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Modeling Co-occurring Health Conditions to Predict Health Utility from Generic Health-Related Quality of Life Indexes

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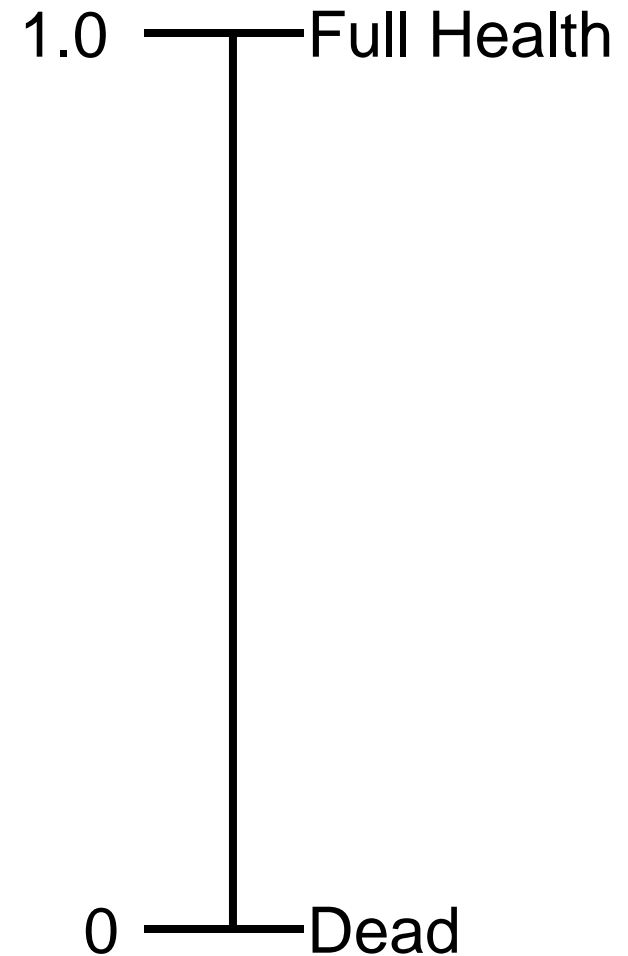
This work was funded by a NIH Integrated Training for Physician Scientists Training Grant (GM008692), a dissertation grant from the Agency for Healthcare Quality and Research (R36 HS016574), and P01 AG020679 from the National Institute on Aging (“Norms and Performance Comparisons for 5 Health Indexes”)



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Health Utility

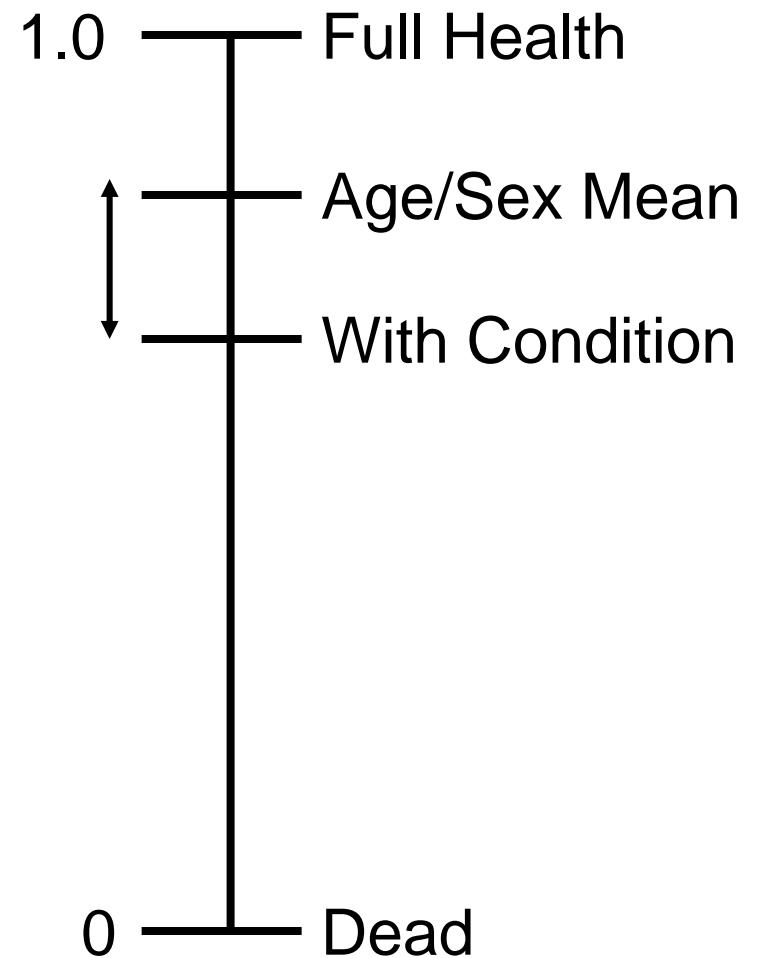
- ❑ Preference-based summary score of health
- ❑ Scores from generic health-related quality of life (HRQoL) measures are appropriate for cost-effectiveness analyses





Health Utility Impact of a Single Condition

- ❑ Cost-effectiveness analyses often require an estimate of the health utility change of a single health condition
- ❑ When longitudinal change data are not available, estimate differences in cross-sectional scores
 - Current recommendations suggest comparing scores from a sample with a health condition to age- and sex-adjusted normative values





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Estimating the Impact of a Single Condition When Other Conditions are Present

- ❑ We should be able to estimate the impact of a single health condition on health utility independent of the other health conditions in large datasets
- ❑ It is not known how the impacts of multiple health conditions combine:
 - Additive
 - Multiplicative
 - Minimum
- ❑ A rule for modeling the impact of multiple health conditions is particularly useful for estimating condition impact catalogs



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Data

- ❑ 1998-2004 Medicare Health Outcomes Survey
- ❑ SF-6D health utility score from the SF-36 version 1
- ❑ Exclude proxy respondents and surveys with less than 80% completed

