

# Effects of Playing High School Football on Mental Health in Early Adulthood: An Observational Study

Joint work with Sameer Deshpande\*, Jordan Weiss, and Dylan Small\*\*

\*First Author; \*\* Principal Investigator

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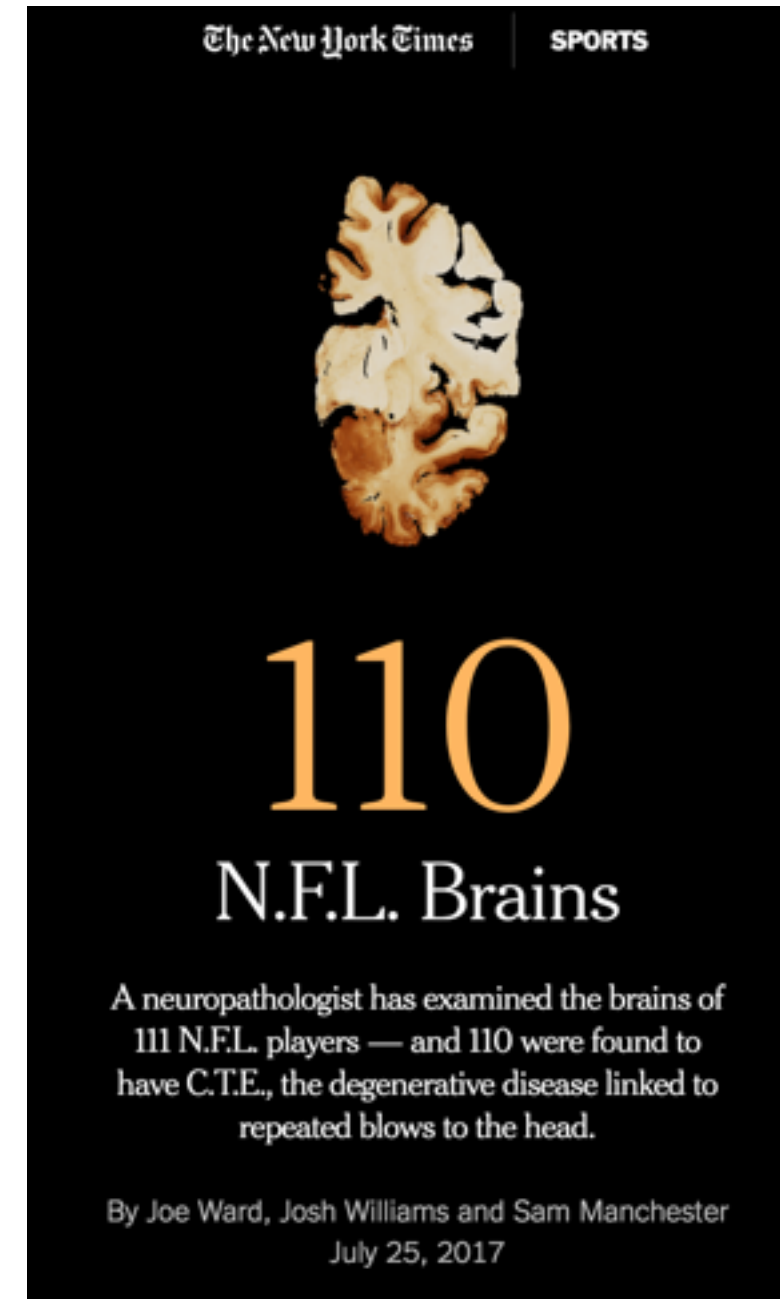
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- **Increasing public awareness:** High profile studies and news coverage linking Chronic Traumatic Encephalopathy (CTE) and football has brought the issue of safety in HS football to the fore.
- **Current evidence:** remains inconclusive...

