Identifying Thresholds of Quality in Early Child Care and Education: A Non-Parametric Approach



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Overview

- Background
- Data Sources
- Analytic Approaches
 - General Additive Modeling
 - Spline Analyses
 - Framework for this project
- Examples
- Summary



Background

- Led by CYFS Faculty Affiliates
 - Julia Torquati
 - Helen Raikes
- Substantively speaking...
 - Evidence that better quality generally predicts more optimal child outcomes



Aims

- <u>Aim 1</u>: To determine whether there are *specific thresholds* of quality that predict children's developmental outcomes
- <u>Aim 2</u>: To determine whether thresholds of quality vary as a function of:
 - (1) age of child in child care;
 - (2) ethnicity of teacher/provider;
 - (3) ethnic match/mismatch of teacher/provider and child;
 - (4) income level of child



Background

- Linear vs. Non-Linear models
 - Linear the norm
 - Implies relationship same across all points
 - \circ Non-linear are necessary
 - The strength of association between predictors and outcomes varies across the measurement scale such that particular levels of quality are significantly more strongly associated with children's development.



Early Head Start Sample

	Family Child Care	Center Care
14 months	151	372
24 months	159	416
36 months	134	500
60 months	80	1043



QUINCE Sample

- Children ages 20 months 5 years
- Assessed fall and spring
 - T2 focus of this analysis

Family Child Care	Center Care
650	652



Analytic Approaches

- General Additive Modeling
 - Empirically driven
 - Non-parametric approach
 - Identify relationship between variables
 - Linear vs. Non-linear
 - Identify possible thresholds
 - Based on shape of observed relationship



Analytic Approaches

- Spline Models
 - Identify empirical thresholds obtained from GAM via the model comparison between linear and non-linear associations
 - Investigate the change of associations within each interval of quality of care classified by empirical thresholds
 - Predict associations based on the spline model



Introduction to GAM

- Generalized Additive Model (GAM)
 - Each term is estimated using a univariate smoother
 - The estimate explains how the dependent variable changes with the corresponding independent variables

$$\eta = s_0 + \sum_{i=1}^p s_i(X_i)$$

where S_i are smooth functions defines the additive component.

- The smoothers are estimated by using the backfitting algorithm and the local scoring algorithm
 - The algorithms are implemented in PROC GAM in SAS and gamm package in R



Spline Analyses

• For identification of thresholds, the following two models are compared

Linear model $Y_i = \beta_0 + \beta_1 \cdot X_i + e_i$ Spline model $Y_i = \beta_0 + \beta_1 \cdot X_i + \beta_2 \cdot T_i^1 + \beta_3 \cdot T_i^2 + e_i$

• To investigate the effect of moderator, the following two models are compared

Spline model $Y_i = \beta_0 + \beta_1 \cdot X_i + \beta_2 \cdot T_i^1 + \beta_3 \cdot T_i^2 + e_i$ Model with Moderator $Y_i = \beta_0 + \beta_1 \cdot X_i + \beta_2 \cdot T_i^1 + \beta_3 \cdot T_i^2 + \sum_{k=1}^{K} C_i^k + e_i$



Framework for Identifying Thresholds

- Thresholds were identified
 - Empirically via GAM
 - Based on a priori research/knowledge



Example 1

- Begin by implement GAM approach
 - Non-parametric approach to investigating the relationship between quality measures of child care (ITERS or ECERS, FDCRS, and CIS) and child outcomes
- Want to determine:
 - Relationship linear or non-linear?
 - If non-linear, are thresholds implied?



Example 1

- To examine the relationship between ITERS quality measure and sustained attention outcome measure at 14 month in the EHS sample
- Using PROC GAM in SAS

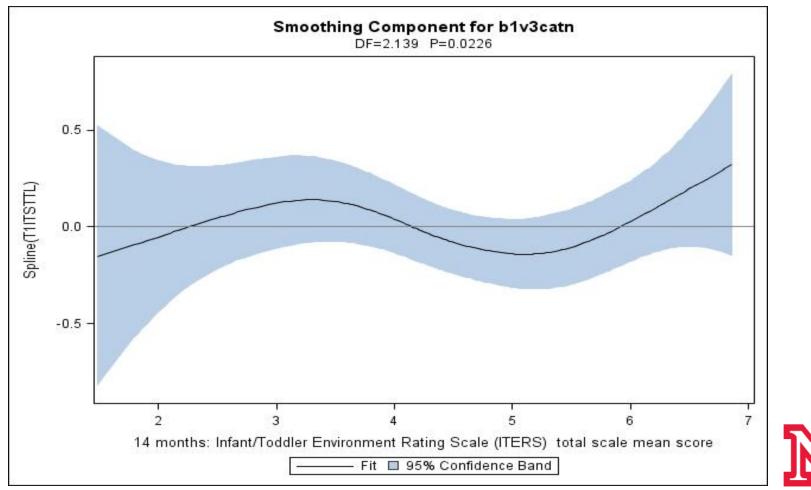
run;