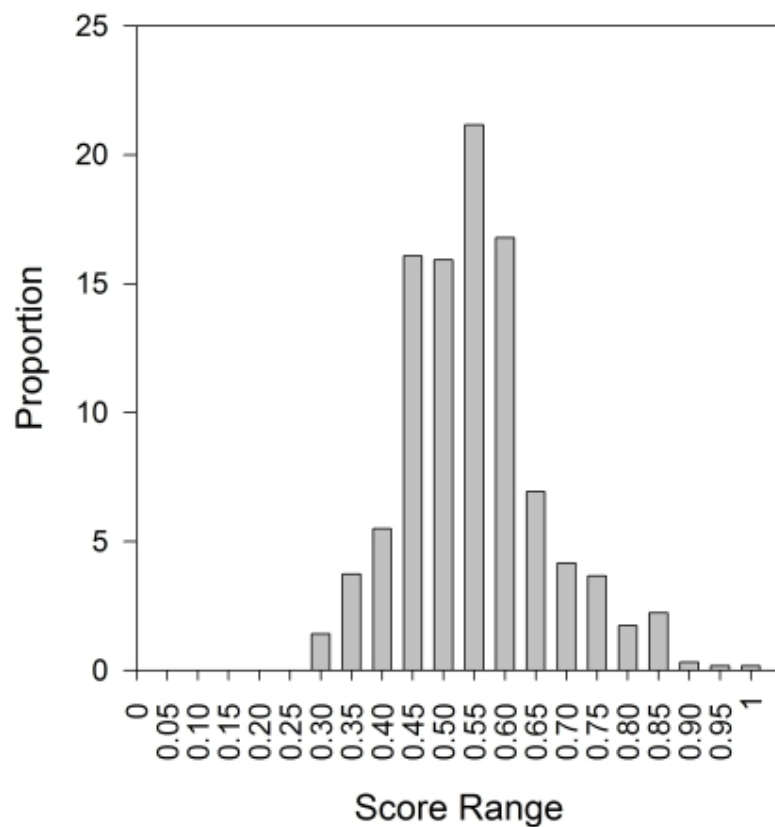
- ❑ 103,484 observations used for model estimation
- ❑ 102,945 observations used for model testing
- ❑ Split data into two groups:
 - aged 65 and older
 - under age 65



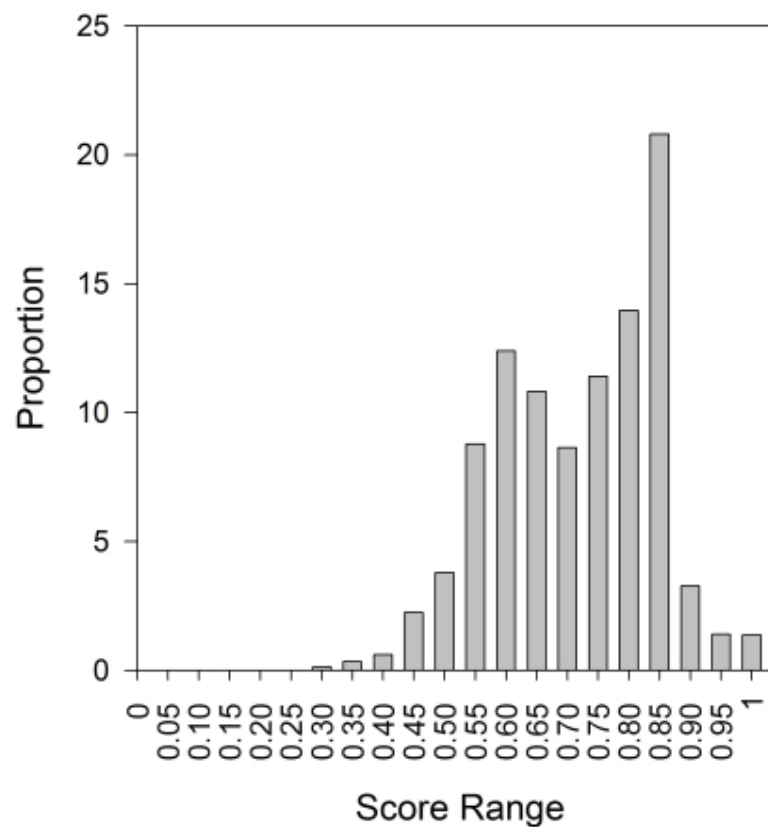
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Distribution of SF-6D Scores

Age 64 and Younger



Age 65 and Older





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Models

- ❑ All models fit in WinBUGS 1.4.1
- ❑ Censored at 1.0 and 0.3

- ❑ SF-6D summary score modeled with:
 - 15 health conditions

 - Age in years
 - Sex
 - Education
 - Income



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4 Combination Rules

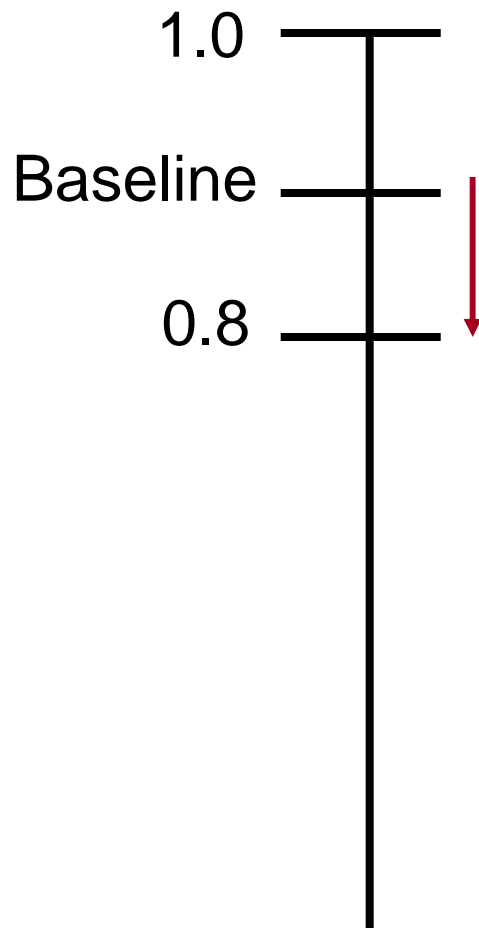
- Additive
- Minimum
- Multiplicative
- Count



