

*Eunice Kennedy Shriver* National Institute of Child Health and Human Development

# First Year of Driving: The Effect of Passenger on Teens' Crash/Near-Crash Risk

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# Teen Drivers -Vulnerable Road Users

Teen drivers are over represented in crashes statistics

- In 2015, 5% of the licensed U.S. drivers were teen drivers
  - 12% of all police-reported crashes
  - 9% of total fatal crashes
- 1,886 teen drivers (15-20yrs.) were killed and 195,000 injured in 2015

(NHTSA, 2017)

# Driving Independently

- Driving SOLO with some restrictions
- Elevated crash risk after obtaining provisional licensure
  - Highest for the youngest
- Increase in risky driving behaviors
  - Distracted driving
  - Speeding

(Curry et al., 2017; Gershon et al., 2018)

#### Contributing Factors to Teens' Over Involvement in Crashes



## Research Goals & Contributions

Identify predictors for teens crash/near-crash (CNC) rates during the first year of independent driving

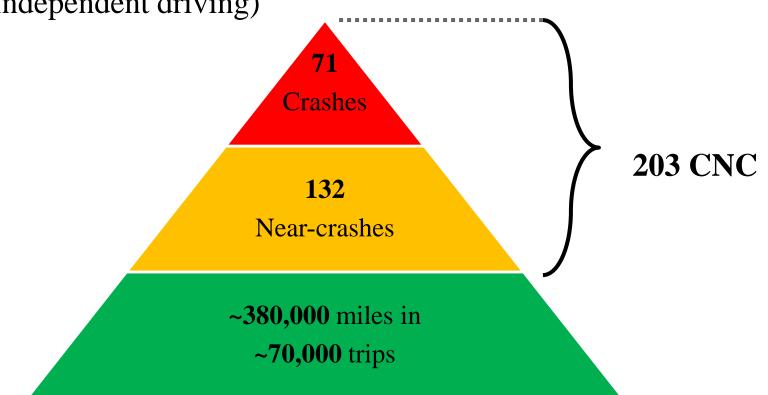
- Demographics measures (e.g. gender, vehicle access)
- Environmental measures (e.g. time of day, road condition)
- Psychosocial measures (e.g. sensation seeking, parental trust)

#### **Study Contributions:**

- Naturalistic data set
- Using CNC as an outcome measure of risky driving
- Prolonged data collection

# The Supervised Practice Driving Study

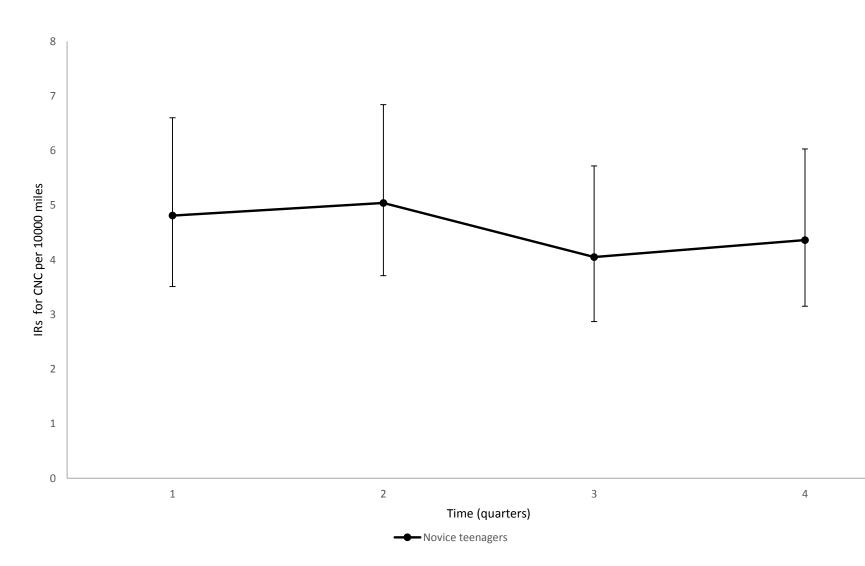
- 82 newly-licensed teens (16.44 yrs., SD=.32) 53% females
- Data collection period 2010-2014 (up to 12 months independent driving)