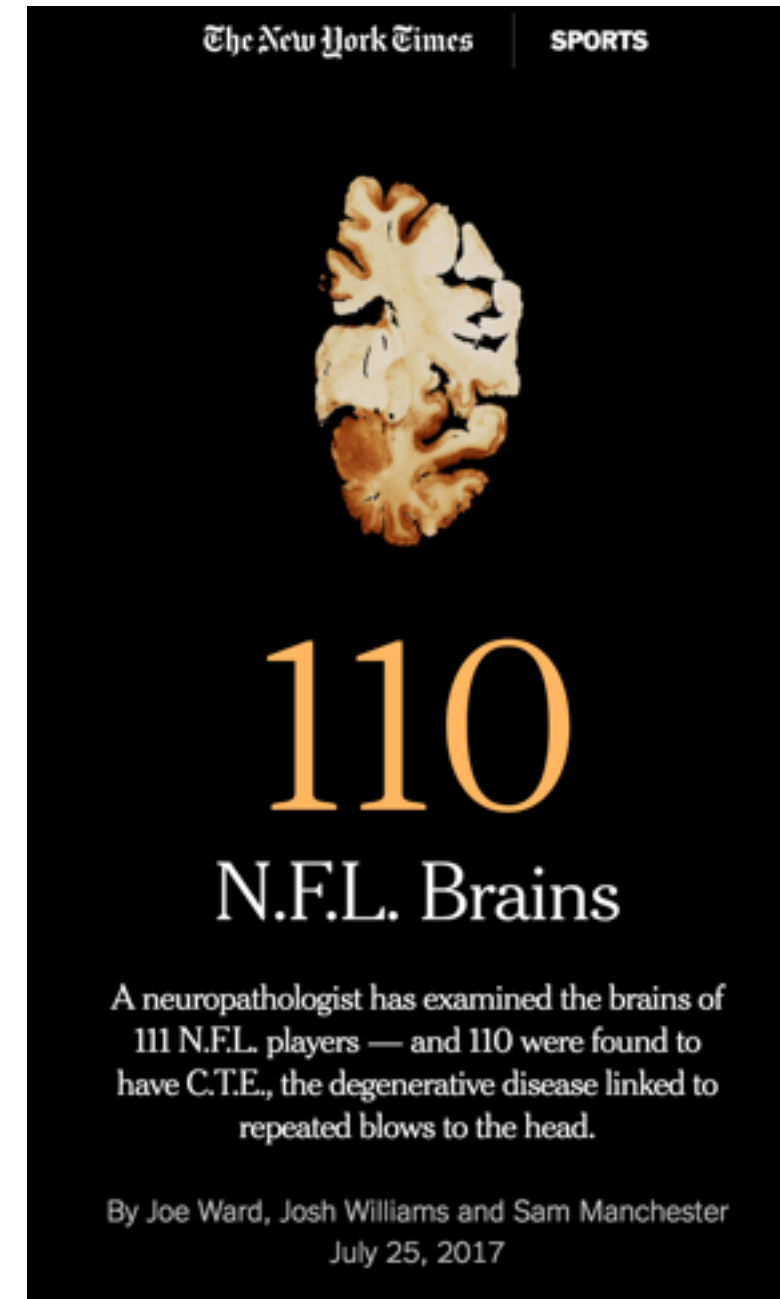
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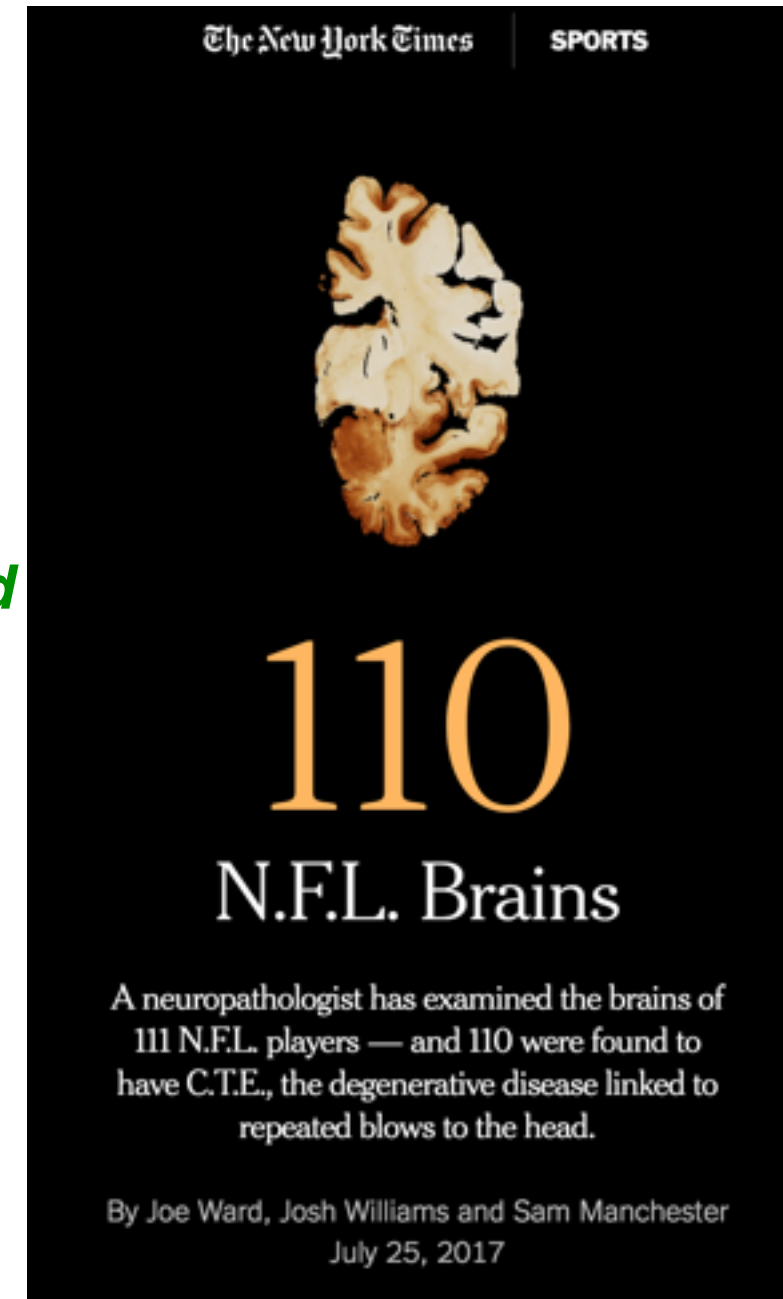
- **Boston University Brain Bank Study (BU):** “CTE was neuropathologically diagnosed in 177 players across all levels of play [among a convenience sample of 202 deceased players of American football] from a brain donation program, including 110 of 111 NFL players,” (McKee et al., 2017).