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Additive Model

Condition A: -0.1



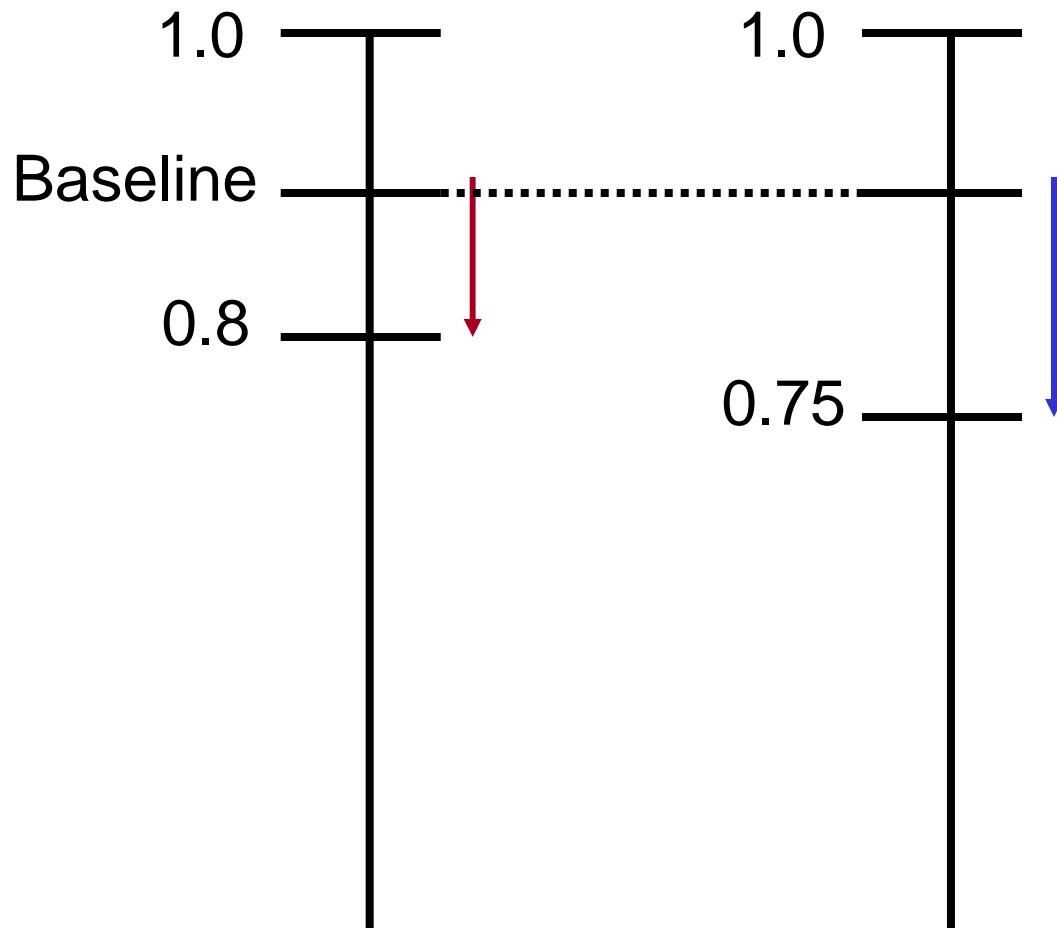


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Additive Model

Condition A: -0.1

Condition B: -0.15



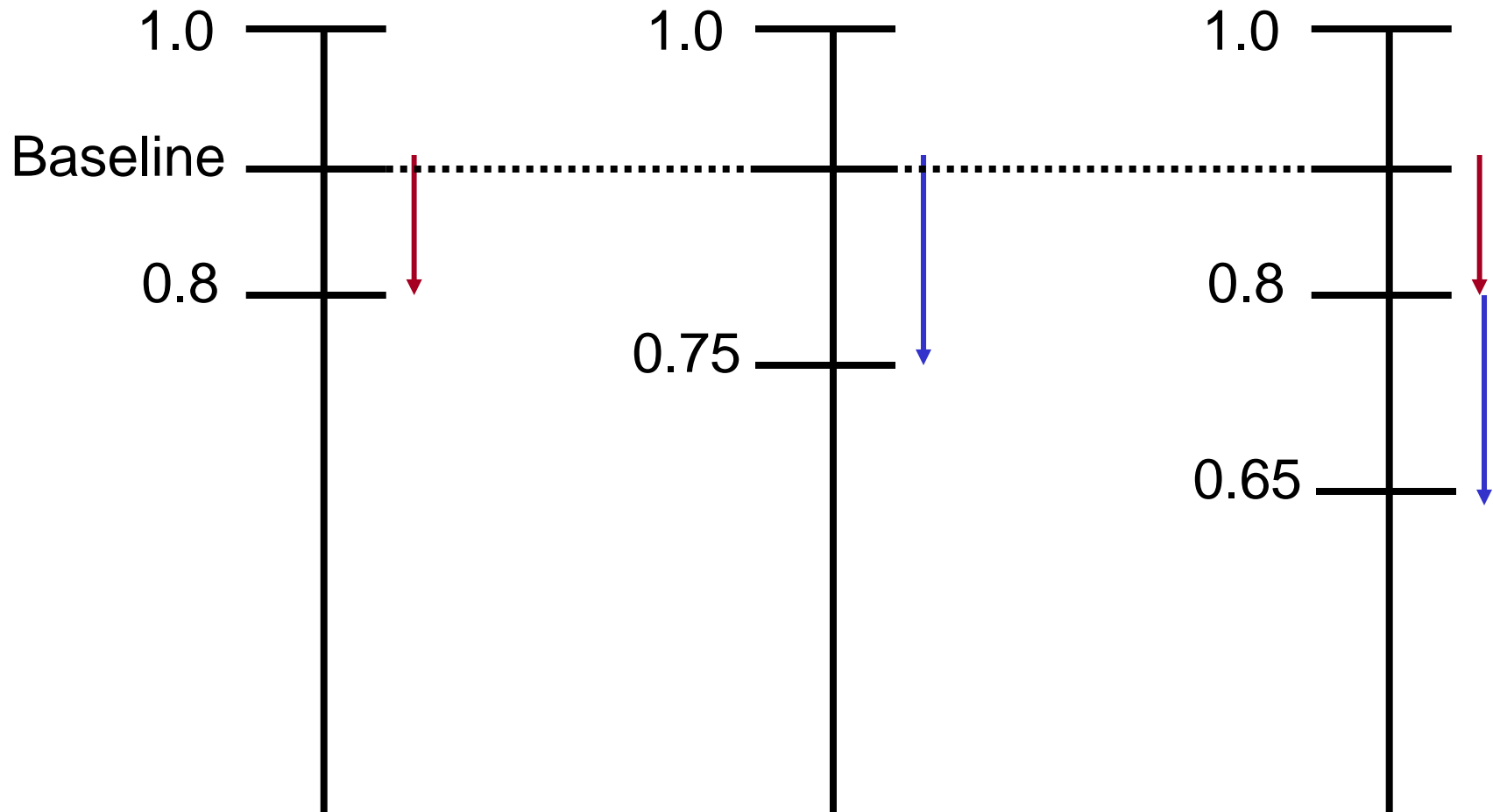


Additive Model

Condition A: -0.1

