037* (.019) 041* (.017) Years Living in the U.S. (ref. = 1 year) 5 025 037 (.019) 041* (.017) Years Living in the U.S. (ref. = 1 year) 5 025 037 (.019) 040* (.017) 15 024 (.019) 040* (.017) English Fluency (ref. = no English) Adequate English 122*** (.018) 045** (.016) Unaccented English 122*** (.018) 045** (.016)	013	(.021)	061***	(.018)	
EITC 059** (.021) 089*** (.019) Police Record (ref. = no record) White-collar crime .120*** (.020) .103*** (.017) Violent crime .201*** (.021) .147*** (.019) Stereotypical immigrant crime .216*** (.020) .185*** (.019) Minor infraction .005 (.019) 010 (.015) Occupation (ref. = unemployed) .005 (.018) 001 (.017) Low-status, informal 024 (.018) 001 (.017) Low-status, informal 063** (.020) 004 (.017) High-status, formal 063** (.020) 004 (.017) High-status, formal 087*** (.019) 026 (.018) Gender (ref. = man) Woman 004 (.012) 020 (.011) Age (ref. = 20) 30 049** (.018) 004 (.018) 50 037* (.019) 041** (.017) <td>042*</td> <td>(.021)</td> <td>039*</td> <td>(.018)</td>	042*	(.021)	039*	(.018)	
Police Record (ref. = no record) White-collar crime	001	(.022)	041*	(.017)	
White-collar crime .120*** (.020) .103*** (.017) Violent crime .201*** (.021) .147*** (.019) Stereotypical immigrant crime .216*** (.020) .185*** (.019) Minor infraction .005 (.019) 010 (.015) Occupation (ref. = unemployed) .005 (.018) 001 (.017) Low-status, informal 111*** (.019) 064*** (.017) High-status, formal 063** (.020) 004 (.017) High-status, formal 087*** (.019) 026 (.018) Gender (ref. = man) Woman 004 (.012) 020 (.011) Age (ref. = 20) 30 049** (.018) 004 (.018) 40 033 (.019) 041** (.018) 50 037* (.019) 037* (.017) 60 037* (.019) 041* (.017) Years Living in the U.S. (ref. = 1 year) <td>077***</td> <td>(.021)</td> <td>063***</td> <td>(.018)</td>	077***	(.021)	063***	(.018)	
Violent crime .201*** (.021) .147*** (.019) Stereotypical immigrant crime .216*** (.020) .185*** (.019) Minor infraction .005 (.019) 010 (.015) Occupation (ref. = unemployed) .005 (.018) 001 (.017) Low-status, informal 024 (.018) 001 (.017) High-status, formal 063** (.020) 004 (.017) High-status, formal 087*** (.019) 026 (.018) Gender (ref. = man) Woman 004 (.012) 020 (.011) Age (ref. =20) 30 049** (.018) 004 (.018) 40 033 (.019) 041* (.018) 50 037* (.019) 037* (.017) 60 037* (.019) 041* (.017) Years Living in the U.S. (ref. = 1 year) 5 025 (.019) 041* (.017) 15	a a a distributi	(000)	4 O Salvabada	(010)	
Stereotypical immigrant crime .216*** (.020) .185*** (.019) Minor infraction .005 (.019) 010 (.015) Occupation (ref. = unemployed)	.111***	(.020)	.106***	(.016)	
Minor infraction .005 (.019) 010 (.015) Occupation (ref. = unemployed) 024 (.018) 001 (.017) Low-status, informal 111*** (.019) 064*** (.017) High-status, informal 063** (.020) 004 (.017) High-status, formal 087*** (.019) 026 (.018) Gender (ref. = man) Woman 004 (.012) 020 (.011) Age (ref. =20) 30 049** (.018) 004 (.018) 40 033 (.019) 041* (.018) 50 037* (.019) 037* (.017) 60 037* (.019) 041* (.017) Years Living in the U.S. (ref. = 1 year) 5 025 (.019) 023 (.017) 10 024 (.019) 040* (.017) 15 025 (.019) 042* (.017) 20 047* (.019)	.197***	(.020)	.162***	(.018)	
Occupation (ref. = unemployed) 024 (.018) 001 (.017) Low-status, informal 111*** (.019) 064*** (.017) High-status, informal 063** (.020) 004 (.017) High-status, formal 087*** (.019) 026 (.018) Gender (ref. = man) 004 (.012) 020 (.011) Age (ref. =20) 30 049** (.018) 004 (.018) 40 033 (.019) 041* (.018) 50 037* (.019) 037* (.017) 60 037* (.019) 041* (.017) Years Living in the U.S. (ref. = 1 year) 5 025 (.019) 041* (.017) 10 024 (.019) 040* (.017) 15 032 (.020) 021 (.017) 15 032 (.020) 021 (.017) 15 047* (.019) 042*	.197***	(.021)	.216***	(.017)	
