# Understanding Effect Sizes in Research - english

## I. Introduction

**Definition:** - **Effect Size:** A statistical measure that quantifies the size or magnitude of an observed effect in a study.

## II. Importance of Effect Sizes

1. **Complement to Significance Testing:**
   - *Significance vs. Practical Relevance:* While significance testing indicates if results are unlikely due to chance, effect sizes provide information on the practical importance of the findings.
2. **Interpreting Results:**
   - *Go Beyond p-values:* Effect sizes help researchers interpret the magnitude of differences or relationships, moving beyond the binary "significant" or "non-significant" distinction.

## III. Types of Effect Sizes

1. **Cohen's d:**
   - *Definition:* Measures the standardized difference between two means.
   - *Interpretation:* Small ($d = 0.2$), medium ($d = 0.5$), and large ($d = 0.8$) effect sizes.
2. **Pearson's r:**
   - *Definition:* Represents the strength and direction of a linear relationship between two variables.
   - *Interpretation:* Ranges from -1 to 1, with 0 indicating no correlation, and -1 or 1 indicating a perfect negative or positive correlation, respectively.
3. **Odds Ratio (OR):**
   - *Definition:* Commonly used in logistic regression, quantifying the odds of an event occurring in one group compared to another.
   - *Interpretation:* An OR of 1 suggests no effect; values greater or less than 1 indicate increased or decreased odds.
4. **Beta ($\beta$ - Standardized Regression Coefficient):**
   - *Definition:* Represents the change in the standard deviation of the dependent variable for a one-standard-deviation change in the independent variable.
   - *Interpretation:* Allows for the comparison of the strength of different predictors.
5. **B (Regression Coefficient):**
   - *Definition:* Represents the change in the dependent variable for a one-unit change in the independent variable in regression analysis.
6. **R-squared:**

- *Definition:* In regression, it indicates the proportion of variance in the dependent variable explained by the independent variable(s).
- *Interpretation:* Values range from 0 to 1, with higher values indicating a greater proportion of variance explained.

7. **Adjusted R-squared:**
   - *Definition:* Similar to R-squared but adjusts for the number of predictors in the model.
   - *Interpretation:* Helps mitigate the risk of overfitting in models with multiple predictors.

# IV. Practical Considerations

1. **Context Matters:**
   - *Discipline-Specific Norms:* Effect size benchmarks may vary across different psychological subfields.
2. **Confidence Intervals:**
   - *Incorporate Uncertainty:* Effect sizes are more informative when accompanied by confidence intervals, providing a range of plausible values.

# V. Reporting Effect Sizes

1. **In Research Papers:**
   - *Include in Results Section:* Report effect sizes alongside p-values for a comprehensive understanding of study outcomes.
2. **Meta-Analysis:**
   - *Facilitates Comparisons:* Effect sizes enhance the comparability of studies in meta-analyses, contributing to a more comprehensive synthesis of evidence.

# VI. Eta Squared ($\eta^2$ and Omega Squared ($\omega^2$)

1. **Eta Squared ($\eta^2$):**
   - *Definition:* In ANOVA, it represents the proportion of variance in the dependent variable explained by the independent variable(s).
   - *Interpretation:* Values range from 0 to 1, with higher values indicating a greater proportion of variance explained.
2. **Omega Squared ($\omega^2$):**
   - *Definition:* Similar to eta squared but provides a less biased estimate, especially in smaller sample sizes.
   - *Interpretation:* Like eta squared, values range from 0 to 1.

# VII. Conclusion

**Key Takeaways:** - Effect sizes provide valuable information about the practical significance of study results. - Use multiple effect size measures for a comprehensive understanding of the observed effects. - Consider discipline-specific norms and incorporate confidence intervals for a nuanced interpretation.

# VIII. References


- Cohen, J. (1988). Statistical Power Analysis for the Behavioral Sciences. Hillsdale, NJ: Lawrence Erlbaum Associates.