

STATISTICAL TRAINING AT THE UNIVERSITIES—SOME VIEWS

S. RAO JAMMALAMADAKA

Department of Statistics and Applied Probability
University of California, Santa Barbara (USA)

1. INTRODUCTION

Statistics, as a discipline which incorporates inductive reasoning, is one of the major building blocks on which new scientific advances are based. Statisticians help scientists and others make a guarded leap of faith from the (observed) sample to the (incompletely observed) entirety and thus play a pivotal role in the progress of science.

In dealing with uncertainties around us in everyday life, from weather prediction to forecasting the performance of our investments, we all "play statistician". So it is very appropriate that everyone of us acquire some "statistical literacy"—just to be reasonably informed citizens. This exposure to basic ideas of probability and statistics should take place at high schools, and it is gratifying that this is indeed happening in many countries. This topic has been adequately dealt with elsewhere (see e.g. Råde (1981)) and I shall not dwell on it here.

My objective is to share with you some ideas regarding statistical training at the university level, i.e., in training Bachelors, Masters and Ph.D. degree students. These views evolved over a period of twenty-five years during which I have been involved in statistical education at the university level—mostly in the USA—and from my own statistical training which has been at the Indian Statistical Institute, Calcutta, under such people as C.R. Rao, P.C. Mahalanobis and J.B.S. Haldane—to name just a few.

I believe that, irrespective of whether one is training a student for a Bachelor's degree or a Ph.D. degree, the training in statistics should always involve an appropriate mixture of both theory and applications. Statistics derives its strength, usefulness as well as research motivation from applications and from the practical problems arising in other fields. It is the most interdisciplinary or multi-disciplinary subject. Given this, our main challenge is to find this "proper mix or balance between theory and applications. This balance, in my view, depends very much on what we are training the students for—training people to fight the fires or training them to invent and build new fire-fighting tools. We do, of course, need both kinds of training. The majority of those we train at the universities at the Bachelors and Masters level work for industry and government and may be classified as the "firefighters". They should know how to use the given tools effectively and know their limitations. These students should concentrate more on acquiring a broad knowledge of the statistical techniques along with the applications (including

computing, consulting and communicating) and less on the theory. On the other hand, one needs to understand a large body of theory and possess good mathematical skills to develop new statistical tools. However, even these “theoretical-types” who are working say for the Ph.D. degree, should constantly bear in mind the immediate or at least potential applications of these new statistical tools. These new tools are meant not so much to enrich and embellish the subject of statistics but to solve practical problems dealing with the real world.

Although this viewpoint is not new, I believe it bears repeating and will be addressed in a bit more detail in the next section. In Section 3, these ideas are given a more concrete shape in the form of an outline of courses etc. and related to our experience at the University of California, Santa Barbara (UCSB), in starting the B.A./B.S. degrees in Statistical Science, M.A. (Statistics) and Ph.D. (Statistics).

2. THEORY AND APPLICATIONS

Mahalanobis (1965), echoing a theme of Fisher, describes statistics as a “key technology”—much like medicine or engineering. He distinguishes science as “the effort to know nature more adequately” and technology as “the effort to use scientific knowledge for the fulfillment of specific purposes” and says further that “A scientist may have a wide or narrow range of interest or specialisation. (But) a technologist must have knowledge and experience of a wide range of scientific subjects”. He quotes Fisher who pointed out that, “a professional statistician, as a technologist, must talk the language of both theoreticians and practitioners”. I fully agree with this viewpoint and believe we should implement this philosophy in training statisticians, to the maximum extent possible. However from my own personal experiences as a student, I should caution against the possibility of getting carried away too far, in providing this broad-based education to statisticians. In 1960, I was among the first batch of students selected for a Bachelor of Statistics (B.Stat.) degree at the Indian Statistical Institute—a degree for training “professional statisticians with a broad base”, formulated by Fisher, Mahalanobis, Haldane and C.R. Rao among others. A somewhat detailed description of this degree program is contained in the article “A multidisciplinary approach for teaching statistics and probability” by C.R. Rao (1969). While this degree program has been highly successful (and continues on a more modest level, even today), it might have been overly ambitious in trying to expose students to the multitude of subjects where statistics finds applications. This meant the students, besides learning mathematics and statistics, had to learn the quantitative techniques and the methodology used in a wide range of subjects including economics, biology, geology, genetics, engineering, chemistry, physics, psychology, sociology, demography etc. etc. To top it off, Professor J.B.S. Haldane taught us a course which he simply termed “General Science”—but to us, it meant an endless hopping from one subject to another—from discussions on space ships to

mining, from gerontology to philology, organic chemistry to microbiology and on and on. It was certainly a most laudable attempt at training statisticians with a broad base—a most ambitious one.