• It produces parameter estimates, estimated degree of freedom, and smoothing plot(s)



Example 1 – GAM Plot

• ITERS and Sustained Attention



Example 1 – SAS Output

The GAM Procedure Dependent Variable: b1v3catn Smoothing Model Component(s): spline(T1ITSTTL)

Summary of Input Data Set	
Number of Observations	335
Number of Missing Observations	2666
Distribution	Gaussian
Link Function	Identity

Regression Model Analysis

Parameter Estimates

Parameter	Parameter Estimate	Standard Error	t Value	Pr > t
Intercept	5.15073	0.21544	23.91	<.0001
Linear (T1ITSTTL)	-0.04984	0.04601	-1.08	0.2795

Smoothing Model Analysis

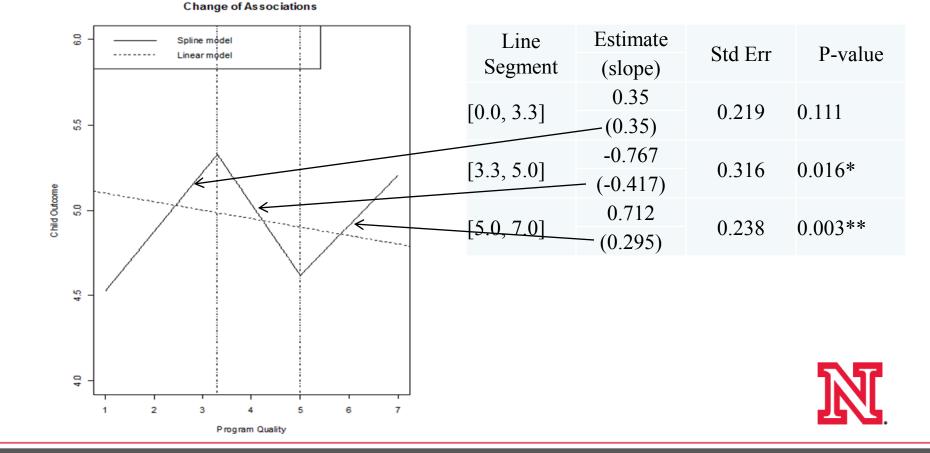
Analysis of Deviance

Source	DF	Squares	Chi-Square	Pr > ChiSq
Spline(T1ITSTTL)	2.13902	8.274623	7.8755	0.0226
		Sum of		



