

Discn. w. Pat. Winston re: my staying at MIT past March 31, 74!

He discussed it w. Marv.

My impressn.: W. doesn't feel that ind. Inf. research is terribly relevant to A.I. — so, as long as I'm working on [GAI], he'd

rather not have me here. If t. nature of t. project "changed" he would perhaps, reconsider.

Mr. Marvin likes to support **A I** on a broader front —

so he'd like me to stay, even if I work on induction theory.

I think M & W tentatively decided that M. & I might go out

and hunt for funds for me a lone, pretty much. As W. puts it

[W. doesn't like to get funding from much > 1 source, because of additional work involved.]

"As long as they have room for me on t. 3rd floor — it's O.K. w. him"

— t. implication being that as far as W. is concerned,

he has no great desire to keep me here, but ~~if~~ as long as I don't do any harm, it's O.K. w. him. At any rate —

W. seemed ~~rather~~ hostile! a.g. sop up any funds or room, @

So, anyway — discuss this w. Marv. — 05 as well as

how to go about shopping for funds. Mention Licklider

as good person to talk to as far as suggesting Gov.

agencies. RLE Brooks

Discn. w. Marv.: Apparently there is a big money shortage now.

Also much disagreement on ratio of theoretical to practical A.I. work. —

W. Marvin & Seymour ~~are~~ for more theory & Pat for more practical.

Anyway M. suggests I write proposal. He will discuss with Seymour,

sources of funds. Size of project to be presented (i.e. no. of

people on it) will depend on how much money agency wants

to spend. Preliminary proposal draft: What I have done

in induction, what needs to be done, what I intend to do,

and why it is important.

Proposal:

01.66.90
(613.90)
spec

Training Sequences for Mechanized Learning. (perhaps better title)

Project for research in Mechanized ~~Learning~~ Learning.

Introduction:

The background of the present

~~For many years, the~~

project, is the theoretic analysis of the induction ^{process} ~~process~~

— the process by which a person with ~~perception~~ is able to discover regularities in the world about him, and use ~~the~~ ~~these~~ these discoveries ~~to make~~ ~~prediction~~ on the successful prediction of new phenomena he has not yet seen.

~~In ~~summary~~.~~

The project itself will consist of the construction of ~~learning sequences from a variety of fields,~~ ~~and~~ sequences of problems of increasing difficulty, from various fields, and the ~~construct~~ parallel construction of machines that are able to solve these problems ~~as they~~ as they are presented.

The present theory of inductive inference ^{problem} makes it possible to devise such sequences.

It tells us how much of a "heuristic jump" is ^{needed to solve} ~~contained in~~ a new problem, if we understand just how the machine has solved problems up to that point. The size of the "heuristic jump" tells us just how much

hard ware and/or ~~hardware~~ time, the machine is expected to need to find a solution to the new problem.

At first the problems will be chosen from ~~well~~ well understood areas in mathematics and formal linguistics. As the total teaching sequence grows larger and the machine has incorporated many ^{heuristic} ~~heuristic~~ devices into ~~its~~ its discovery mechanisms, it will be possible to give the machine problems that are not so completely understood from a greater variety of fields.

Why this work is important (depends on agency)
 possible applications (military, non-military,

depending on agency)

Applications, importance: (1) Devising the seqs for ~~the~~ people is aided by analysis of machines' learning of seqs. (2) Steps toward very high power machine that could learn to understand English text is other data that has not been very carefully designed text. (perhaps go into this at some length - how, ordinarily, it is very difficult to ~~teach~~ teach machines things - one must be very careful in devising the seqs.

① Previous work in induction theory is what it can do
 It can tell us if a discovery procedure is adequate. ② It suggests
 ways to devise discovery procedures.

② ~~What work on generalization~~ Much of the ~~work~~

V.G. work ~~on~~ in induction in H.F. has been

based upon introspection & protocol analysis. Plans written were
 attempts to have E. way humans do.

Induction theory gives us a uniform ~~method~~
 set of constructs for expressive ~~and~~ such induction systems -
 for criticizing them, for improving them. It

makes it possible to select features from several different
 induction programs and combine them to make a

program more powerful than any of the ~~original~~

programs.

③ The goal of the research is not simply to find good ways

to do mechanized induction. - Rather it is to find such

methods that are discoverable by a machine that has

had an ~~acceptable~~ acceptable training sequence.

We want a machine that learns how to do induction from

a training sequence that is ~~relatively~~ specific

oriented toward teaching specific induction techniques.