On a much more modest scale, I advocate that all undergraduate training in statistics should incorporate at least one (preferably two or more) application area(s). If one holds the idea that statistics is a “technology” like medicine or engineering, one should develop a speciality in one or two application areas, like an electrical engineer or a pediatrician. Statistical ideas could be learnt in the context of this speciality and this provides both motivation and a concrete setting for developing the statistical theory and concepts. For instance, in U.S. universities, this can be achieved at the Bachelors level through what is called a “double-major”—majoring in statistics alongside say economics or biology. If it takes a bit longer to obtain a double-major, so be it. One does not get a degree in medicine or engineering without putting in some extra time and effort.

In the U.S. where students typically spend four years to obtain a Bachelors degree, I believe the students should spend approximately equal amounts of time and effort on statistics during the final three years—the first year is mostly lost to general education requirements. Under (a) mathematics, (b) computing, (c) statistics and (d) application area(s). In mathematics, students should have a year of calculus and a year of linear/matrix algebra. If time permits additional courses on advanced calculus, real and complex analysis, numerical analysis, operations mathematics and operations research should be recommended. Students should have sufficient mathematical ability to work with e.g. Jacobians of transformations, linear models in matrix notation etc. Under (b), they should be proficient in computers and be able to write programs in one or two languages. Also to be included here is a course on “statistical computing” which teaches students at least two software packages for statistical analysis. Given the current job market, the more the students pack in here, the more employable they are. In statistics, the courses should include (i) a one-year course on probability and mathematical statistics which typically includes derivations of sampling distributions and parametric and some nonparametric inference, (ii) a course (of at least half-year duration) on linear models and generalized linear models with discussions on regression and analysis of variance and (iii) at least a basic course on sample surveys and on design of experiments. Other optional courses offered include time-series analysis, nonparametric methods, quality control and reliability, simulation methods. Finally, under (d), application areas, students should take some “serious courses” in the subject area(s), specifically those that focus on quantitative methodologies used in that area(s).

At the Masters degree level, it should be possible to separate students into at least two main streams depending on their goals and interests. For instance, an Applied Statistics

and a Mathematical Statistics stream. While the Applied stream trains students for jobs in government and industry (the “firefighters”), the other stream emphasizes more on theory and mathematics and prepares them for a Ph.D. Students doing Masters in applied statistics should possess a broad knowledge of a wide range of statistical techniques and methods—applied multivariate analysis, categorical data analysis, some biostatistics including survival analysis, sampling methods, time-series etc. They should also be able to effectively utilize statistical packages for data analysis, be able to solve general statistical problems from industry or government as consultants, and communicate with other scientists. At UCSB, we found that students who obtain their B.A./B.S. degrees in one of the substantive areas (e.g. biology or psychology) are equally successful at pursuing this stream, if they are willing to make up the deficiencies. There is an excellent report by the American Statistical Association Section on Statistical Education on this type of training (ASA (1980)) that I would recommend to your attention.

The other (Mathematical Statistics) stream consists of potential researchers—those who intend to obtain a Ph.D. in statistics. Their coursework should emphasize more statistical theory (optimality of various procedures, nonparametric and robust methods, decision theory etc.) and additional courses in probability theory, stochastic processes as well as other relevant mathematics courses.