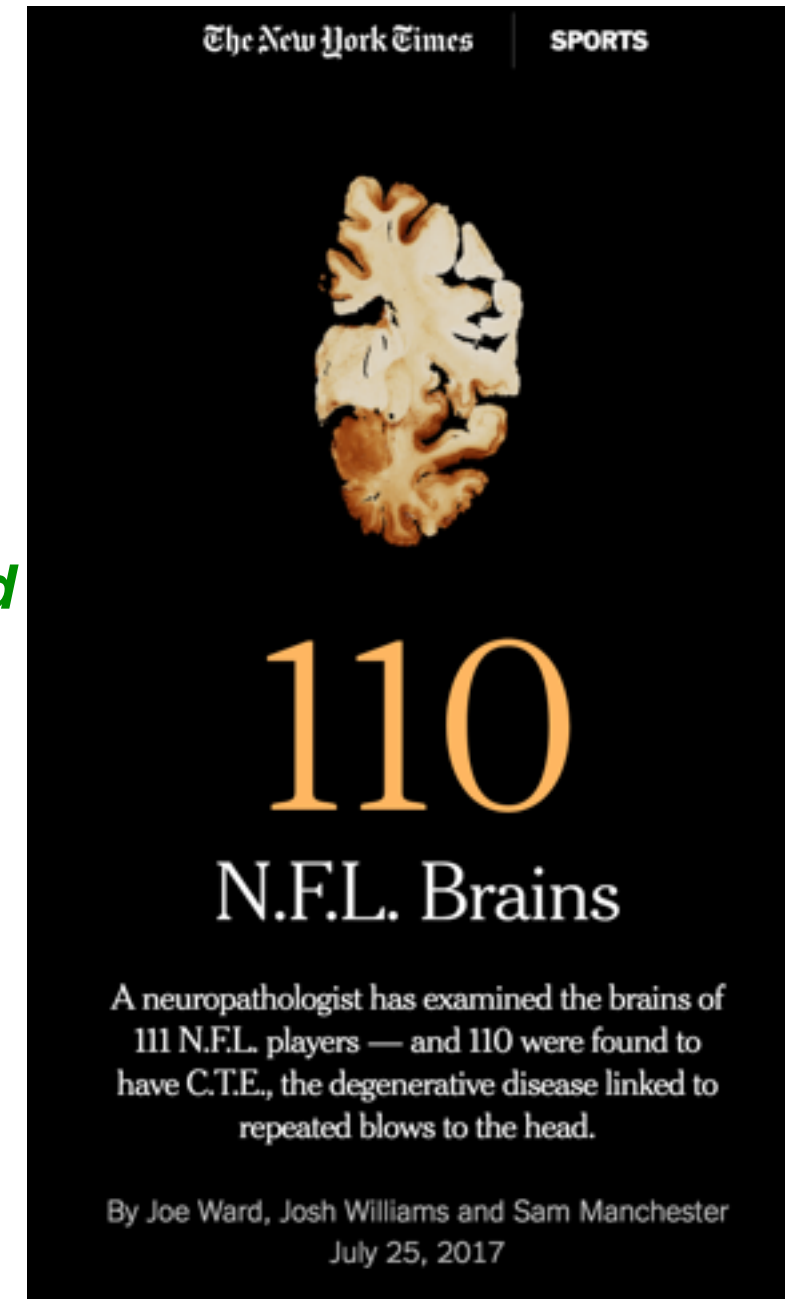
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  - (1) **Severe referral bias — donation to brain bank associated with dementia and depression status**, (2) *neuropathology overlaps broadly with several neurodegenerative comorbidities (Schwab and Hazrati, 2018), (3) Does not generalize well to most common levels of exposure (i.e. youth and HS football).*



