Nikou, Roshan

From: Sent: Graduate.Council.Web.Site@www.uky.edu Wednesday, December 03, 2008 10:20 AM

To:

Nikou, Roshan Price, Cleo

Cc: Subject:

Investigator Report

AnyForm User: www.uky.edu

AnyForm Document: http://www.research.uky.edu/gs/GCInvestigatorReport.html
AnyForm Server: www.uky.edu (/www/htdocs/AnyFormTurbo/AnyForm.php)

Client Address: 76.177.13.44

College/Department/Unit:

= BST701

Category:_

= New

Date_for_Council_Review:

= 12/4/2008

Recommendation_is:_ = Approve Investigator: = Kert Viele E-mail_Address = viele@ukv.edu

1 Modifications:

2 Considerations:

= Course appears straightforward and useful. I recommend approval

3_Contacts: = Adam Branscum general discussion

4__Additional_Information:

AnyForm/PHP3 0.1

AnyFormRandomSeqNo: 36508096

1.	Submitted by the College of Public Health Date: March 10, 2008						
	Department/Division proposing course: Biostatistics						
2.	Proposed designation and Bulletin description of this course:						
	a. Prefix and Number BST 701						
	b. Title* Bayesian Modeling in Biostatistics *If title is longer than 24 characters, write a sensible title (24 characters or less) for use on transcripts: Bayesian Biostatistics						
	c. Courses must be described by <u>at least one</u> of the categories below. Include the number of <u>actual contact hours per week</u> for each category, as applicable.						
	(
	d. Please choose a grading system:						
	e. Number of credit hours: 3						
f. Is this course repeatable? ES 🔲 NO 🗵 If ES, maximum number of credit hours:							
	g. Course description:						
TThis course provides an introduction to Bayesian ideas and data analysis applied to the biosciences. The cour							
	illustrates current approaches to Bayesian modeling and computation in biostatistics.						
h. Prerequisite(s), if any:							
	required: BST 760 (Advanced Regression) and BST 676 (Biometrics II) or equivalent						
	i. Will this course be offered through Distance Learning? If ES, please circle one of the methods below that reflects how the majority of the course content will be delivered:						
	Internet/Web- Interactive Extended campus Kntucky Educational Television Other						
	Please describe "Other":						
3.	Teaching method: ☐ N/A or ☐ Community-Based Experience ☐ Service Learning Component ☐ Both						
4.	To be cross-listed as: Prefix and Number Signature of chair of cross-listing department						
5.	Requested effective date (term/year): Spring / 2011						

6.	Course to be offered (please check all that apply): Fall Spring Summer						
7.	Will the course be offered every year?	□ 1	YES	\boxtimes	NO		
	If NO, please explain: This course will be offered once every three semesters.						
8.	Why is this course needed? This course is a requirement in the proposed PhD in Epidemiology/Biostatistics.						
9.	a. By whom will the course be taught? Adam Branscum or Chong Wang	-		_			
	b. Are facilities for teaching the course now available?		Y S		NO		
	If NO, what plans have been made for providing them?						
10.	What yearly enrollment may be reasonably anticipated? 5-10 students						
11.	a. Will this course serve students primarily within the department?	⊠ ē	YŚ		No		
	b. Will it be of interest to a significant number of students outside the department? If ES, please explain.	_	₹S	_	NO		
	The course will be a requirement for the proposed Ph.D. in Epidemiology/Biostatistics. Some of the students in that program may consider Epidemiology their home department.						
12.	Will the course serve as a University Studies Program course [†] ? If ES, under what Area?		¥S	×	NO		
	†AS OF SPRING 2007, THERE IS A MORATORIUM ON APPROVAL OF NEW COURSES FOR	USP.					
13.	Check the category most applicable to this course:						
	relatively new – now being widely established						
	not yet to be found in many (or any) other universities						
14.	Is this course applicable to the requirements for at least one degree or certificate at UK	⊠ 8	M S		No		
15.	Is this course part of a proposed new program?		e s		NO		
	If ES, please name: PhD in Epidemiology and Biostatistics				- -		
16.	Will adding this course change the degree requirements for AN program on campus? If ES [‡] , list below the programs that will require this course:		e s	\boxtimes	NO		
	<u> </u>						