Students who wish to pursue a Ph.D. degree should develop considerable breadth in statistics and familiarity with applications, before they are allowed to narrow down to their chosen research topic. This might be achieved, for example, through various course requirements and/or “qualifying” examinations. It is unfortunately still possible to get a Ph.D. in statistics in most U.S. universities and not ever be exposed to basic ideas of survey sampling, design of experiments, time-series, computing or even real applications. I quote here again from Mahalanobis (1965) about a Ph.D. from a well-known U.S. university who was seeking a job in India: “the results of his investigation were given in an abstract mathematical form and their implications were not immediately clear to me. As the post was for work mostly at an operating level, I asked him whether he had any occasion to use his results in practice. He was surprised at this question and said ‘no’. I asked him whether he knew or could think of any situation in which his results could be used in practice; he thought over my question for a while and then again said ‘no.’” This state of affairs has changed very little. To quote from Box (1980), “the student may receive verbal and nonverbal messages that tell him that the kind of Ph.D. thesis that is motivated by a scientific problem with real data is not likely to be so acceptable as a pretty narrowly introspective mathematical exercise. Again the junior faculty member knows that he has only a few short years to make it to tenure. In many cases he will be judged by faculty who have no understanding of, or sympathy for, application. He may suspect that when his

publications are assessed, those with any taint of application will be down-weighted or entirely discounted." Students often get a Ph.D. degree in statistics for abstract exact "statistical mathematics" which have no connection to real data analysis. The thesis committee and supervisors should make certain that the research topic the student chooses to work on has potential use if not of immediate applicability. This test of "applicability of results" along with "knowledge of computing", should be made requirements for all students doing the degree in statistics.

Students of statistics, no matter what the level, should get this sense that statistical theory as well as the need for new theory is driven by applications. The best place to teach applications is alongside the relevant theory regardless of whether it is an elementary or an advanced course. Teachers do not always find this easy or convenient. Most of us are used to tossing coins and dice, for 'concrete' examples. It seems that the training (or retraining) actually begins at the faculty level, with the faculty developing interests in some applications. This can be achieved through joint appointments or by faculty from statistics departments on temporary assignments to government or private agencies, to acquaint themselves with real world problems. Participation in a statistical consulting laboratory at the university (or even at the university) can be both an educational tool for the students and an extremely useful resource for the faculty (or the university) and/or even private consulting are other ways for statisticians to keep in touch with reality.

3. SOME EXPERIENCES AT UCSB

Courses in statistics have been offered at the University of California, Santa Barbara (UCSB) for over 15 years, mostly as part of the Mathematics curriculum. In the Fall of 1970 a separate "Department of Statistics and Applied Probability" was created at UCSB. The Department now has close to 14 faculty members and offers:

- (a) B.A. and B.S. in Statistical Science;
- (b) M.A. (Statistics) with three streams leading up to Applied Statistics, Mathematical Statistics and Operations Research specializations; and
- (c) Ph.D. (Statistics).

The two Bachelors degrees require a substantial amount of coursework in mathematics, computer computing and statistics, as described in the earlier section. However the degrees do not, in time, require coursework in application areas that I discussed before. We hope we can in the future offer several double-majors, as for instance, statistics and economics, statistics and computer science, statistics and biology. The B.A. degree requires 10 one-quarter (a quarter consists of 10 or 30 fifty-minute sessions of instruction) courses in statistics at the upper division (3rd

year) level, while the B.S. degree requires 13 such one-quarter courses. Following is an abbreviated list of courses (all are one-quarter, unless mentioned otherwise) categorized as:

- Requirements:** Probability and Mathematical Statistics (3 quarters)
Design and Analysis of Experiments
Regression Analysis
Statistical Computing
- Options:** Sampling Techniques
Statistics in Industry
Sequential Methods
Nonparametric Methods
Operations Research
Ranking and Selection Methods
Applied Stochastic Processes (2 quarters, includes some time-series)
Actuarial Statistics and Risk Theory (3 quarters)
- Others:** Internship in Statistics
Independent Studies in Statistics

While the “independent studies” course is a vehicle for learning any other topic that is not taught as a regular course, the internship encourages students to do faculty-supervised academic internship in industrial or research firms in the area. This is an excellent way for students to relate the academic coursework to real problems that exist out there.

An M.A. degree with any of the three specializations requires about 11 one-quarter courses (mostly graduate with a few approved upper division courses allowed). An abbreviated list of graduate courses (all are 3-quarters unless mentioned otherwise), currently offered include:

- Statistical Theory
- Multivariate Analysis
- Statistical Decision Theory
- Linear Models
- Life Testing and Reliability (one quarter)
- Operations Research
- Case Studies in Operations Research (one quarter)
- Advanced Statistical Methods
- Seminars and Projects in Statistical Consulting (one or two quarters)
- Probability Theory
- Advanced Probability Theory
- Stochastic Processes
- Advanced Stochastic Processes

Then we have a catch-all course called “Seminars in Probability and Statistics”, where the

topics (generally advanced) vary with the instructor's interests.