Low-status, informal 024 (.018) 001 (.017) Low-status, formal 111*** (.019) 064*** (.017) High-status, informal 063** (.020) 004 (.017) High-status, formal 087*** (.019) 026 (.018) Gender (ref. = man) 004 (.012) 020 (.011) Age (ref. =20) 30 049** (.018) 004 (.018) 40 033 (.019) 041* (.018) 50 037* (.019) 037* (.017) 60 037* (.019) 041* (.017) Years Living in the U.S. (ref. = 1 year) 5 025 (.019) 041* (.017) 10 024 (.019) 040* (.017) 15 032 (.020) 021 (.017) 15 032 (.020) 021 (.017) 15 047* (.019) 042* (.017)	.010	(.019)	007	(.015)	
$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$	010	(010)	006	(016)	
High-status, informal	018 076***	(.019)	006 085***	(.016)	
High-status, formal 087*** (.019) 026 (.018) Gender (ref. = man) 004 (.012) 020 (.011) Age (ref. = 20) 049** (.018) 004 (.018) 30 033 (.019) 041* (.018) 40 033 (.019) 041* (.018) 50 037* (.019) 037* (.017) 60 037 (.019) 041* (.017) Years Living in the U.S. (ref. = 1 year) 5 025 (.019) 041* (.017) 10 024 (.019) 040* (.017) 15 032 (.020) 021 (.017) 20 047* (.019) 042* (.017) English Fluency (ref. = no English) 047* (.018) 045** (.016) Adequate English 122*** (.018) 056*** (.016) Good, accented English 122*** (.018) 045** (.016) Unaccented English 116*** (.017) 056***	076**** 036	(.019)		(.016) (.016)	
Gender (ref. = man) Woman	036 049*	(.020) (.019)	031 060***	(.016)	
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	049	(.019)	000	(.010)	
Age (ref. =20) 30	014	(.012)	.001	(.010)	
30	014	(.012)	.001	(.010)	
40 033 (.019) 041* (.018) 50 037* (.019) 037* (.017) 60 037 (.019) 041* (.017) Years Living in the U.S. (ref. = 1 year) 025 (.019) 023 (.017) 10 024 (.019) 040* (.017) 15 032 (.020) 021 (.017) 20 047* (.019) 042* (.017) English Fluency (ref. = no English) 047* (.018) 045** (.016) Good, accented English 122*** (.018) 045** (.016) Unaccented English 116*** (.017) 056*** (.016)	023	(.020)	045**	(.016)	
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	029	(.020)	064***	(.016)	
60	040*	(.019)	053***	(.016)	
Years Living in the U.S. (ref. = 1 year) $ \begin{array}{ccccccccccccccccccccccccccccccccccc$	043*	(.020)	051**	(.016)	
$ \begin{array}{cccccccccccccccccccccccccccccccccccc$.0.2	(1020)	.001	(.010)	
10 024 (.019) 040* (.017) 15 032 (.020) 021 (.017) 20 047* (.019) 042* (.017) English Fluency (ref. = no English) 122*** (.018) 056*** (.016) Good, accented English 122*** (.018) 045** (.016) Unaccented English 116*** (.017) 056*** (.016)	022	(.019)	022	(.016)	
15	026	(.019)	030	(.016)	
20 047* (.019) 042* (.017) English Fluency (ref. = no English) 122*** (.018) 056*** (.016) Adequate English 122*** (.018) 045** (.016) Good, accented English 116*** (.017) 056*** (.016) Unaccented English 116*** (.017) 056*** (.016)	036	(.020)	014	(.016)	
English Fluency (ref. = no English) Adequate English Good, accented English Unaccented English122***122*** (.018)045** (.016)045** (.016)16*** (.017)056***	051**	(.020)	034*	(.016)	
Adequate English 122*** (.018) 056*** (.016) Good, accented English 122*** (.018) 045** (.016) Unaccented English 116*** (.017) 056*** (.016)		` /		, ,	
Good, accented English	079***	(.017)	089***	(.015)	
	088***	(.017)	090***	(.015)	
Education (ref. = less than HS degree)	090***	(.018)	086***	(.015)	
High school degree .002 (.021)034 (.019)	032	(.020)	003	(.018)	
Some college024 (.021)063*** (.019)	068**	(.021)	025	(.018)	
College degree047* (.021)058** (.019)	072***	(.022)	029	(.018)	
Master's degree083*** (.021)088*** (.019)	063**	(.021)	084***	(.017)	
PhD058** (.021)054** (.019)	070***	(.021)	046*	(.018)	
Constant .676*** (.041) .461*** (.041)	.632***	(.042)	.498***	(.038)	
Observations 6,741 6,741 7,162 7,162	6,722	6,722	8,377	8,377	
R-squared .069 .069 .048 .048	.051	.051	.059	.059	

