

# Sensitivity analysis

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- **overt bias** could occur when outcome analysis is carried out despite imbalance on the observed covariates  $X$
- there is no guarantee that matching results in balance on variables that we did not match on (e.g. unobserved variables)
  - ↳ if these unobserved variables are **confounders**, then we have **hidden bias**  
⇒ ignorability assumption would then be violated (**unmeasured confounding**)

## Main idea of sensitivity analysis:

- if there is hidden bias, how severe would it have to be before conclusions changed
  - ↳ change from stat. significant to stat. insignificant result
  - ↳ change in direction of treatment effect (sign  $\pm$  changed)

## Sensitivity analysis cont'd:

- $\pi_j$  ... prob. of subject  $j$  receiving treatment  
 $\pi_k$  ... " — subject  $k$  receiving treatment
- suppose subjects  $j$  and  $k$  are perfectly matched, so that observed covariates  $X_j$  and  $X_k$ , are the same
- now if  $\pi_j = \pi_k$ , then we have to conclude, that there is no hidden bias

- assume following inequality

$$\frac{1}{\Gamma} \leq \frac{\frac{\pi_j}{(1-\pi_j)}}{\frac{\pi_k}{(1-\pi_k)}} \leq \Gamma \quad (\text{capital gamma})$$

odds of treatment person  $j$   $\left( \frac{p_{\text{treat}}}{1-p_{\text{treat}}} \right)$

odds of treatment person  $k$

odds-ratio

- $\Gamma$  ... is an odds ratio
  - ↳ if.  $\Gamma = 1$ , then no overt bias
  - ↳ if.  $\Gamma > 1$ , then hidden bias
- if  $\Gamma$  is very close to 1, then the **no hidden bias assumption** (i.e. ignorability assumption) is barely to violated
- when we estimate / test treatment effect, we assume no hidden bias
  - ↳ i.e.  $\Gamma = 1$
  - ↳ thus we can carry out sensitivity analysis by gradually increasing  $\Gamma$  until the evidence of treatment effect goes away:
    - if evidence goes away with little amount of increasing  $\Gamma$  (e.g.  $1 \rightarrow 1.1$ ), then we conclude, that our treatment effect inference is very sensitive to unmeasured confounding
    - however, if evidence remains even after large increase of  $\Gamma$  (e.g.  $1 \rightarrow 5$ ), then we conclude, that our analysis of treatment effect inference is robust to hidden bias from unmeasured confounding
- R packages for sensitivity analysis: sensitivityfull, sensitivity2x2xk.