It is only in this way that we can obtain a machine

that will learn by its experience with problems

that have not been selected with the goal of

training the machine in any way. It is such

a machine that would be most able to deal with the

& greatest variety of problem appearing in the real world.

01:68.40 : (1) Previous work in induction theory: what it can do
 — It can tell us ~~if~~ a discovery procedure is adequate. (2) It suggests ways to devise discovery procedures.

(2) ~~Most work on getting a machine~~ Much of the ~~work~~
 V.G. work ~~in the area~~ on induction in A.F., has been based upon introspection & protocol analysis. Programs written were ~~an~~ attempts to have it. way humans do.

Induction theory gives us a ~~uniform method~~ uniform ~~method~~ set of constructs for expressing ~~the~~ such induction systems — for criticizing them, for improving them. It makes it possible to select features from several different induction programs and combine them to make a program more powerful than any of the ~~original~~ original programs.

(3) The goal of the research is not simply to find good ways to do mechanized induction. — Rather it is to find such ~~methods~~ ^{themselves} that are discoverable by a machine that has had an ~~acceptable~~ acceptable training sequence. We want a machine that learns how to do induction from ^{not} a training sequence that is ~~relatively~~ ~~non~~ specifically oriented toward teaching specific induction techniques. It is only in this way that we can obtain a machine that will learn by its experience with problems that have not be selected with the goal of training the machine in any way. It is such a machine that would be most able to deal with the ~~the~~ greatest variety of problem appearing in the real world.

01: 73.90 : (9) (a) The Nature of t. (Ind. Int.) Problem: Under Induction, learning. How it is done by ~~Animals~~ living creatures - How to do it mechanically.

USF?
Ford
Foundn.

(b) Why it is an impt. problem: ~~a~~ - Understanding human learning ^{in humans} for better teaching - Understanding ^{creative} / discovery process so we can devise envts., tng., to encourage it - ~~Designing~~ ~~machines~~ ~~for learning~~ ~~under cond~~ in which it is not feasible for humans to ~~employ~~: e.g. Extraterrestrial

NASA extraterrestrial envts for Nasa (by accel., by radiation, by pressure

NIH ^{low temperatures} by temp. / , etc); Medical Learning / over very large population over very large period of time. Well beyond capability of any human. Present statistical techniques employ simple models for induction - are very limited in their capability. We want to get much more clever induction for very large amounts of data. A human Md. dies with his experience, for t. most part not passed on - ^{a potentially infinite} data taught. We want to retain t. learning of all lifetimes.

ARPA
DOD
Army
Air Force

For making intelligent decisions much rapidly than humans can ^{attach in a way that was not anticipated, not too brittle}
(Nuclear Attack - Explain how / machine could be used ^{to get confidence in it.} How Machine would have to be tested under various simulated roads. compensates for lack of brilliance, by extreme speed)

Learning to recognize signatures of Missiles of various types.

Learning to recognize Nuclear test v.s. earthquakes.

Navy
Army
Electronics
Background

Integration of Data from many modalities (IR, Radar, sonar, intelligence reports) very rapidly, to infer position of Nuc. submarines, missiles, missile bases,

NIH

A very fast machine could ~~reach~~ ^{reach} thru skin of a patient and remove an appendix in less than a second, ^{making exploratory observations of}
- On the way in, observing various conditions that may or may not make an appendixotomy the best thing to do, perhaps deciding not to do it, after 2 or 3 milliseconds penetration.

01: 74.90: NASA ; (} Very rapid emergencies occurring in the piloting of aircraft or spacecraft.

Human is too slow to respond, Situation is too complex for routine programming for foreseeable emergencies.

- Industrial production
- Detecting Epidemics at early level
- Info Retrieval
- General Assistant (Getting from Mars)

As photos

NIH : / A central medical intelligence system, an induction machine could ~~data~~ discover an incipient epidemic by ~~then~~ noticing various ^{apparently} ~~is~~ isolated symptoms in a great number of people over a large population.

The ~~interrelation~~ ^{correlation} of such symptoms with weather conditions, ~~the~~ ^{potential} disease counts ~~in other~~ ^{countries} would not be detected by ~~usually use~~ ^{ordinarily} available statistical methods. The induction techniques ~~can~~ ^{can} be used to obtain the best possible predictions from the available information, using

the ~~available~~ ^{available} computing resources available. Suspicion of an epidemic could be ~~confirmed~~ ^{confirmed} by ~~various~~ ^{various} further medical tests. Such a central medical system ~~can~~ ^{could} recommend various measures to ~~prevent~~ ^{prevent} the epidemic well before it ever got started.