Note: Clustered standard errors are in parentheses. *p < .05; **p < .01 ***p < .001 (two-tailed tests).

Table S4. Global Interaction Models by Respondent Region Predicting Suspicions of Illegality, Coefficients and SEs

	Midwest	Region	Northeast	Region	South R	egion	West	Region
	Coef.	(SE)	Coef.	(SE)	Coef.	(SE)	Coef.	(SE)
Country of Origin (ref. = Mexico)	2361.	(52)	2001.	(52)		(52)	2301.	(52)
El Salvador	024	(.045)	017	(.053)	.039	(.038)	018	(.047)
Cuba	.0024	(.043)	017 013	(.053)	008	(.038)	097*	(.047)
	029		013 073		008 028		.001	, ,
Argentina		(.042)		(.056)		(.036)		(.046)
Haiti	048	(.043)	.010	(.054)	018	(.040)	087	(.048)
Nigeria	012	(.044)	.025	(.058)	033	(.040)	126*	(.050)
Somalia	031	(.046)	062	(.045)	.001	(.039)	036	(.047)
Syria	.007	(.043)	047	(.052)	.021	(.037)	032	(.050)
Pakistan	036	(.045)	040	(.052)	019	(.040)	102*	(.048)
India	054	(.045)	022	(.051)	037	(.040)	094*	(.047)
China	063	(.042)	091	(.052)	044	(.041)	048	(.052)
Korea	105*	(.044)	005	(.050)	022	(.041)	111*	(.050)
Cambodia	007	(.047)	032	(.056)	029	(.039)	121*	(.050)
Poland	013	(.045)	100	(.056)	023	(.038)	152**	(.047)
Italy	076	(.044)	095	(.053)	051	(.039)	113*	(.049)
Canada	045	(.043)	102*	(.048)	057	(.038)	060	(.051)
Receives Govt. Benefits (ref. = none)	043	(.043)	102	(.0+0)	037	(.030)	000	(.031)
Welfare	012	(027)	001	(024)	003	(025)	013	(020)
		(.027)	.001	(.034)		(.025)		(.029)
SSI	045	(.026)	079**	(.030)	025	(.024)	025	(.031)
Medicaid	076**	(.027)	020	(.031)	029	(.024)	033	(.029)
Food stamps	037	(.026)	028	(.035)	013	(.025)	018	(.027)
EITC	087***	(.026)	114***	(.032)	039	(.023)	053	(.030)
Police Record (ref. = no record)								
White-collar crime	.110***	(.026)	.147***	(.029)	.093***	(.021)	.123***	(.028)
Violent crime	.194***	(.026)	.213***	(.031)	.180***	(.024)	.143***	(.028)
Stereotypical immigrant crime	.188***	(.025)	.250***	(.031)	.204***	(.023)	.220***	(.028)
Minor infraction	021	(.022)	.045	(.027)	008	(.021)	.017	(.024)
Occupation (ref. = unemployed)		(, -)		(/		(- /		()
Low-status, informal	017	(.024)	043	(.027)	012	(.021)	.019	(.027)
Low-status, formal	091***	(.024)	115***	(.028)	092***	(.022)	034	(.025)
High-status, informal	007	(.026)	069*	(.028)	057*	(.022)	.000	(.028)
High-status, formal	034	(.024)	077*	(.020)	076***	(.022)	026	(.026)
Gender (ref. = man)	034	(.024)	077	(.030)	070	(.022)	020	(.020)
	006	(015)	004	(010)	002	(014)	02.4*	(017)
Woman	.006	(.015)	004	(.018)	002	(.014)	034*	(.017)
Age (ref. $= 20$)	000	(000)	0.71	(004)	0.45	(000)	0.504	(005)
30	020	(.023)	071*	(.031)	017	(.022)	062*	(.027)
40	035	(.024)	129***	(.029)	036	(.023)	030	(.029)
50	067**	(.022)	123***	(.031)	001	(.022)	035	(.027)
60	058*	(.023)	093**	(.031)	029	(.022)	029	(.027)
Years Living in the U.S. (ref. $= 1$ year)								
5	.010	(.024)	001	(.030)	043*	(.022)	049	(.027)
10	.007	(.024)	040	(.031)	051*	(.021)	035	(.027)
15	027	(.025)	029	(.032)	032	(.021)	.004	(.029)
20	014	(.025)	035	(.032)	061**	(.021)	058*	(.027)
English Fluency (ref. = no English)		()		()		()		()
Adequate English	071**	(.022)	082**	(.027)	074***	(.019)	118***	(.027)
Good, accented English	071 077***	(.022)	052*	(.027)	075***	(.020)	149***	(.026)
_	065**			` ′	088***	(.020)	146*** 146***	
Unaccented English	003***	(.022)	056*	(.027)	088****	(.019)	140	(.026)
Education (ref. = less than HS degree)	002	(004)	010	(000)	017	(000)	016	(000)
High school degree	003	(.024)	013	(.033)	017	(.023)	016	(.033)
Some college	012	(.026)	059	(.034)	063**	(.023)	014	(.032)
College degree	030	(.026)	095**	(.034)	049*	(.024)	011	(.030)
Master's degree	079**	(.025)	043	(.035)	087***	(.023)	061*	(.030)
PhD	025	(.025)	063	(.034)	082***	(.024)	031	(.030)
Constant	.517***	(.053)	.582***	(.062)	.558***	(.049)	.570***	(.067)
Observations	4,186	4,186	2,579	2,579	5,124	5,124	3,220	3,220
<i>R</i> -squared	.058	.058	.075	.075	.053	.053	.065	.065
Note: Clustered standard errors are in t								