	[‡] In order to change the program(s), a program change form(s) must also be submitted.									
17.	The major teaching objectives of the proposed course, syllabus and/or reference list to be used are attached.									
18.	Check box if course is 400G- or 500-level, you must include a syllabus showing differentiation for undergraduate and graduate students by (i) requiring additional assignments by the graduate students; and/or (ii) the establishment of different grading criteria in the course for graduate students. (See SR 3.1.4)									
19.	9. Within the department, who should be contacted for further information about the proposed new course?									
Name	: <u>A</u>	dam Branscum		Phone:	257.147	6	Email:	abran3@email.uk	cy.edu	
20.	Sign	atures to report a	approvals:							
	4-1-08			1	hichan	d Kry	scio	y Department Cha	Kupsio	
•	DAT	E of Approval b	y Department Faculty	pri	nted name		Reported b	y Department Cha	ii	signature
		(0-21	0-08	1	Linda	Alex	ander	Kuch	A Alexan	lec
-	D.	ATE of Approva	l by College Faculty	pri	nted name	7.71.54	Reported	by College Dean		signature
,				-	. ,			/		
	* I	OATE of Approv	val by Undergraduate uncil	ргı	nted name	Repor	ted by Und	lergraduate Counc	il Chair	signature
				Be	im A	TH4681	JE	mut !	<u> </u>	
	* D.	ATE of Approva	l by Graduate Council	pri	nted name	Rep	orted by C	Fraduate Souncil	Chair	signature
		8/19/08		i	Heidi	Ander	son	المتعلق	uale	<u>-</u> -
·	*		oval by Health Care uncil (HCCC)	pri	nted name	Reported	by Health	Care Colleges Con	uncil Chair	signatu re
,	* D.	ATE of Approva	l by Senate Council	. <u> </u>		Repor	ted by Off	ice of the Senate (Council	
•	* D.	ATE of Approva	l by University Senate			Repor	ted by Off	ice of the Senate (Council	

BST 701: Bayesian Modeling in Biostatistics

Course Description: This course provides an introduction to Bayesian ideas and data analysis applied to the biosciences. The course illustrates current approaches to Bayesian modeling and computation in statistics. The course will cover the following topics: Bayesian philosophy, prior, posterior and predictive distributions and inference, one and two population models based on normal, binomial, Poisson, and survival distributions, conjugate, informative, and reference prior distributions, linear and generalized linear models, survival analysis, ANOVA, missing data, diagnostics and model selection, aspects of modern Bayesian computational methods including Markov chain Monte Carlo methods (Gibbs sampler and Metropolis Hastings algorithm) and their implementation and monitoring, with applications to biomedical data analysis.

Course Structure: 3 credit hours (2.5 hours of lecture, 1 hour of laboratory)

Prerequisites: BST 760 (Advanced Regression) and BST 676 (Biometrics II) or equivalent.

Initial Offering: Spring 2011

Philosophical Statement: Statistical models are developed for the purpose of addressing scientific questions. For each scientific question that is posed and for which data are collected, the truth is sought by developing statistical models that are useful in this regard. Bayesian statistics represents one of the major approaches to the whole of statistical modeling. Through hands-on experience with real data from a variety of applications, students will learn the basics of designing and carrying out Bayesian analysis, and interpreting and communicating the results. BST 701 will provide students an introduction to practical Bayesian methods for biostatistics data analysis, and illustrate the utility of the Bayesian philosophy in public health. While aimed at biomedical scientists, the methods presented apply to all disciplines in which data are collected to address scientific hypotheses. BST 701 has methodological, computational, and data analysis components. Although statistical methodology will be presented to the extent needed for students to understand the models and methods, the course emphasizes practical applications over theory.

Objectives: On successful completion of the course, students will be able to

- 1. understand the Bayesian approach to statistical data analysis and how it compares to other modes of statistical inference
- 2. design, develop and implement Bayesian methods to address biomedical hypotheses
- 3. incorporate prior information into statistical inferences
- 4. make posterior inferences on parameters of scientific interest
- 5. conduct Bayesian analyses of complex biostatistics data sets
- 6. create WinBUGS programs to fit Bayesian models
- 7. interpret and communicate results from Bayesian data analysis

References: Lecture notes will be provided by the instructor. One of the following may be used as the official textbook:

- 1. Gelman, A, Carlin, J, Stern, H, Rubin, D (2004). Bayesian Data Analysis. Chapman & all.
- 2. Carlin, B, Louis, T (2000). Bayes and Empirical Bayes Methods for Data Analysis. Chapman & all.

Detailed Outline:

I. Fundamentals of Bayesian Biostatistics

Why Bayesian statistics? Approaches to statistical inference and an overview of Bayesian ideas

Probability review

Prior distributions

Likelihood

Bayesian inference: point and interval estimation, hypothesis testing, prediction

Single parameter models: binomial data, Poisson data, normal data Two sample problems

> Comparing means, proportions, rates, median survival times Effect measures: relative means, mean difference, odds ratio, risk ratio, risk difference, rate ratios, and median time to event, survival probabilities

Multivariate normal models, Wishart distribution and Jeffreys' prior Introduction to the WinBUGS software

II. Bayesian Computation

Monte Carlo integration, Law of Large Numbers, asymptotic approximations

Markov chain Monte Carlo (MCMC) algorithms: Gibbs sampling,

Metropolis-Hastings

Adaptive rejection sampling

Slice sampling

WinBUGS for Gibbs sampling

Convergence monitoring and diagnosis: multiple starting values, Gelman-Rubin diagnostic, autocorrelation function, thinning, initial values Inference from MCMC: method of composition, credible intervals, HPD intervals, posterior density estimation, posterior errors.