Industrial Use of robots for industrial production

would be much facilitated if they could be taught a very complex task in the way a human is taught by watching someone else perform the task. At the present time, robots are very limited in what they can do.

Tasks involving complex interaction of effectors and sensors are usually well beyond their capabilities.

In cases where such interactive behavior is possible, ~~the~~ ^{each} ~~specific~~ ^{job} tasks must be ~~programmed~~ ^{specifically} programmed.

A very difficult and time consuming task. An inductive machine could learn such tasks very quickly after ~~observing~~ ^{observing} a human perform the task a few times.

1. 77.40
75.40 spec.

The Generalized informational assistant (Charvillat's idea)

TM78

Council
in Lib.
Resources
(Fund Found)

This is an info retrieval device, in which one describes one's problem (or what one thinks one's problem is) & it.

To Mrs. 28
630-730P
Adv. corp
195 Albany St
Cambridge

machine asks Q's of one & makes suggestions of ideas areas or of things to read or of relevant discoveries, activities of people, inventions, conferences, places to go, kinds of expts. to try (both Gedanken & actual). This assistant would also help as a "memory aid." One would describe it & thing one wanted to think of, along w. various probable contours & it would ask Q's then try to guess what was wanted.

Demo of
52'x69"
proj
color TV.

To ~~forget~~ over all general A.I. goals. Indirect research

is one area of AI in which any success can usually be translated into useful progress in most any other area of A.I.

POD
ARPA

Learning to integrate various environmental & political & economic symptoms to decide whether an enemy country was preparing for an attack on some other country.

General emphasis of abilities of high level learning machines to deal w. unforeseen events, for which they were not specifically programmed, & give list of situations in which this would be vital.

For Motivation, see: New Scientist. 10 Mar 66: "New Learning Machines for Future Aerospace Systems. (D.R. Morris, A.C. Spake, - USAF Avionics Lab. W. Pitt ~~AFB~~ AFB, Ohio. pp 626-628

➡ ① Present state of Ind. research; How long will it be before we get progress? Just why we are more likely to make breakthroughs now.

② On George's work: How big problem was to obtain probys to be used in decision theoretic Model. How now Induction Theory is in some position to give real help with this problem - perhaps solved completely.

(Govvy wrote 3 proposals, all rejected - perhaps because of Nixon's misunderstanding of funds. - New funds have been restored. to NIM)

1) Perhaps work w. Govvy to get Money from NIH
Phone Passy to ask her who to talk to.

5m x 360 = 1.8 x 10⁹

The she may not know specifically, she can give info on the hierarchy.

2) Discuss ~~my~~ my early work in induction (1956 IRE) ~~What its problems were~~ (Utility evaln. - no basic understanding of induction theory)

3) Discuss ~~more~~ more modern methods of dealing w. P&S problem - its successful applicn. to some aspects of grammar induction, where other methods have failed. Its "adequacy" in ~~the~~ SS error sense.

4) The proposed 3 pronged attack on ~~probabilistic~~ induction:
a) Basic top. seq. starting w. very simple concepts. use of ~~math~~ math, Alg, math, a formal linguistics in top. seq. where material is very well understood.

b) ~~Top~~ Top. Seq. ~~is~~ of an ongoing complex problem. Attempt to find "factors" for the concepts used. What ^{we have} ^{very} ^{clear} defn. of adequacy in this context.

Use of Med diagnosis, induction of concepts from "black world", ~~from~~ from Chess, from Electronic ckt analysis, from ^{computer} ~~computer~~ Pgm debugging.

c) Continued work on basic problems in induction theory that are most relevant to applications. e.g. (1) Better methods to combine perms. (2) Optimum ^(least cost) soln. of theoretical problem of finding best changes in "production" in response to a gn. "A corpus". (3) Work on ~~the~~ theory.

Boony ^{thru} ^{530P}

$$\text{flux} = M \times W_1$$

$$u^{-1} = m \frac{1}{2} g^{\frac{1}{2}} t^{-1}$$

$$hV = au$$

$$u^{t-1} = m \frac{1}{2} t^{-2}$$

$$h = m \frac{1}{2} g^{\frac{1}{2}} t^{-1}$$

$$e =$$

$$\frac{h}{Q} = \frac{m \frac{1}{2} g^{\frac{1}{2}} t^{-1}}{m \frac{1}{2} g^{\frac{1}{2}} t^{-1}}$$

$$Q = u^{-1} m \frac{1}{2} g^{\frac{1}{2}} t^{-1}$$

$$\frac{h}{e} = \frac{m \frac{1}{2} g^{\frac{1}{2}} t^{-1}}{u^{-1} m \frac{1}{2} g^{\frac{1}{2}} t^{-1}}$$

$$= u^{-1} m \frac{1}{2} g^{\frac{1}{2}} t^{-1}$$

labor flux

$$m \frac{1}{2} g^{\frac{1}{2}} t^{-1} Q^{-1}$$

al change = Q

$$\text{flux} \times e = m \frac{1}{2} g^{\frac{1}{2}} t^{-1}$$

$$= \frac{h}{t} = e$$

80.01

3374 Pro

Solomonoff (Mar 67)

