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The Effects of Extended Practice
on Performance in a
Tracking Task

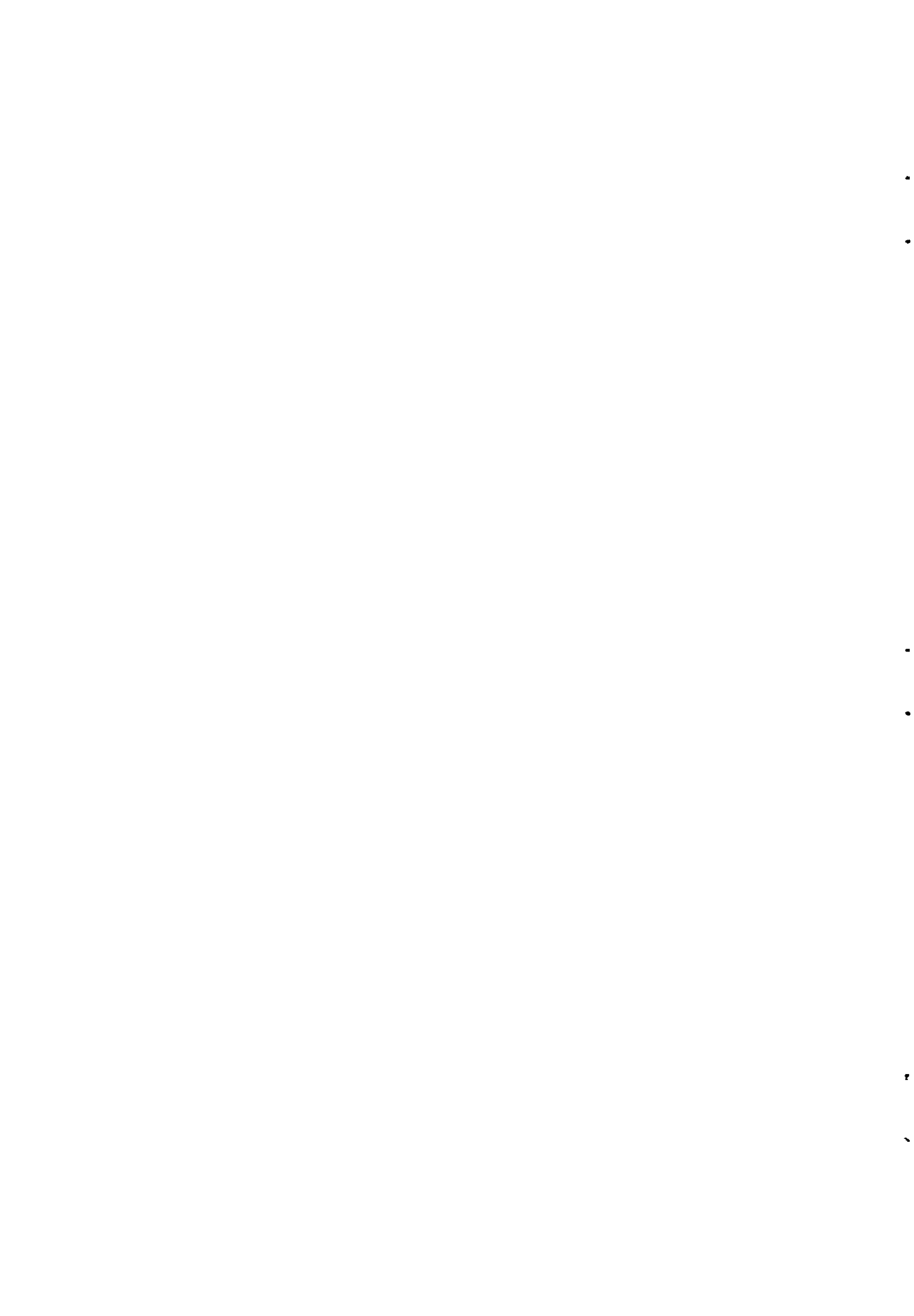
By

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The Effects of Extended Practice on
Performance in a Tracking Task

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SUMMARY

Experimental measurements of human controller performance have been made during extended periods of practice in **visual** sine-wave tracking tasks. It has been found that, irrespective of task difficulty, **RMS error scores** decreased to such **small** magnitudes that differences in scores due to different task variables would have no practical significance. Thus the averaged values of steady scores when tasks are **well** learnt are meaningless for **subject** or task difficulty comparisons. It has been shown that the performance scores vary in **an** exponential manner **with** the number of task **repetitions** and it is proposed that **an** empirical constant related to the rate of decrease of scores be used as a measure of relative task difficulty.

Introduction

Performance scoring is a method of measuring the proficiency of a human pilot in a manual control task. Often this kind of **experimental** approach is used to compare the relative skills of different controllers, or the **relative** merits of different control systems. Experience has led to the notion that there exist "plateaus of performance" for both **subject** variations and task variations. Thus, if different subjects were to practice a particular control **task** for a period of time, it is assumed that the **performance** scores would eventually reach steady **values** as **shown** in Fig. 1.

During the first N_1 repetitions the scores of both subjects are **continually** decreasing. During the repetitions **numbered** N_1 to N_2 the scores of the subjects A **and** B are the **constant** values S_A and S_B . During the repetitions **numbered** N_2 to N_3 the respective scores are unchanged.