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**A Minnesota and  
Wisconsin Collaboration**



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  - (1) **No adjustment for important confounders** (e.g. adolescent IQ, general health, etc.), (2) *relevance of exposure to current players*, (3) *no measure of dose.*



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  - (1) *Relevance of exposure to current players*, (2) **Sensitivity of cognitive outcome measures to mTBI-related disease**, (3) *no measure of dose*.



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- **Exposure:** indicated in Wave I questionnaire that they participated or intended to participate in HS football (n=661 men in core Add Health sample).
- **Control:** 2,216 control subjects, 808 participated or intended to participate in a non-collision sport (e.g. basketball or swimming) and 1,408 were non-athletes.

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  - ★ *Personality (e.g. Angry Hostility Scale)*

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- **Anticipate unmeasured confounding:** Consider multiple comparisons that address plausible sources of unmeasured confounding while not sacrificing power.
- **Avoid data-snooping:** Like a RCT, publicly register a pre-analysis plan prior to joint analysis of outcomes and exposure (Rubin, 2007).

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  - ★ *Matching closely on covariates guards against misspecification of propensity score and **reduces variance**.*
- **Rich set of baseline covariates:** sociodemographic status, school performance/engagement, life plan, social engagement, psychological characteristics, physical functioning.

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- **Solution:** Developed an ordered testing procedure that seeks reassurance that the natural comparison is not an oversimplification of these *versions* while not sacrificing power in the main comparison (Hasegawa et al., 2018).

# Testing in order: all controls, versions of control, and equivalence of controls

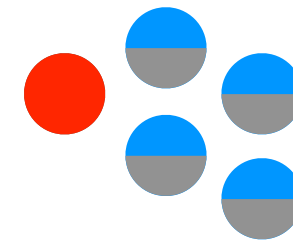
● Football Players    ● Non-collision Athletes    ● Non-athletes    ● All controls



# Testing in order: all controls, versions of control, and equivalence of controls

- **Match 1:** treatment vs. all controls — maximize power.