TMSO

Proposal for the Application of Recent Progress
in the theory of Inductive Inference to Practical
Problems in Machine Learning, Prediction and Training
Sequence Generation.

The last decade has witnessed important progress in understanding the theoretical mechanisms of inductive inference. Research by Chaitin, Kolmogorov, Solomonoff and Willis, has clarified much in an area that had up to this time been dominated by philosophical speculation, intuition and inspired guesswork.

Although these new results have been applied to some practical problems - such as linear regression and curve fitting, the goal of these studies has been not to obtain new techniques, but to show that the new formulations of induction yield correct solutions for problems that are already well understood.

The proposed applications will be of several kinds. In previous artificial intelligence work programs have been written to solve problems in mathematics, linguistics, and electronics, and to play games such as chess and checkers. These programs have been very specific, in the sense of being devised for very narrow task environments, and being of little value for slightly different kinds of tasks.

The proposed work in induction will be toward machines that are able to work over a much broader spectrum of problem environments - machines which learn

3319 pro

2

usually

TMI
31

to work problems by being shown an adequate set of mostly correct examples.

Much of the previous work of this sort has met with very limited success.

shorter In some cases - i.e. perceptron-like machines - Minsky and Papert have made the nature of the limitations quite clear - the machines are linguistically incapable of expressing the needed concepts.

Induction theory makes it very easy to construct machines that do not have this linguistic limitation.

Another difficulty with previous induction machines (e.g. those of Uttley and Rosenblatt) is that the number of examples needed to learn a particular concept was inordinately large. In some cases, the machine would have to be shown all cases of the concept in order to be able to learn it - so that for such problems the machine performs no inductive inference at all.

Again induction theory permits us to deal with this difficulty in a most general way. It tells us the minimum number of examples that are needed to discover a particular concept (Schubert 1973, Willis 1970) as well as what sort of training environment is needed, and suggest algorithms for most rapid learning.

These proposed investigations involve programming a machine so that it can learn to work problems in a particular field.

pro

will

TM 82

A somewhat more general approach that will be pursued would train a machine just as a child is trained - by starting out with a very simple set of problems and by a suitable sequence of training examples, bring the machine to any needed level of competence in a particular field. The initial machine need have fewer induction techniques built into it than in the previously described machines.

Again, induction theory tells us which sets of built-in techniques are adequate for this kind of learning. It is felt that this sort of machine would probably be able to work problems better than those previously described, and from a much broader field of problem environments. Its training sequences would, however, be much longer and as in the case of a human being most of the training would not be in fields directly related to the problems the machine would ultimately be solving.

A third area of research will be continued work on fundamental problems in induction theory that are most relevant to applications. One of these problems involves optimum methods of combining various successful induction techniques.

Another concerns optimum methods of computation to obtain the best possible results in the least amount of time, using any specific hardware configuration, or ways of finding optimum hardware to solve the particular kinds of problems that occur in induction. These hardware-software optimization problems occur

in all areas of practical computation, but in induction theory, it appears to be easier to state the problem in a general, unambiguous way that suggests promising approaches.

01:83.90: My present idea of this proposal:

- 1) Abstract (short summary)
- 2) Introduction (longer summary - also tells what the various sections are about)
- 3) Main body: Various sections that were ~~summarized~~ ^{summarized} (in 2)
- 4) Appendices: Papers & Reports & "working papers" relevant to the proposal. [I.E.S. review: willis: Tug says (1962) perhaps my working paper on induction results.]

The main body describes the problem, tells a little why it's important & gives some applications. (The main thrust of why it's important & applications, will be in an auxiliary letter - geared to the agency.)

Next, what previous work has been done by others on this problem, along with criticisms of it.

Next, Give summary of main deficiencies of previous approaches. ~~XXXXXXXXXX~~

Give outline of progress in induction theory of last ~10 yrs. ^{Eurus, Winston}
Give the main points that are relevant to overcoming the difficulties of the "previous approaches".