Condition B: -0.15

A & B: -0.25

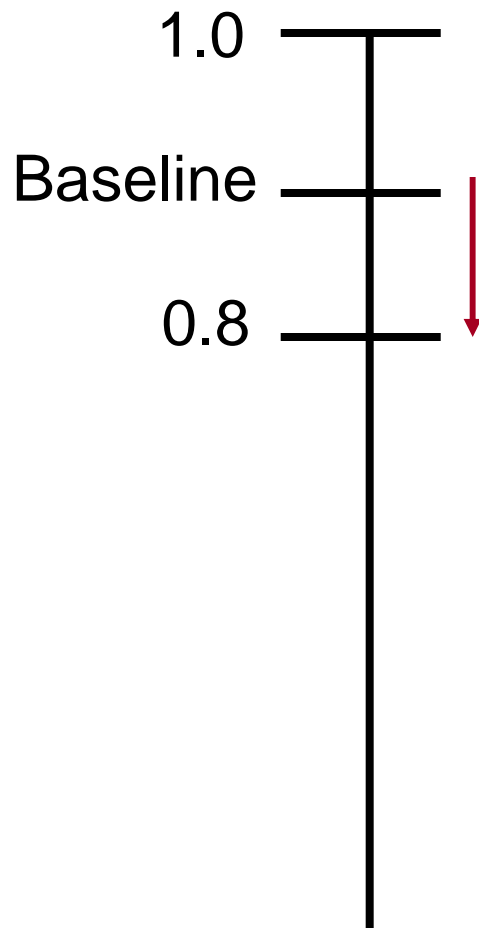




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Minimum Model

Condition A: -0.1



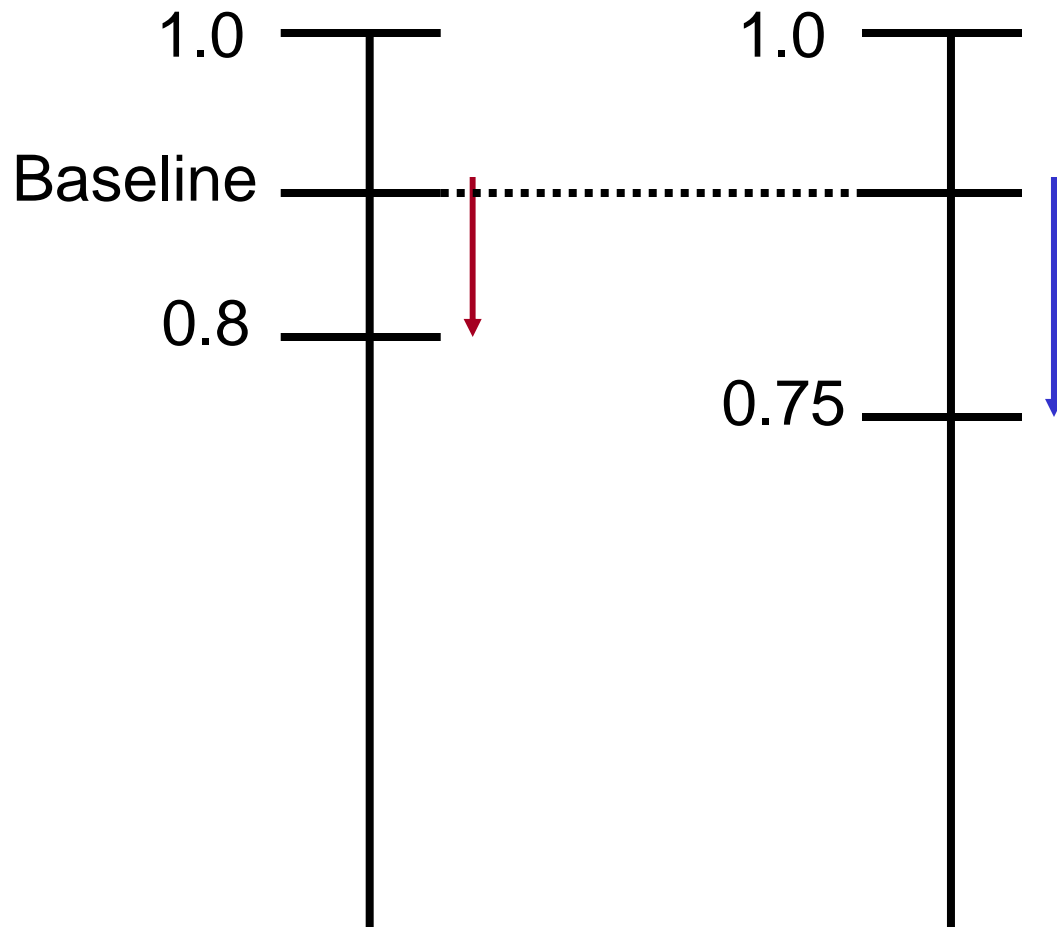


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Minimum Model

Condition A: -0.1