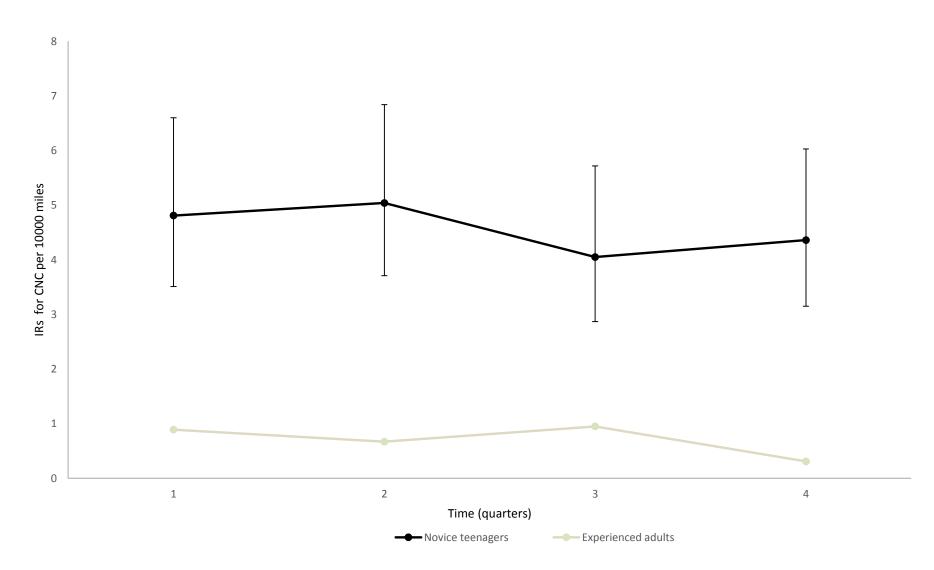
## Method

- In-vehicle Data Acquisition System (DAS) included:
- Multi-axis accelerometer
- Global Positioning System
   (GPS)
- Video cameras monitoring driver's face, hand and body positioning, driver's forward,
   rear views, and the car dashboard

#### Results: Crash/Near-Crash Rates



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	Factor	IRR	95% CI
	Gender		
Results	Male Vs. Female	1.5	0.98-2.31
	Vehicle ownership		
	Owned vs. Shared	1.18	0.75-1.85
	Passengers		
	Yes vs. None	0.6	0.44-0.81
	Adult vs. None	0.28	0.15-0.54
	Teen vs. None	0.64	0.50-0.83
	Time of day		
	Night Vs. Day	1.23	0.90-1.67
	Weather condition		
	Wet vs. Dry	0.78	0.49-1.25
	Self reported risky driving		
	High vs. Low	0.55	0.06-4.97
	Risk perception		
	High Vs. Low	0.51	0.07-3.89
	Sensation seeking		
	High Vs. Low	1.81	0.28-11.84
	Friends substance use		
	High Vs. Low	2.01	0.36-11.35
	Friends risky driving		
	High Vs. Low	1.76	0.31-10.20
	Parents trust		
	High Vs. Low	0.97	0.07-1.95
	Parents knowledge		
	High Vs. Low	1.77	0.17-18.81
	Parents limits		
	High Vs. Low	0.81	0.13-5.21

#### Results

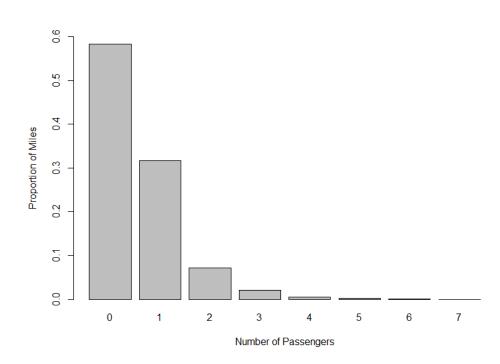
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• Miles driven by passenger type

Passenger Type	Alone	Teen	Adult	child
Total miles (%)	60	30	8	2

#### Results

• Number of passengers



• Teen driver and passenger gender

		Passenger's gender	
		Female	Male
Driver's	Female (miles %)	64	36
gender	Male (miles %)	44	56

#### **Discussion and Conclusions**

- Passenger presence was associated with decreased crash/near-crash rates among novice teen drivers.
  - Adult passenger was associated with dramatic decrease in CNC rates.
  - Teen passenger was also associated with some decrease in CNC rates
- The effect of teen passenger is somewhat inconsistent across studies.
  - Passenger gender, number of passengers, and other passenger characteristics should be examined

## Limitations and Challenges

- Naturalistic driving studies provide extremely rich data
  Passenger data were coded at the beginning and end of a trip
- Passenger related attributes can change during the trip
  - Presence, type, number, and gender
  - Front seat vs. back seat passengers
- The approach we are considering is to work on the CNC dataset and randomly sampled baselines.

# Thank You!

#### Collaborators

- Bruce Simons-Morton, NICHD/DIPHR/SBSB
- Kellienne R. Sita, NICHD/DIPHR/SBSB
- Johnathon P. Ehsani, Johns Hopkins Bloomberg School of Public Health
- Chunming Zhu, NICHD/DIPHR
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