(In describing the work of others, ~~do not~~ do not mention deficiencies in it unless they are extremely clearly there. Also, when mentioning deficiencies of an approach, mention the essential progress in A.I. that was obtained via that approach.

These ^{own} remarks are made w. understanding that the people whose work I refer to will probably be reviewing this proposal!)

01: 84.40 : Abstract: (use title of 80.01) Sec 107.01-21 for revision most successful

The proposed investigations will ~~continue the previous work~~ ^{apply the recent results} in induction and generalize it ^{approaches to} induction theory to ~~use~~ the most successful / inductive ~~programs~~ ^{available} at the present time. ~~What~~ It should be possible for programs to ~~learn~~ to arrive at their present proficiency by a suitable training sequence given to an initially less ~~able~~ ^{able} induction program, ~~and~~ by continued training, ~~eventually~~ ^{it would go} on to greater proficiency in more general problem universes.

A parallel investigation will devise a training sequence starting from ^{the} most elementary problems and continuing on up to the most difficult ones. A machine will be devised that is able to follow along in learning to work this ~~and~~ training sequence.

Along with these two ~~programs~~ program developments, ~~the~~ the theoretical analysis of induction will continue, with emphasis on its application to practical problems. y y

23

The "Introduction" starts with 80.01 - 83.40. - perhaps use that as a long "abstract".

29

A "true" introduction: More like an extended table of contents.

→ Description of just what the inductive inference problem is — give a formalization & some examples of problems that can be ~~into~~ expressed in that form. Mention a few areas in which a machine that could do this would be of much value (e.g. weather prediction in essentially different way from usual methods, — Mod diagnosis - decision theory).

01:105,40:

Previous work in this field: Uttlay, Rosenblatt, Kachan, Solomonoff, Fogel, KGS, IE3 (summary 1964 paper for list), Evans, Winston.

Discuss difficulties, inadequacies of this work. (perhaps refer to [Note lack of work on probability - explain why its impl. - origin decision theory

How recent induction research will overcome these diffs & inadequacies. Mention specific inadequacies that are common to several of past attempts. Show very specifically - say in 1 or 2 particular cases, how induction research will solve these problems.

to operator or not.
 Shall we treat it as an enemy attack?
 probability of snow.

A more detailed description of new approaches that will be pursued: (1) give example of application to Winston's work (2) Give example of ~~the~~ simple training in Arith, alg, calca, analysis. - That all of "prog" will be done by giving examples - & examples will be gen. w/ less & less cases as we continue. That the ^{sequence} ~~prog.~~ will be very similar to that given to a child.

Discuss Practical problems that need be solved - give specific problems & expected solns &/o approaches

Discuss Hardware Methods of use of hair memory, of many ALU's in ll of cache memory, of Program on  mean flip time vis. min. poss. flip time. Perhaps insert a ~~new~~ little math. on hair memory utilization.

explain why its imp. - that TM feasibility depends on hardware.
 That speed $\approx 2 \times 10^5$.

short Abstract: (use title of 80.01)

from 105.01-23

TM07

01; 106.40: ↓ ↓ The proposed investigations will ~~utilize~~ ~~utilize~~ recent results in inductive inference theory to ~~the~~ ^{best} overcome the limitations of the inductive inference programs that ~~are~~ ~~are~~ ~~are~~ presently available. These ~~programs~~ programs will be revised so that by presenting to them ~~a~~ suitable training sequences of usually correct examples, their internal structures will be modified so that they will be able to deal with more difficult problems or problems in different fields.

A parallel investigation will devise a machine and an associated training sequence that will start from very elementary problems and work up to more difficult ones. Such a machine is expected to be more versatile than the machine previously described, but it will require a much longer training sequence.

Along with these program developments, the theoretical analysis of induction will continue, with emphasis on its application to practical problems,

See 80.01-83.40 for a longer ~~an~~ introduction ~~to~~ abstract.

Extended introduction $\hat{=}$ Giving contents & summary of f. following sections. (See 105.29-106.40 for preliminary ~~the~~ approach to introduction)

This will be a brief statement of what each section is about.

107.40

Introduct:

① ~~The nature of the induction problem~~ The nature of the induction problem + 1408
and ~~why it is important to get a mechanized solution.~~ ^{The value of a} ^{good} mechanized solution.

Weather prediction, medical diagnosis, discovery of patterns in data, ~~and~~ extrapolation of economic ~~data~~ ^{traces}, and ~~research~~ ^{all} of human learning behavior ~~are~~ ^{are} examples of inductive inference.