Condition B: -0.15



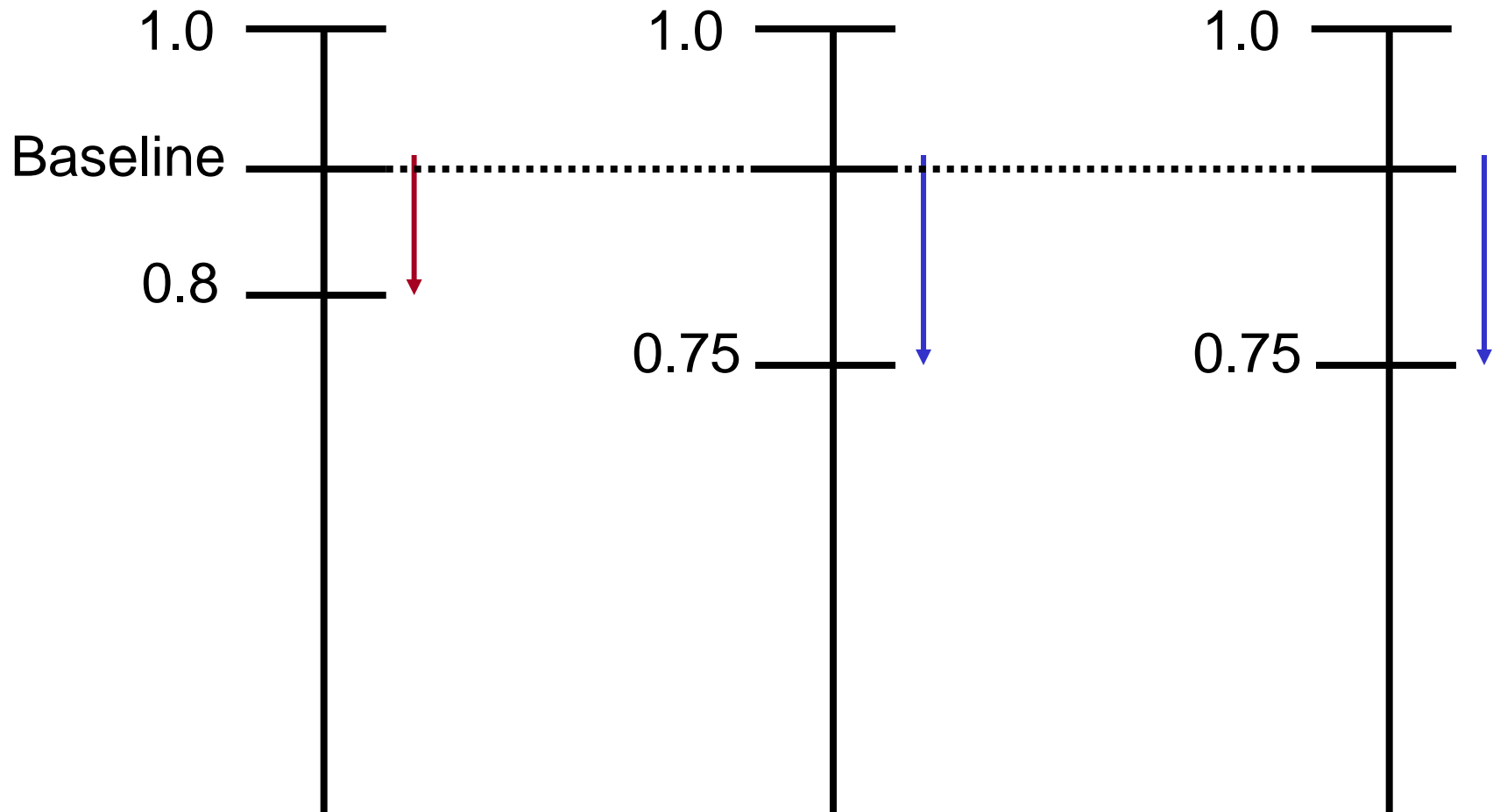


Minimum Model

Condition A: -0.1

Condition B: -0.15

A & B: -0.15





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Multiplicative Model

Condition A: -0.1



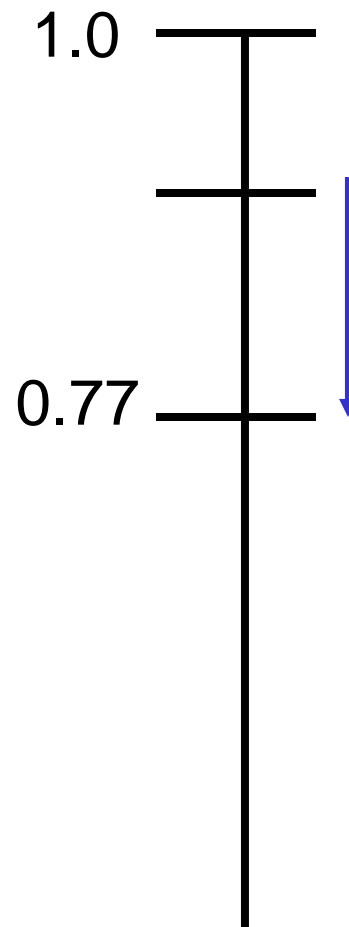
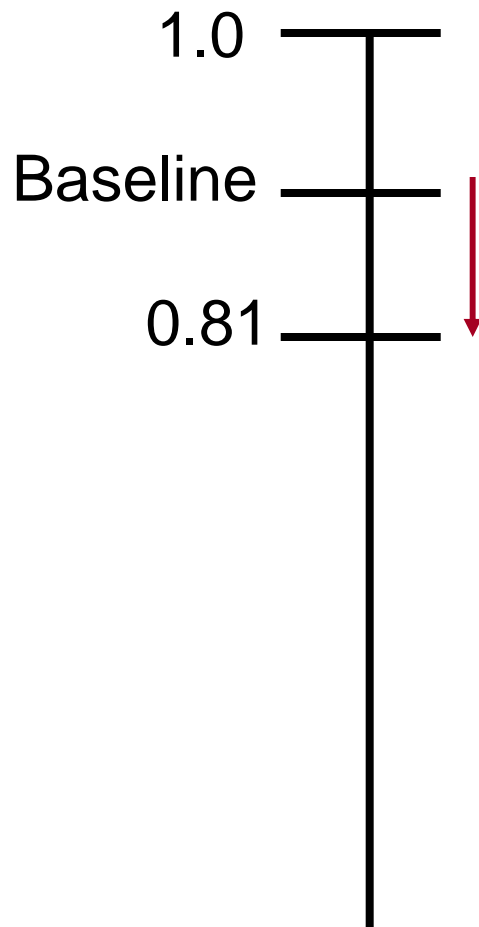