Note: Clustered standard errors are in parentheses. *p < .05; **p < .01 ***p < .001 (two-tailed tests).

Design Appendix

Introductory Text: Next, identifying immigrants who are in the United States illegally is a government priority. In order to help us better understand this process, you will be presented with pairs of profiles describing different immigrants currently living in the United States. For each pair of profiles, please look at the information carefully, and then indicate whether you suspect each immigrant is in the U.S. illegally. Even if you aren't entirely sure, please indicate your best guess about the legal status of each immigrant.

Dependent Variable 1: Do you think these immigrants are illegal/undocumented? In other words, do you think they lack proper documentation status to be in the United States?

Response Options: Yes, No

Dependent Variable 2: Many local law enforcement organizations are asking for citizens to assist them in their efforts to identify immigrants living in the U.S. illegally.

If you saw these immigrants [location of interaction treatment], would you consider asking local law enforcement to investigate their legal statuses?

Response Options: Yes, No

See Table 1 in the main text for all randomizations

Survey Experiment Screen Shots

Figure S1. Screen Shot of Survey Experiment Introduction



Next, identifying immigrants who are in the United States illegally is a government priority. In order to help us better understand this process, you will be presented with pairs of profiles describing different immigrants currently living in the United States. For each pair of profiles, please look at the information carefully, and then indicate whether you suspect each immigrant is in the U.S. illegally. Even if you aren't entirely sure, please indicate your best guess about the legal status of each immigrant.

You will receive the following categories of information about each immigrant:



Figure S2. Screen Shot of Dependent Variable 1 and Presentation of (Randomized) Profiles



Please read the two profiles carefully. Then, please indicate whether you suspect each immigrant is living in the United States illegally.

	Immigrant 1	Immigrant 2
Education	Did not finish high school	College degree
Recipient of	Welfare	Welfare
Government Benefits		
National Origin	Nigeria	El Salvador
Current Area of	Within 10 miles of southern	Within 100 miles of
Residence	border	southern border
Occupation	Day laborer	Freelance graphic designer
Years Since Arrival in	20 years	1 year
the U.S.		
Gender	Woman	Woman
Police Record	Littering	No police record
English Proficiency	Good, accented English	Adequate English
Age	60	50

Do you think these immigrants are illegal/undocumented? In other words, do you think they lack proper documentation status to be in the United States?

	Yes	No
Immigrant 1	0	0
Immigrant 2	0	0

Figure S3. Screen Shot of Dependent Variable 3 with Randomized Location of Interaction Treatments and Presentation of (Randomized) Profiles



	Immigrant 1	Immigrant 2
Education	Did not finish high school	College degree
Recipient of Government Benefits	Welfare	Welfare
National Origin	Nigeria	El Salvador
Current Area of Residence	Within 10 miles of southern border	Within 100 miles of southern border
Occupation	Day laborer	Freelance graphic designer
Years Since Arrival in the U.S.	20 years	1 year
Gender	Woman	Woman
Police Record	Littering	No police record
English Proficiency	Good, accented English	Adequate English
Age	60	50

Many local **law enforcement organizations** are asking for citizens to assist them in their efforts to identify immigrants living in the U.S. illegally.

If you saw these immigrants standing outside of a welfare office, would you consider asking **local law enforcement** to investigate their legal statuses?

	Yes	No
Immigrant 1	0	0
Immigrant 2	0	0