**Match 1:**  
 $n_{\text{treated}}=578$ ,  $n_{\text{all\_control}}=1385$



Football Players



Non-collision Athletes



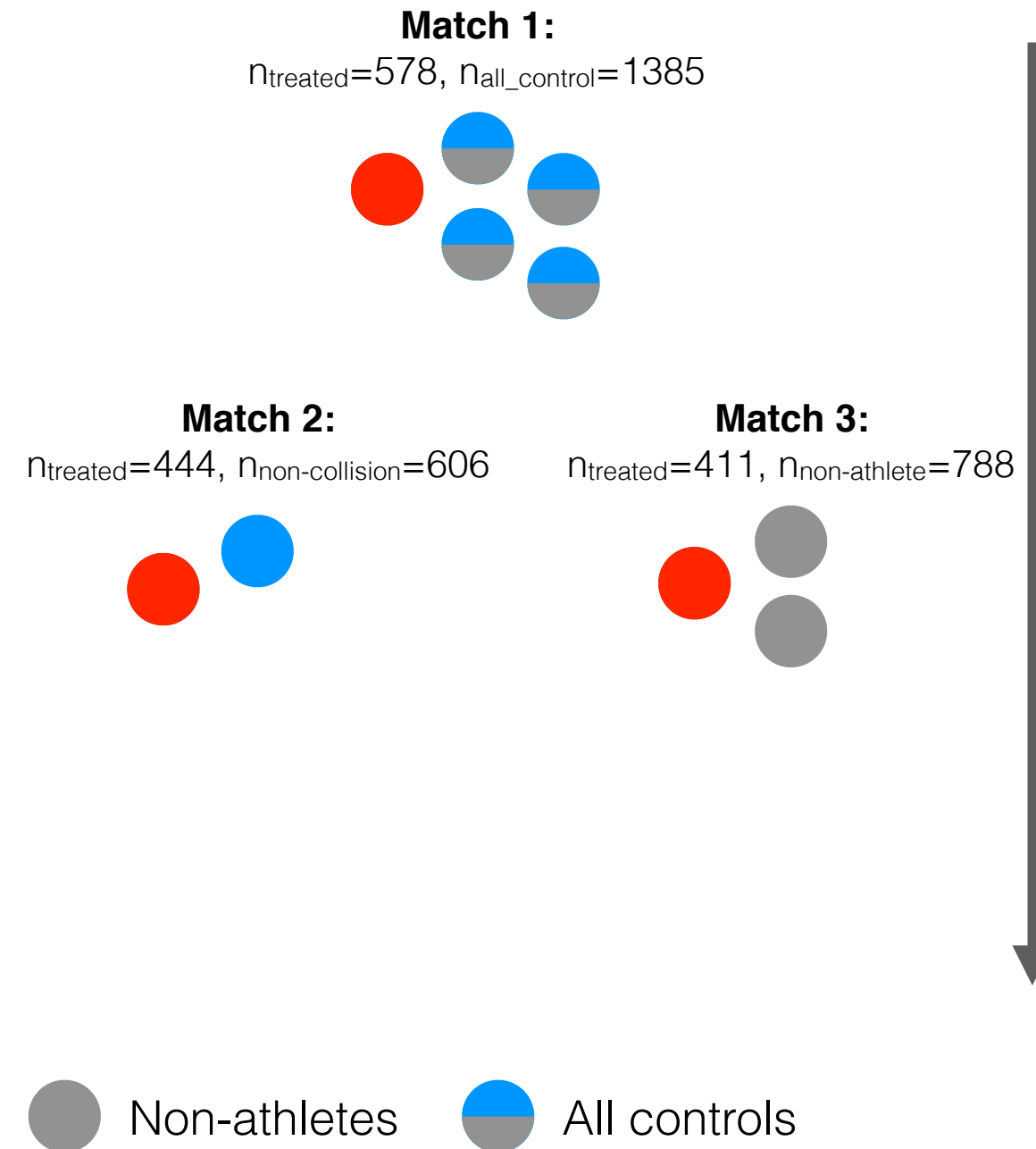
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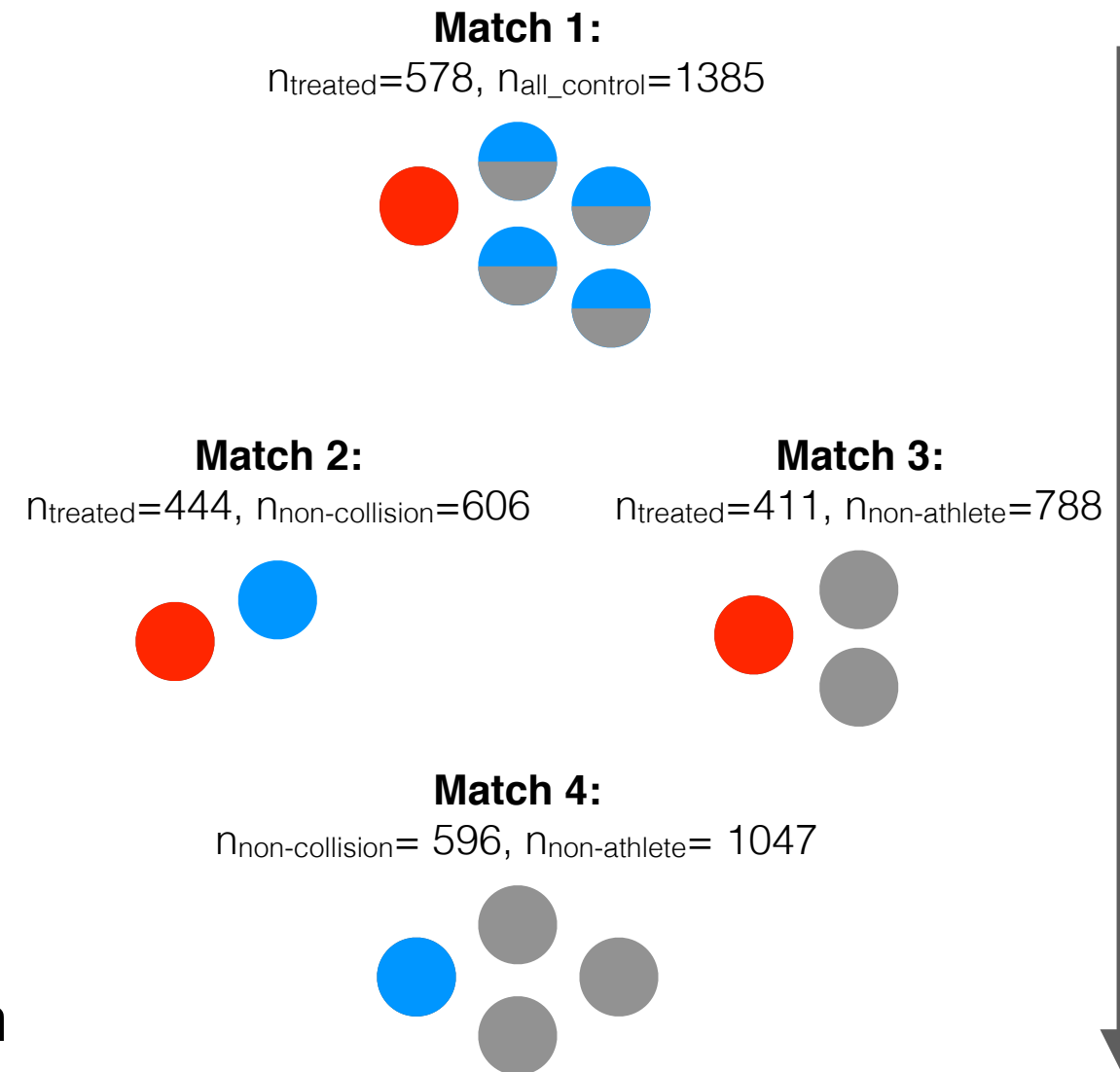
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- **Match 2 & 3:** treatment vs. non-collision athletes & treatment vs. non-athletes — are athletes inherently different than non-athletes? do positive effects of playing a sport counteract negative effects from head-injury exposure?