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Multiplicative Model

Condition A: -0.1

Condition B: -0.15



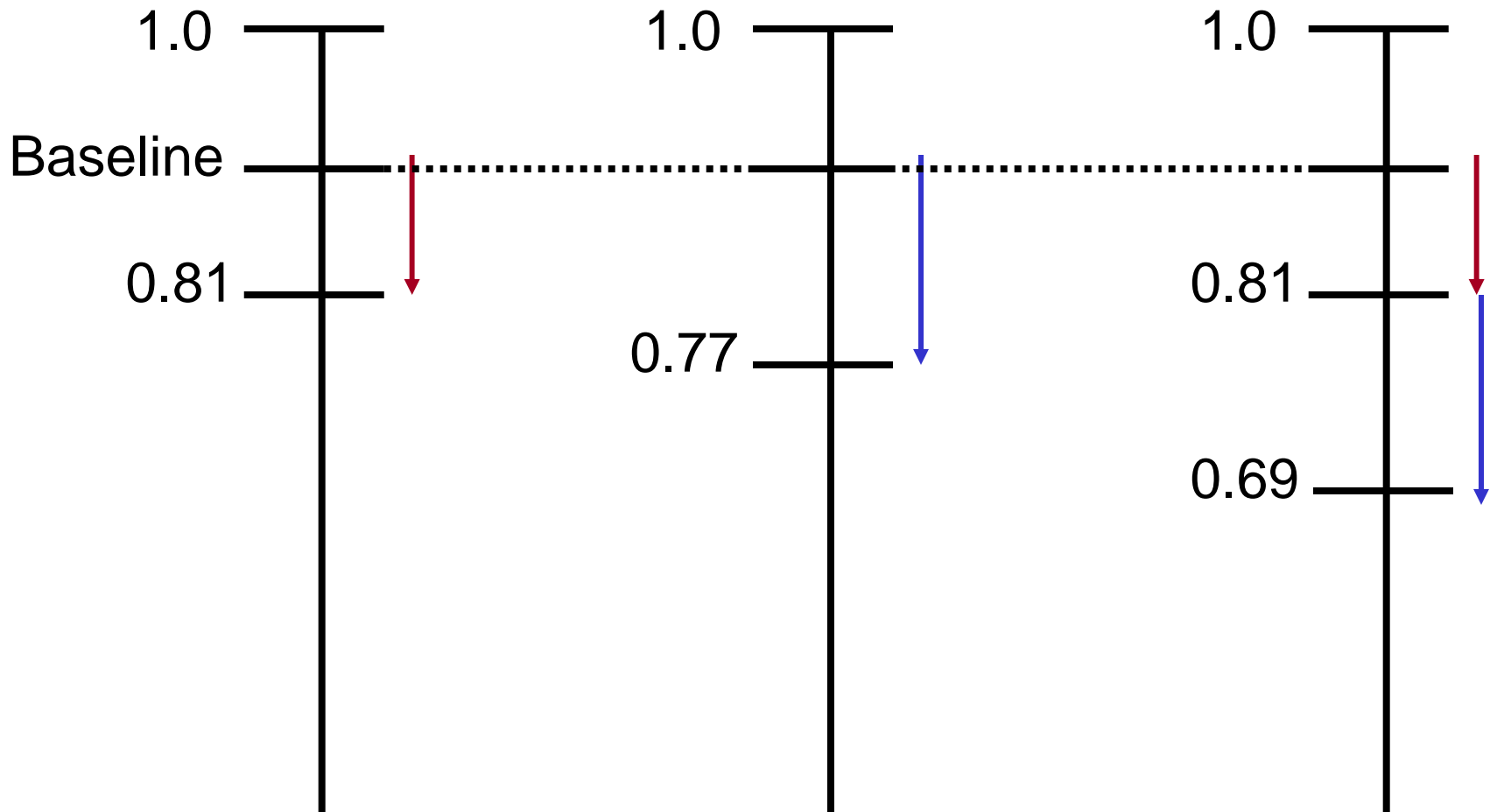


Multiplicative Model

Condition A: -0.1

Condition B: -0.15

A & B: -0.21



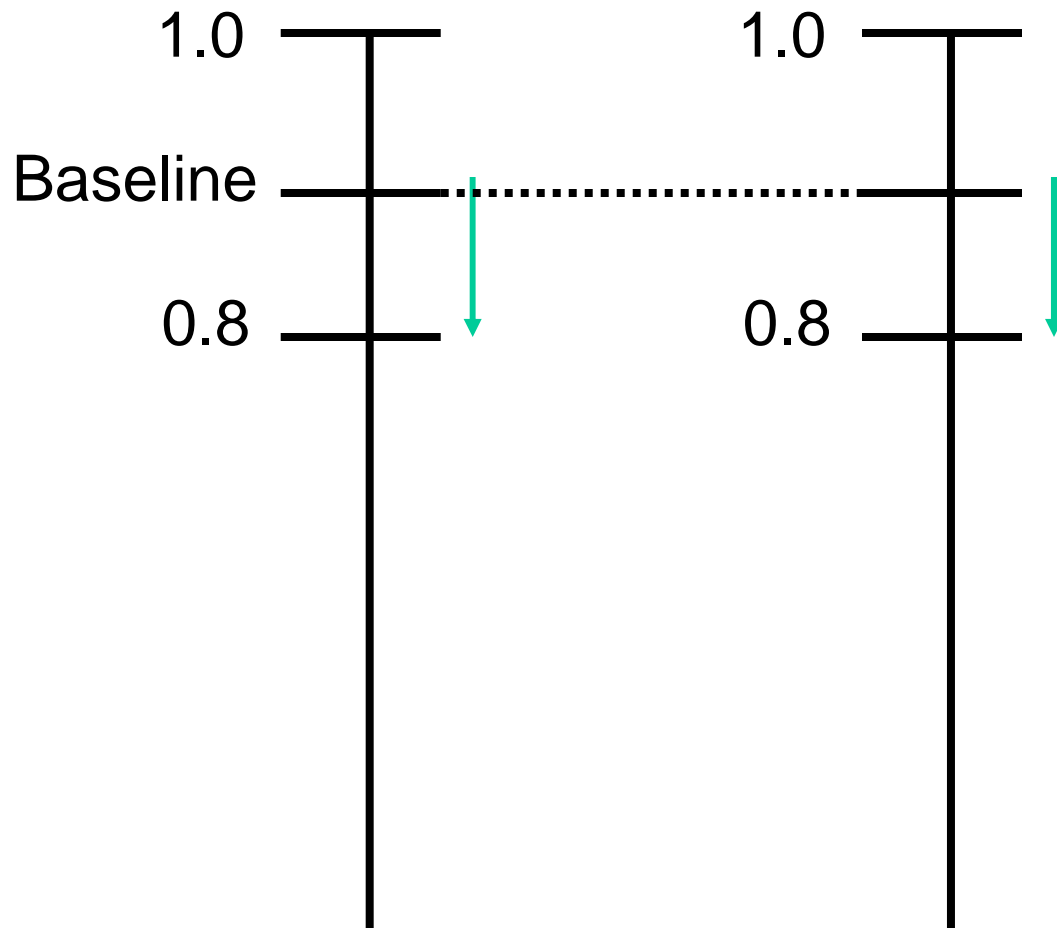


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Count Model

Condition A: -0.1

Condition B: -0.1



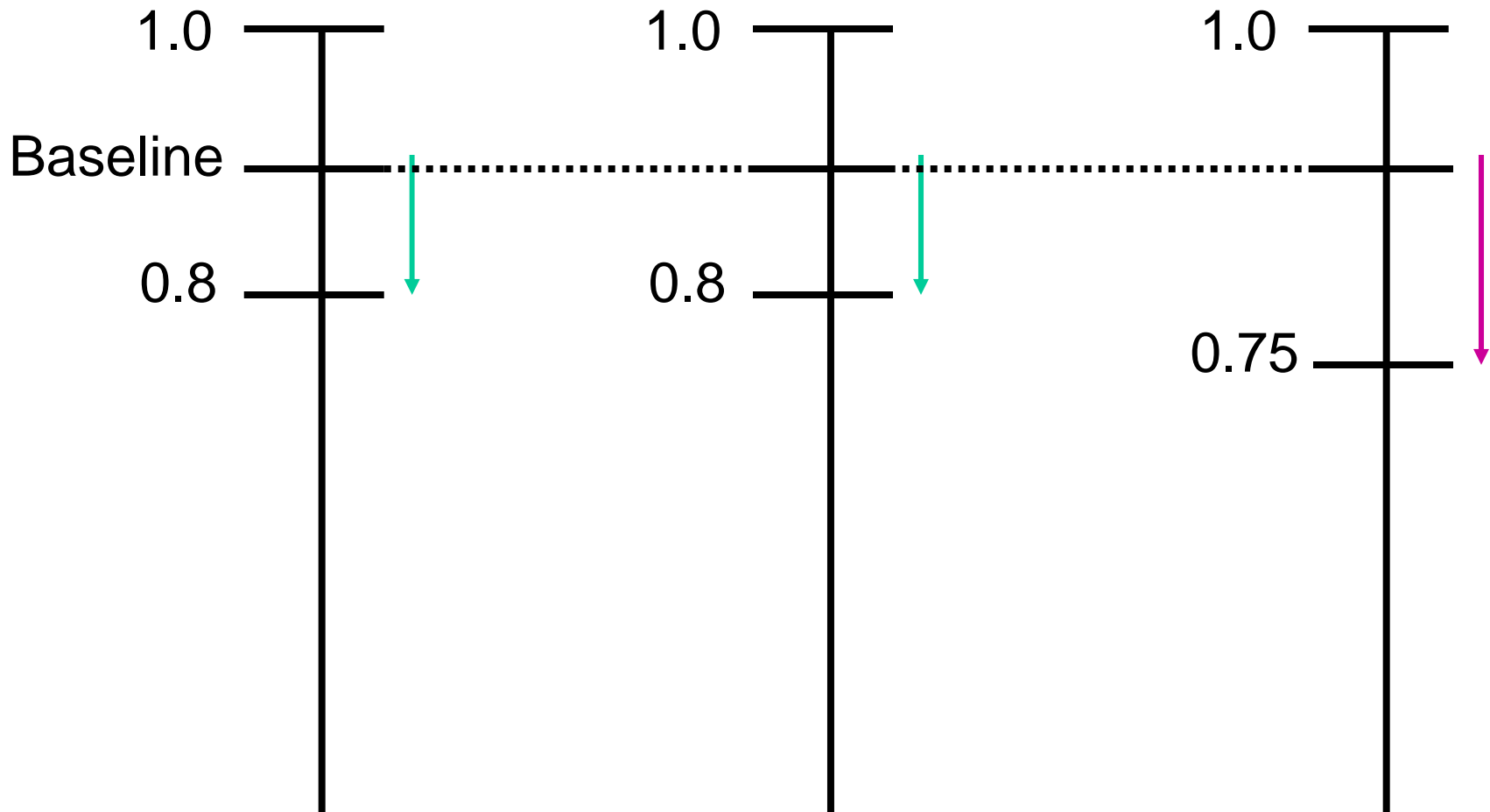