III. Bayesian Regression

Likelihood construction

Normal regression: Bayesian linear model

Bayesian ANOVA and MANOVA

Bayesian Generalized Linear Models

Logistic regression

Poisson regression

Negative binomial regression

Bayesian residuals

Survival Analysis

Cox proportional hazards models

Accelerated failure time models

Prior distributions: Data augmentation priors and conditional means priors

WinBUGS implementations

IV. Model Selection

Bayes information criterion (BIC)

Deviance information criterion (DIC)

Bayes factors

Savage-Dickey ratio

Pseudo marginal likelihoods and Bayes factors

Bayesian model averaging

Reversible jump MCMC

Predictive model selection

Assessment: Grades will be based on problem sets and lab assignments (50%), a final examination (25%) and class project (25%).

Grading Scale:

Grade	<u></u> %
Α	90-100
В	80-89
C	70-79
E	60-69

LaRoche, Adrea S.

From:

Brothers, Sheila C

Sent:

Monday, September 22, 2008 8:42 AM

To:

LaRoche, Adrea S.

Subject:

FW: HCCC Transmittal - Program Change: MS in Athletic Training

Attachments:

PhD Epi Bio Final Signatures.pdf; FW: important-EPI 714; FW: regarding the New Program

Proposal for the PhD in Epidemiology and Biostatistics

Follow Up Flag: Flag Status:

Follow up Flagged

Categories:

Curricular Items

Don't let the subject line fool you – this is for a PhD in Epidemiology.

Sheila

Office of the Senate Council Phone: (859) 257-5872

From: Lindsay, Jim D.

Sent: Friday, September 19, 2008 2:20 PM

To: Nikou, Roshan; Jackson, Brian A

Cc: Brothers, Sheila C; Anderson, Heidi Milia; Flanagan, Rebecca; Alexander, Linda A; Kryscio, Richard

Subject: RE: HCCC Transmittal - Program Change: MS in Athletic Training

September 19th, 2008

TRANSMITTAL

TO:

Brian Jackson, Roshan Nikou

Graduate Council

FROM:

Jim Lindsay

Health Care Colleges Council

At its August 19th 2008 meeting, the Health Care Colleges Council approved the following proposal and is now forwarding it to the Graduate Council to approve:

College of Public Health

New Program: Ph.D. in Epidemiology

Attached are the materials to implement the requested action.

cc:

Linda Alexander

Becki Flanagan

Richard Kryscio

Shelia Brothers

Heidi Anderson

Jim Lindsay

Health Care Colleges Council Coordinator
Associate Provost for Faculty Affairs Office
University of Kentucky, 205 Frazee Hall
Lexington, KY 40506-0031 Ph. (859) 323.6638
www.uky.edu/Provost/AcademicCouncil/council.php

1.	Sub	omitted by the College of Public Health Date: March 10, 2008				
	Dep	partment/Division proposing course: Biostatistics				
2.	Proj	posed designation and Bulletin description of this course:				
	a.	Prefix and Number BST 701				
	b.	Title Bayesian Modeling in Biostatistics *If title is longer than 24 characters, write a sensible title (24 characters or less) for use on transcripts: Bayesian Biostatistics				
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	(_ (_					
	d.	Please choose a grading system:				
	e.	Number of credit hours: 3				
	f.	Is this course repeatable? YES NO NO If YES, maximum number of credit hours:				
	g.	Course description:				
		TThis course provides an introduction to Bayesian ideas and data analysis applied to the biosciences. The course illustrates current approaches to Bayesian modeling and computation in biostatistics.				
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		Internet/Web- Interactive based video Extended campus Kentucky Educational Television (KET/teleweb)				
		Please describe "Other":				
3.	Tea	aching method: N/A or Community-Based Experience Service Learning Component Both				
4.	To	be cross-listed as: Prefix and Number Signature of chair of cross-listing department				
5.	Rec	quested effective date (term/year): Spring / 2011				

6.	Course to be offered (please check all that apply):	mer					
7.	☐ YES ☒ NO						
	If NO, please explain: This course will be offered once every three semesters.						
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	11 110, white plants have been made for providing them.						
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*If applicable, as provided by the University Senate Rules. (http://www.ukv.edu/USC/New/RulesandRegulationsMain.htm)

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