# Testing in order: all controls, versions of control, and equivalence of controls

- **Match 1:** **treatment** vs. **all controls** — maximize power.
- **Match 2 & 3:** **treatment** vs. **non-collision athletes** & **treatment** vs. **non-athletes** — are athletes inherently different than non-athletes? do positive effects of playing a sport counteract negative effects from head-injury exposure?
- **Match 4:** equivalence test of **non-collision athletes** vs. **non-athletes** — seek further reassurance that the comparison in Match 1 is appropriate.

● Football Players    ● Non-collision Athletes    ● Non-athletes    ● All controls





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- Open to advice, comments, and suggestions from experts on Add Health to help us refine our analysis plan.

# Thank you!

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# BACKUP SLIDE [1/2]: An additional feature of our observational study design

- **Minimize variance without making additional modeling assumptions:** Agnostic covariance adjustment using flexible machine learning algorithms (Rosenbaum, 2002).

# BACKUP SLIDE [2/2]: Flexible covariance adjustment: Bayesian Additive Regression Trees (BART)

- Under hypothesis of an additive treatment effect  $\tau$  can infer potential outcomes had all subjects received control.
- Can use machine learning algorithms to “model” potential outcomes under control to reduce variance coming from residual imbalances in prognostic covariates after matching.
- Correctness of “model” not required for valid inference, that comes from randomization or like-randomized observational study.
- BART is a flexible, Bayesian regression algorithm with nice regularization properties.