Count Model

Condition A: -0.1

Condition B: -0.1

A & B: -0.15





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Deviance Information Criteria

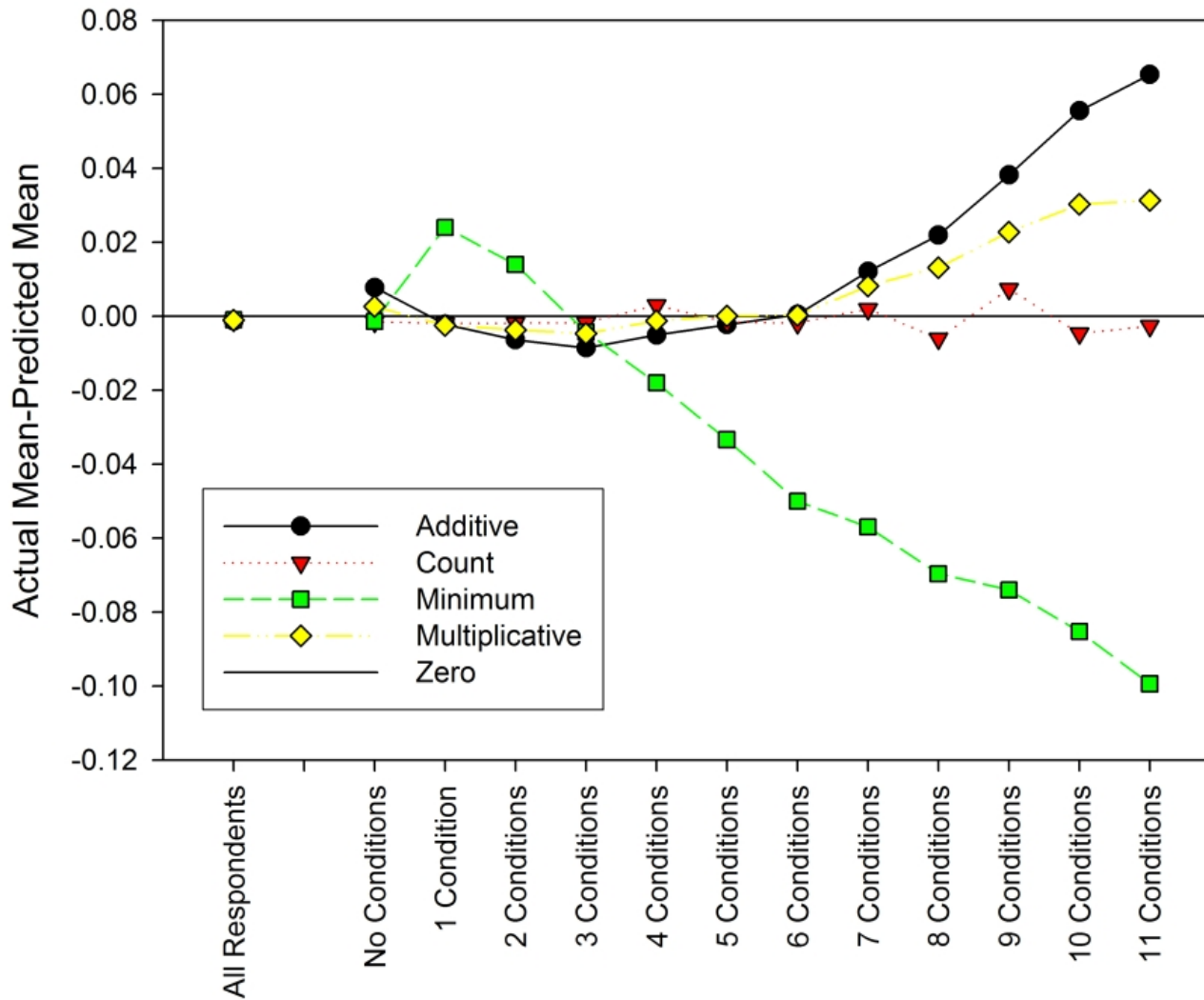
- ❑ Measure of overall model fit

- ❑ The multiplicative model has substantially improved model fit compared to all other models
 - In both age groups
 - In models both adjusted and unadjusted for age, sex, education, and income