Mention that all learning processes are examples of induction.

All of these problems can be usefully expressed in the form of extrapolation of a sequence of symbols.

Ordinarily ~~the~~ induction is performed by humans, sometimes with aided ~~the~~ ^{aid} by computers.

There are several reasons why we want to mechanize this procedure. ~~unfortunately~~ A major goal is the purely scientific one of understanding the induction process in humans. Such a understanding would be invaluable in ~~and~~ teaching or otherwise enhancing the learning ~~and~~ ^{creative} process in humans. Our models of human learning of complex tasks are ~~is~~ extremely rudimentary. ~~An understanding of a working~~ induction machine would ~~be~~ ^{be} a very good model ~~to~~ ^{to} ~~begin~~ ^{begin} preliminary ~~be~~ ^{be} developed ~~into~~ ^{into} a model for human learning.

In weather prediction, we have only recently begun to get ^{completely} mechanized predictions that are better than human predictions. Humans seem to ~~do~~ ^{make} these predictions significantly differently from the ways now used by machines. ~~Moreover~~ and at present, machine ^{weather} prediction is severely limited by the size and speed of ~~a~~ available computers. One aspect of the proposed induction research will be in the direction of combining the methods used by humans ~~with~~ ^{with} those presently used by machines. Another aspect of the proposed work is ~~an~~ ^{the optimization} investigation of induction ~~of~~ hardware efficiency — how can we best ~~design~~ ^{design} hardware or utilize existing hardware, so as to get the best ^{possible} prediction for ~~the~~ ^{whatever fixed we} amount (spent on computing).

In the case of medical diagnosis, relatively little has been done by machine. ^{perhaps mention Gerry's ideas in his Bibliog.} ~~The advantages of mechanization~~ 109.01

1109.10 / facilities to review its past behavior, ~~to~~ successes & failures & decide why it had succeeded or failed & modify future behavior accordingly

T.M.112

so 1) A report of 105.29 as a contents:

1) What is the problem of Inductive Inference and what is the value of a mechanized solution? 105.29;
108.01-
109.15

2) Previous work on Mechanized induction, ~~and its inadequacies~~ ^{unsolved} ~~the~~ ~~unsolved~~ ~~problems~~ ~~of~~ ~~this~~ ~~work~~ — ~~what~~ ~~needs~~ ~~to~~ ~~be~~ ~~done~~, ~~unsolved~~ ~~problems~~ — Problems that remain to be solved.

Partners
July 1
Section
16

3) A The approach to mechanized induction suggested by recent theoretical research. How ~~the~~ ~~inadequacies~~ ~~of~~ ~~previous~~ ~~work~~ can be dealt with.

4) ~~A~~ A description of the proposed work ~~and how~~ ~~it~~ ~~will~~ ~~deal~~ ~~with~~ ~~the~~ ~~difficulties~~ ~~encountered~~ ~~in~~ ~~past~~ ~~induction~~ ~~research~~ ~~→~~ 114.10
~~inadequacies~~ ~~of~~ ~~difficulties~~ ~~encountered~~ ~~in~~ ~~previous~~ ~~work~~ ~~will~~ ~~be~~ ~~dealt~~ ~~with~~ ~~by~~ ~~the~~ ~~proposed~~ ~~work~~ ~~→~~ ~~114.10~~
it will deal with the ~~difficulties~~ ~~encountered~~ ~~in~~ ~~past~~ ~~induction~~ ~~research~~.

5) ~~Unsolved problems that need to be~~

Problems in the optimization of hardware for inductive inference. The concept of "information processing capacity" of an element or complex of elements.

~~Some~~ ~~the~~ ~~methods~~ ~~of~~ ~~highly~~ ~~efficient~~ ~~usage~~ ~~of~~ ~~an~~ ~~optimum~~ ~~methods~~ ~~to~~ ~~utilize~~ ~~memories~~ ~~of~~ ~~different~~ ~~speeds~~, ~~the~~ ~~optimum~~ ~~Use~~ ~~of~~ ~~many~~ ~~small~~ ~~quick~~ ~~central~~ ~~processing~~ ~~units~~ ~~with~~ ~~the~~ ~~sharing~~ ~~of~~ ~~most~~ ~~of~~ ~~main~~ ~~memory~~. How these structures differ from other large parallel processors (e.g. Illiac IV).

6) ~~Some~~ important unsolved problems in induction theory, optimum utilization of a set of memory banks, of different access times, optimum utilization of a set of fast cpu's with large common memory banks.