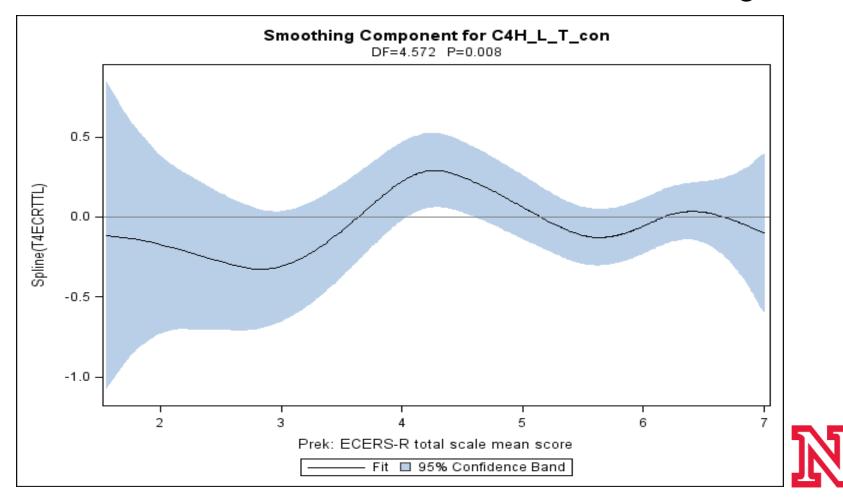
Example 1 – Spline Result

- Testing empirical thresholds 3.3 and 5.0
 - F2=4.628; p-value=0.010



Example 2 – GAM Plot

• ECERS-R and Howes Ladd conflict with teacher rating

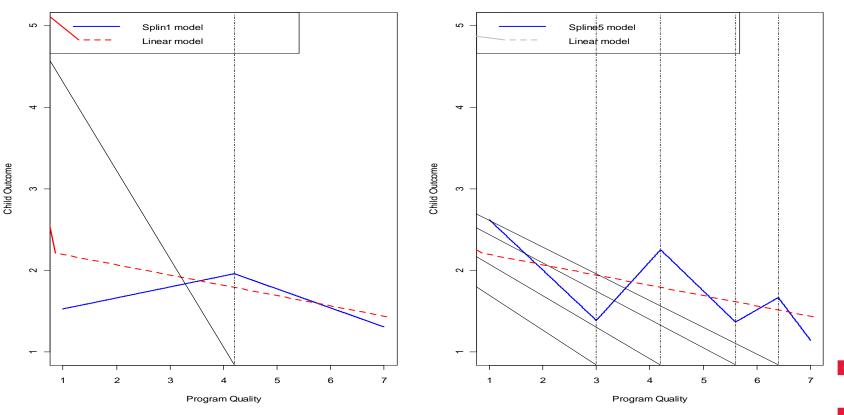


Example 2 – Spline Results

• Comparing possible models

Fitted linear model vs. spline1 model

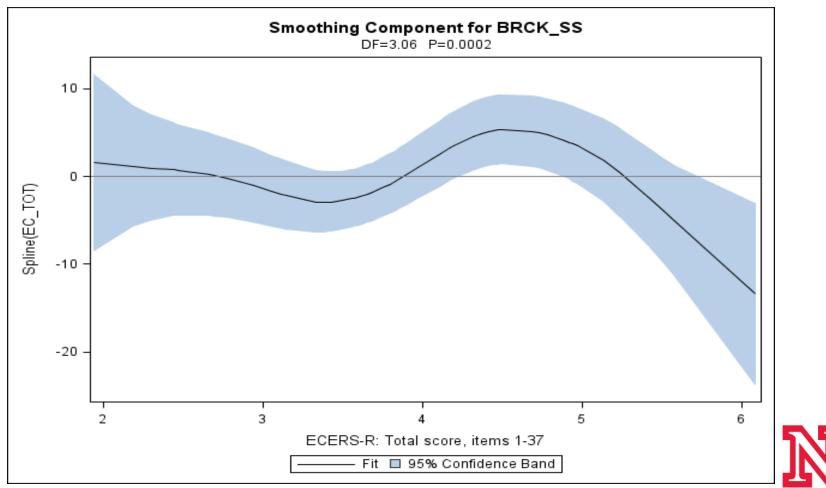
• Practical?