Students in M.A. (Applied Statistics) take the Advanced Statistical Methods course which covers a wide range of topics (not always with formal proofs) and have to do an internship in the statistical consulting lab for a quarter or two. Here they participate in actual consulting, under supervision, and write a project report on their statistical analysis and conclusions, where the students are exposed to the 'inconvenient' real-world problems which may violate traditional assumptions, have missing data and may indeed call for the development of new theory. The operations research stream requires a similar case-study project after the year-long graduate course in operations research. Students in the Mathematical Statistics stream take the more theory-oriented courses (statistical theory, probability theory, etc.). Students are encouraged to keep up their interest in an application area by providing academic credit for courses taken outside the department.

To make sure that the Ph.D. students have the appropriate breadth, they take "qualifying examinations" in mathematical analysis (real and complex variables), which is compulsory and in any three of the following five areas: Mathematical Statistics, Mathematical Statistics, Operations Research, Stochastic Processes and Probability Theory. We still have foreign-language requirements and one of them can be fulfilled by demonstrating proficiency in a computer language—almost all the students opt for this and do a rather substantial amount of computing. Even those who opt for a very theoretical piece of work for their Ph.D. thesis are asked to answer my colleague Joseph Gani's familiar question during their thesis defence: "How do you apply your results to any data sets, real or simulated?" So all in all, I believe we have a good training program that needs only minor fine-tuning and is mostly in line with my recommendations.

4. CONCLUDING REMARKS

Although there is no unique formula which works for training all categories of statisticians, there is, I believe, general agreement that statistical training as well as statistical research, should be driven by applications. We are in the fortunate position of being able to motivate all the theory we teach, through real-world applications and we should encourage synergy between theory and applications to our students. A good understanding of the theory, along with its limitations, an appreciation for applications and familiarity with "statistics and computing" should form the basic themes for any university degree in statistics.

Statistics is a discipline which is growing both in scope and depth, especially during the last few decades. This makes it imperative that we re-examine the courses and curricula on a periodic basis and update them, as necessary. For instance, the tremendous power

computer which is now at our disposal, has revolutionary implications for our discipline and we should take advantage of this. Computer intensive methods like resampling techniques including bootstrapping, iteratively reweighted least squares, density estimation, simulation, dynamic graphics, etc. have all become topics of great interest. See for instance Diaconis and Efron (1983), Speed (1984). Also it is possible to do a large part of data-scrutiny and exploratory data analysis through interactive statistical graphics packages, i.e., plot, look for outliers and try various transformations etc. See for instance Tierney (1990). We should incorporate these topics as well as a host of new and emerging theoretical fields like inference for stochastic processes, robustness and influence functions, directional data analysis, spatial statistics, image reconstruction, interacting particle systems, etc. into our curricula alongside the more traditional courses that we already offer.

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SUMMARY

This paper presents some views regarding the training and academic preparation of statisticians at the University level. Statistics is a "key technology" which helps solve problems arising in a wide variety of disciplines. Therefore our goal should be to train students who not only understand the mathematical methods and the theory but also have a clear appreciation of the practical problems and applications in at least one or two subject areas. The proper balance between theory and applications will, of course, vary depending on the students' goals and needs. The argument is made that there should be no dichotomy splitting the profession into theoretical statisticians and applied statisticians. There is no room in the profession for a theoretical statistician whose work has no relevance to the real world or for one who cannot grasp the basic theory, assumptions and limitations behind the techniques. Some general comments regarding the coursework and a comparison with the degree programs at the University of California, Santa Barbara, are provided.

RESUME

Cet article traite des idées concernant la formation des statisticiens. Notre but devrait être d'enseigner aux étudiants à comprendre la théorie des statistiques et à apprécier les problèmes pratiques. On affirme que la profession ne supporte pas une dichotomie entre ceux qui s'intéressent à la théorie et ceux qui s'intéressent à la statistique appliquée. Finalement, on décrit quelques idées concernant les cours à l'Université de Californie, Santa Barbara.