6) Some important unsolved problems and unexplored areas in induction theory -

113.15
IPC
114.01

01:12.40: 2) Previous work on Mechanized Induction. Problems that remain to be solved.

At this point, I should familiarize myself w. T. Evans & P. Winston's stuff!

Reading my summary of T. Evans in my 66IE3 paper: P1690!
I think his main problem was to devise a lang. to express desc. each object. ~~When~~ These desc. should be \rightarrow it is easy to ~~to~~ devise good "relations", so $A:B = C:X$ can be solved.

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So the problem is devising a lang., g.c. to set of pictures.

10:11.18 4) The general approach: ~~Start with~~ ~~is~~ a model of human induction ~~where~~ either obtained from (introspection), ~~or~~ from protocol analysis. ^{or} Such models have been studied extensively (Newell, Simon, T. Evans to start, Winston, etc.).

~~Define~~ Define each of the concepts used and ^{assign} ~~give~~ is a symbol. Obtain a "bit" for each symbol, ~~on the basis of~~ This is, loosely speaking, ~~the~~

$-\log_2$ of the ~~prob~~ frequency of use of that ~~sy~~ concept in the past. It will also depend on the ~~constraints~~ constraints that are necessary in using that concept.

Using these symbols and their costs, and various ~~computed~~ punctuation costs, ~~write out~~ write out the

prediction algorithm ~~that has been used in the~~ of the models.

Ordinarily, it will be possible to do this ~~in~~ in more than one way.

from these algorithm descriptions we ^{can} obtain 2 new ~~various~~ kinds of information.

037

First, we are able to get probability values, where, in the original algorithm, we were only able to get single "best" predictions.

Second; After using the algorithm for prediction, the bit costs of the concepts used are changed somewhat.

The new bit cost values can be used to ~~use~~ ^{trial} tell how to combine the old concepts to obtain new concepts. The new trial concepts will have bitcosts now roughly equal to the sum of the bitcosts of their component concepts. For trial runs of new predictions, we should use new concepts of minimum bit cost.

These will correspond to concepts that are formed of only a few concepts that have been very useful in the past.

While the foregoing method of obtaining new concepts would be fine if we had an arbitrarily large ^{amount of} computing power available, it is not particularly good in practical cases of moderate complexity. The reason is, we have not taken the computation cost of the old and new concepts into account.

~~What do we, we must~~
 This can be done ^{by} several ways. ^{or (? using 3!)} A fairly general way is to have a second machine observe the behavior of the first. It observes the concepts used by the first machine, and it notes the computation cost associated with various combinations of these concepts.

From these observations it makes a model from which it is able to estimate the expected computation cost of any new concepts that are obtained by combining ~~the~~ old ones.

01: 115. to : ^{computation cost} These / estimates need not be very good for ~~us to be able~~ ^{TM116} to be able to ^{discover} fairly ~~good~~ useful concepts ~~for the first machine~~ but it will take longer to find these concepts if the second machines' estimates are poor.

As the ~~more~~ body of data used by the machines increases, the second machine will be able to make better computation cost estimates on the basis of larger sample size — even if its models are poor.

In a ~~machine~~ machine with much experience or a ~~high~~ probability estimates of this kind, it should be possible to use the first machine to devise ~~improved~~ improved models for the second machine's computation cost estimates.

~~When~~ ^{when} ~~costs~~ expected computation cost estimates are available, they are used, in combination with low ~~total~~ total bit costs to select out new trial concepts that are ~~more~~ most likely to be useful, and which will require a reasonable amount of computation to test.

Re: 114.37 : We must have several ^{BW} ~~prob.~~ prob. models in 11 to get probly values. We always have 4 standard Bern. seq. methods, but usually we can do far better than that by devising ~~prob.~~ ^{variously deriv'd} sequs. using various constraints — or ^{subsequences} of ~~the~~ corpus.

We are interested in going beyond previous induction programs in 2 ways (1) Obtaining good values of probly. This is ~~the~~ well beyond the use of event counting to obtain frequencies. It involves construction of the most general ^{possible} stochastic models possl.

01:16:40:

2) ~~Discovering new concepts to be used in induction~~: Previous

induction systems have been very limited in this respect. ~~And~~
~~It has been able to express~~ Some of the older
ones used little more than the optimization of a fixed
set of parameters ~~to obtain better induction~~.

~~Some~~ More complex systems were able to ~~discover~~ ~~discover~~
new ~~concepts~~ ~~that~~ ^{go} beyond parameter adjustment,
but were still severely limited in the range of ~~environments~~
environmental regularities that they ^{could} ~~discover~~ ~~concepts~~
discovered.