- ❑ Adjusted models have better fit than unadjusted models

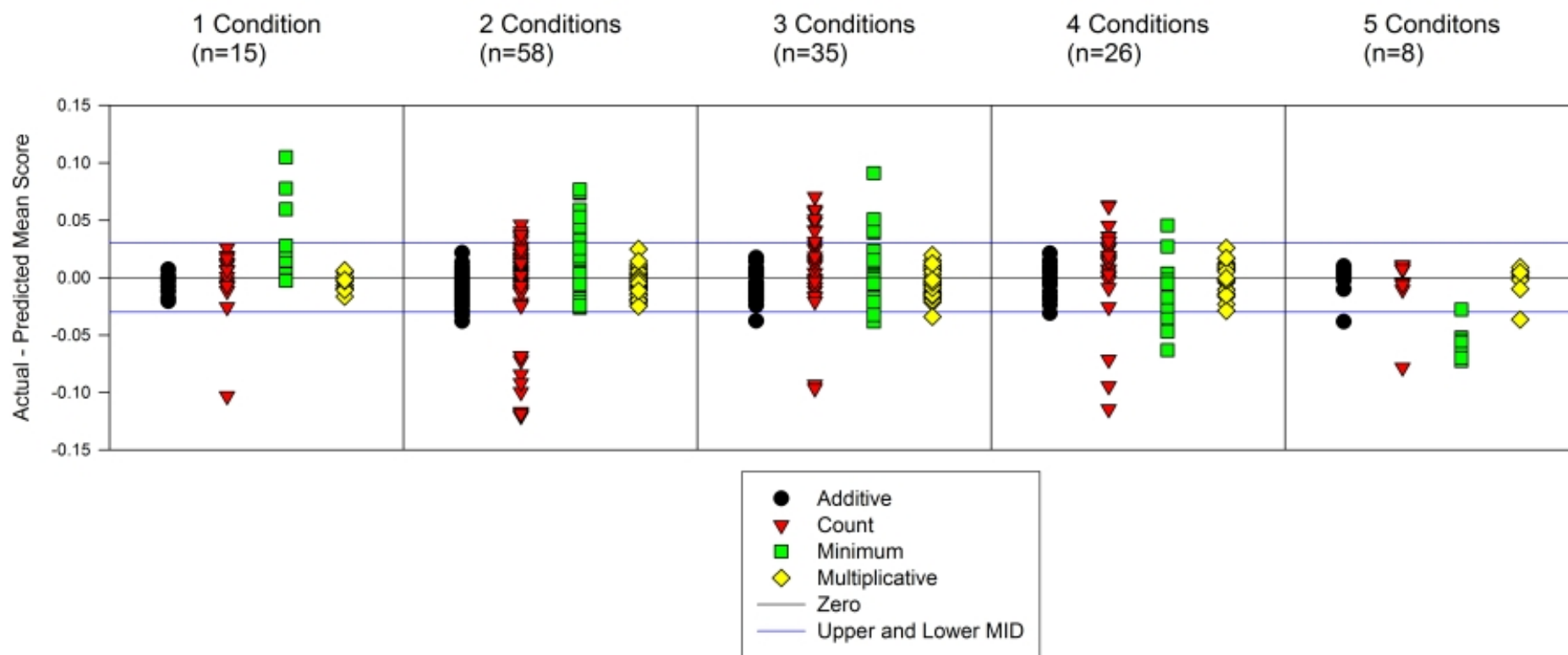


Predictive Error by Number of Conditions





Predictive Error for Named Conditions



Each point represents a group of individuals (n>50) with a particular combination of health conditions (i.e., stroke and a respiratory condition)



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Results from the Multiplicative Model

Health Condition	Rank	Impact on SF-6D Score (SD)
Depression	1	-0.167 (0.002)
Respiratory Condition	2	-0.058 (0.001)
Arthritis of the Hip/Knee	3	-0.057 (0.001)
...		
Diabetes	13	-0.027 (0.001)
Coronary Artery Disease	14	-0.025 (0.001)
Myocardial Infarction	15	-0.009 (0.002)



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Generalizability

- ❑ Similar analyses using two other datasets
- ❑ Medical Expenditure Panel Survey (MEPS)
 - 2000-2002
 - EQ-5D (with US Scoring), n= 50,565
 - SF-6D from the SF-12 version 1, n= 49,474
 - 15 of 128 chronic health conditions
- ❑ Canadian Community Health Survey 1.1 (CCHS)
 - 2000-2001
 - Health Utilities Index Mark 3 (HUI3), n= 116,908
 - 15 of 25 chronic health conditions



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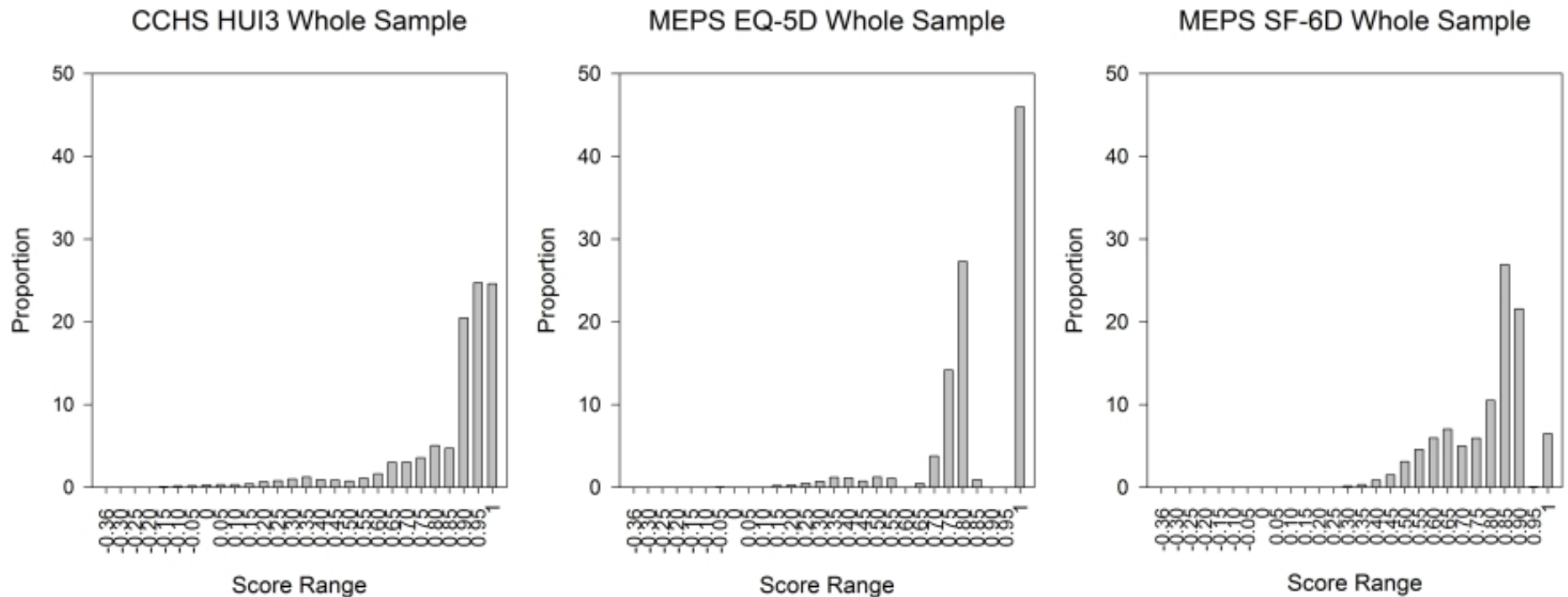
Results

Dataset	HRQoL Measure	Fit Criterion	
		DIC	Predictive Validity
Medicare	SF-6D	Multiplicative	Multiplicative
MEPS	SF-6D	Ambiguous	Multiplicative?
	EQ-5D	Ambiguous	Multiplicative?
CCHS	HUI3	Additive	Additive



General Population Surveys

- Analyses are complicated by high health utility scores, low condition prevalence, and weights





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Strengths and Limitations

□ Strengths

- Large datasets with diverse populations
- Prevalent health conditions which impact HRQoL

□ Limitations

- Did not test entire space of possible combination rules
- Have not fully tested models in CCHS and MEPS
- Generalizability may be limited by HRQoL measure or dataset



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Conclusions

- ❑ When modeling the SF-6D health utility score in the Medicare Health Outcomes Survey, a multiplicative model had substantially improved fit compared to the other tested models
- ❑ All combination rules are imperfect
- ❑ Future work may find generalizable rules for a HRQoL score
 - Overall measures of model fit may not be appropriate for datasets with very healthy respondents



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Thank you!

We look forward to your comments and questions

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