Fitted linear model vs. spline5 model

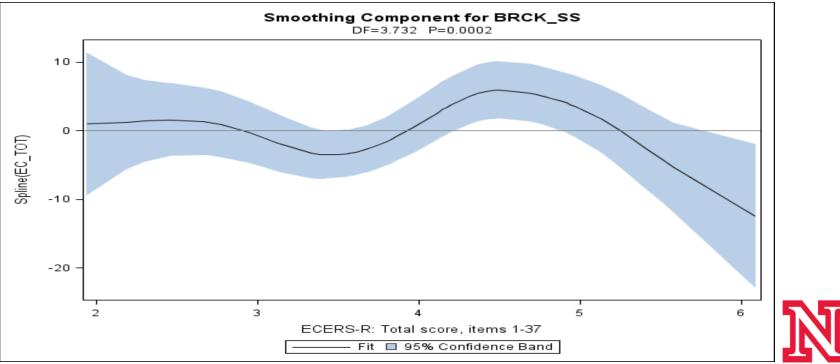
Example 3 – GAM Plot

• Bracken school readiness and ECERS-R



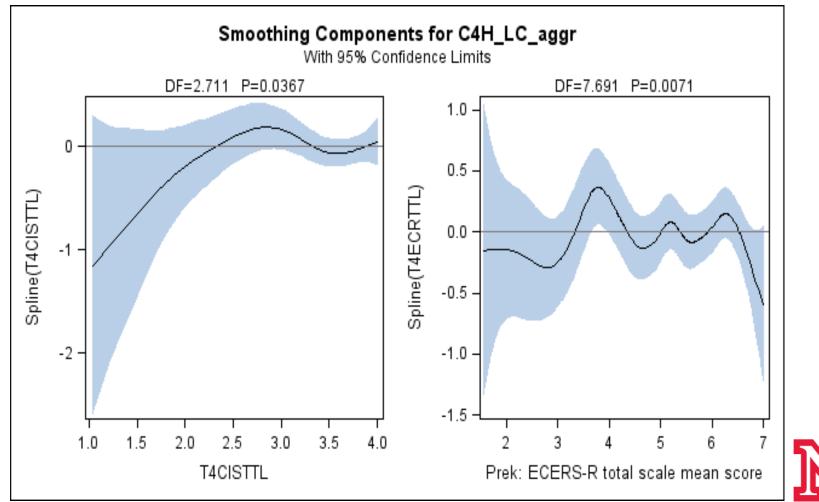
Example 3 – GAM Plot

- Bracken school readiness and ECERS-R
 - by Gender
- Gender significant (p-value=0.0002)
 - Does not significantly impact GAM result



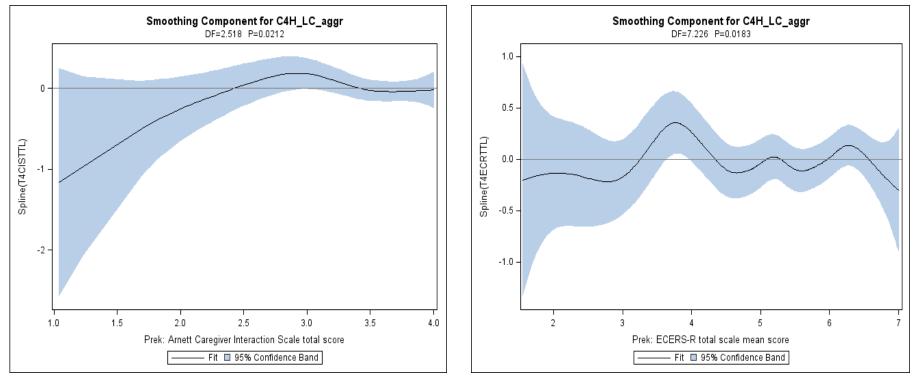
Example 4 – GAM Plots

• ECERS-R & CIS and Howes Ladd aggressive composite



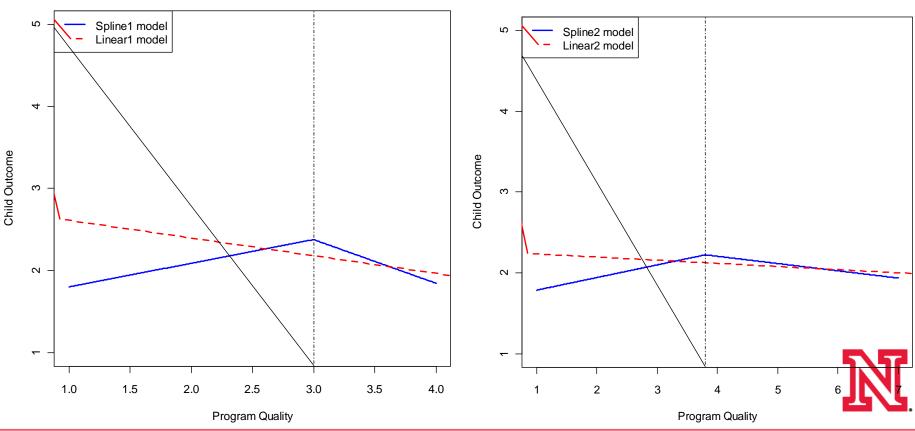
Example 4 – GAM Plots

- What if we look at the relationships separately?
 - The results are almost identical, which is a property of GAM



Example 4 – Spline Results

• Compare the spline model with the suggested thresholds with a linear model on both cases



Linear1 model vs. spline1 model

Linear2 model vs. spline2 model

Utility of the Approaches

- GAM excellent approach to identifying thresholds
 - No a priori knowledge necessary
 - Can be useful though
 - Does not provide for inferential statements
- Spline
 - Can use a priori and/or empirical information
 - Can make inferential statements



Thank You!

Questions?

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