There have been a few systems that were not limited
in this way, that could, in theory, discover any regularity
in the environment that was ~~describable~~ in a ~~finite~~
had a finite description. These systems have all,
however, been heuristically inadequate, so they had no
way to ~~discover~~ ~~complex~~ ~~new~~ ~~concepts~~ in any reasonable
length of time. ~~They could, in theory,~~ Discovering ^{important} ~~concepts~~

new concepts was theoretically / but not practically
achievable. → 119.35-137

~~The set of all prediction algorithms can be listed, and, conceptually,
tested on the
data of the past. ^{As a function of the} ~~past~~ ^{both past}
complexity ~~part, as an index,~~ ~~each~~ ~~of these~~ ~~algorithms~~ ~~can~~ ~~be~~ ~~weighted~~
suitably, and the ~~best~~ ~~set~~ ~~can~~ ~~be~~ ~~used~~ ~~to~~ ~~make~~ ~~predictions~~
range as predictions that are as good as possible with the
given data.~~

138 : 119.37 In practice this procedure is not entirely possible.
Many of the prediction algorithms take an enormous amount
of time to evaluate. Many of them never ~~converge~~ converge at all

and doubt it is impossible to discover ~~the~~ ^{find} a finite amount of time.

Instead of testing all possible algorithms, it is necessary to restrict the set tested to those that are likely to converge in a reasonable time.

If an estimate is obtainable for the convergence time of any prediction algorithm, it is possible to decide ~~which~~ ^{an} optimum order in which the ~~diff~~ ^{can be computed} algorithms should be tested, optimum in the sense of best expected prediction accuracy for a given amount of time spent testing algorithms. This is, in essence,

~~the induction system can be processed~~ At any particular point the testing program ~~can stop and~~ ^{can stop and} stop ~~and~~ ^{stop} the best ~~prediction~~ ^{prediction} algorithms ~~made~~ ^{can} found up to that time ~~can~~ ^{can} be used for prediction.

This is, in essence, the induction system that

we ~~propose~~ ^{propose}

- ① Direct ponder.
 - ② For " " "
 - ③ Mt. Carlo ponder.
- 120.01

The ~~measures~~ ^{measures} criteria method of generation of ~~prediction~~ ^{prediction} algorithms is fairly well understood. It is also possible to list them in a priori likelihood of effectiveness.

The part of the system that needs most work is the ~~measures~~ ^{measures} calculation of expected testing times for the algorithms. For many large classes of algorithms (Give a list of such), ~~method~~ ^{method} it is fairly easy to obtain such estimates, but for the ~~new~~ ^{new} truly original algorithms that will be generated, there is no general estimation method. Various ~~the~~ ^{the} approximation methods will have to be devised and analysed.

It should be noted that ~~the~~

~~estimates method that is not very good, the system~~

335P →
335P

the effectiveness of the whole induction system, is not critically dependant upon the accuracy of the subsystem that estimates algorithm convergence / testing times. Poor estimates will mean that the machine wastes ~~substantial~~ ~~amount~~ time testing algorithms that are ~~theoretically~~ impractical / or impossible, ~~some of which do not start converging. This will slow down the discovery of algorithms that are essential~~ new, but will certainly ~~not~~ permit the discover

Because of ^{this inefficiency} such a system would be able to discover essentially new algorithms, but at a much slower rate. ~~that~~

~~A system with a better testing time estimator.~~

Eventually, if the entire system has worked ^{up to} ~~the~~ ~~point~~ ~~of~~ ~~being~~ ~~a~~ ~~capable~~ ~~of~~ ~~dealing~~ ~~with~~ ~~fairly~~ ~~complex~~ induction problems, it will be able to work on the problem of improving its own algorithm testing time estimator - ~~the~~ ~~same~~ ~~normal~~ ~~inductive~~ ~~inference~~ ~~program~~

At first, the system will use a fixed ~~algorithm~~ ^{program} for estimating testing times of algorithms. This program will be improved from time to time by its (human) operator. ~~or a user~~ ~~will~~

35

The set of production algorithms can be listed and, conceptually at least, each can be tested on ^{known} the / data ~~set~~. Each of these algorithms can be weighted as a function of its informational complexity and ~~the~~ the goodness of its performance on known data. This weighted sum will give predictions that are as good as possible ~~all~~ with respect to the ~~data~~ ~~known~~ available data, → 117.38

37

ol: 119.40:

SN :
118.20 spac

On 1. Method of generating new Pans.

TM120