In order to **draw significant** conclusions from such a set of data, two statistical tests are **required**. The first test is required to show that the mean values of the **scores** in the interval N_2 to N_3 do not differ significantly **from** the mean values **in** the interval N_1 to N_2 . (In **practice**, measured **human performance scores** would not decrease in monotonic order as shown in Fig. 1 because of the random variability **of human** controller behaviour. Neither would the scores in the steady interval N_1 to N_3 be exactly constant, hence the need for statistical tests.)

A/

A second test is required to determine if there is any significant difference between S_A and S_B when the first test shows that no further decrease of averaged score is taking place. If the difference between S_A and S_B were found to be significant then it would be concluded that subject B was a more skilful controller than subject A.

In a similar manner, two different control tasks could be compared on the basis of the performance scores of a particular subject or group of subjects. If a significant difference between the mean steady scores in the two tasks were found it would be concluded that one task was more difficult than the other.

This procedure has generally proved useful as a rough and ready guide for the selection of personnel for machine operation and for the design of dynamic systems. Nevertheless it has shortcomings and there is still a need for improved methods which take account of the effects and potential effects of operator learning on control performance. An increasing amount of doubt has been expressed as to whether plateau levels of performance actually occur, or whether scores would continue to decrease on the average if lengthy practice were continued.

Moreover, there is a class of problems in which the performance scores, e.g., error scores in a tracking task, can become so small that they are difficult to measure accurately. Moreover, provided the measuring apparatus is designed to resolve the smallest difference which has any practical relevance, nothing is to be gained by introducing more sophisticated and more accurate measurements. Any difference detected in the mean scores which might be significant in the statistical sense could be trivial in relation to the task at hand.

The experiments reported in the present paper have illustrated these kinds of phenomena in sine wave tracking tasks. Extended periods of practice have shown that apparently steady plateau levels of performance are of relatively short-term duration and that the error scores continued to decrease on the average as practice progressed. Eventually the scores reached the limitations of the measuring apparatus. Had a more accurate measuring procedure been adopted then any differences which might have been detected among the independent variables would have been trivial.

Thus the state of affairs was that two subjects performed six tasks of different degrees of difficulty, which included twelve independent variables. The final steady mean performance scores were the same in all cases after sufficient periods of practice. It is clear that comparisons among the independent variables cannot be based on final steady performance scores. It is shown, nevertheless, that a significant comparison can be based on the average rate of change of scores with the number of task repetitions.

Experimental Description

The subjects sat in a small enclosed cockpit watching the motion of a spot of light on a 5 in. diameter oscilloscope at a distance of about 24 in. Control was effected with a 20 in joy-stick having low friction and linear spring force gradients of about 1 lb/in. An electrical output from the stick was fed to an analogue computer and control dynamics relating spot movement to stick movement were:

- (i) position control, $x = 2.0$
- (ii) velocity control, $\frac{dx}{dt} = 4.0$
- (iii) acceleration control, $\frac{d^2x}{dt^2} = 4.0$

where/

where x = spot displacement in cm from a reference point

= stick deflection in inches, measured from centre at the hand grip.

Performance scores were measured in both single **axis** and **two axes tasks**. In the single **axis** tasks a sinusoidal input of 0.2 cps nominal frequency was fed to the horizontal plates of the oscilloscope. If this motion were uncontrolled the **maximum amplitudes** of the spot from the reference would have been ± 2.5 cm. The subjects attempted to track the **signal** by controlling the spot to keep it as close as possible to the reference,

In the **two axes** tasks an additional sine wave input was applied to the **vertical** plates of the oscilloscope. This input had the **same** amplitude and frequency as the horizontal input but was displaced in phase by 90° . The uncontrolled motion of the spot was thus a circular path of 2.5 cm **radius**.

Individual tracking thus occupied one minute. During the first 15 **sec.** no scoring occurred. In the next 30 **sec.** the tracking error voltage was squared and integrated in the computer, There was then a rest period of 15 **secs.** Two **subjects** alternately performed sessions of ten one-minute **runs**. At the end of a run the **final** voltage corresponding to the total integrated squared error was recorded on a trace recorder. The traces were later read to the nearest whole number percentage of **RMS error/RMS** input. A **100%** score would, therefore, be obtained with **an RMS error** of 1.8 **cm** on the oscilloscope, a **10%** score with 0.18 **cm** and **5%** with 0.09 **cm**, etc,

The main object of the tests was to observe the effects of extended practice on human operator performance in tracking tasks of known difficulty. A sine wave input was chosen for two reasons. **First**, its simple repetitive form could be learned quickly by the subjects, thus helping to reduce the large number of laborious task repetitions which would be required in any case. Secondly, by using this **same** input throughout it could be assumed that the difficulty of the **tasks** depended only on the nature of the control dynamics and the number of axes being controlled. Thus we could, with reasonable certainty, estimate the relative difficulties of the tasks in qualitative terms which could then be compared **with** the quantitative results. It is known that velocity tracking is a little more difficult than position tracking, and acceleration tracking considerably more difficult than either, Corresponding two **axes** tasks are more than twice as difficult as single axis tasks, the controller has to exert twice the effort to control about both axes simultaneously, and as well as this he has to **make** additional decisions as to which control movements are required for **which axis**.

In **all** cases, task repetitions were continued until scores of **5%** were achieved regularly (i.e., at least five times in a session of ten repetitions). A **5% score** implied a **root-mean-squared** error during the run of less than a **millimetre**, and it was apparent that errors as small as this could easily have been introduced by parallax **in** the display. Hence, as pointed out in the introduction to this report, **any** differences in mean steady scores of this order of magnitude would be classed as "**trivial**" in relation to the nature of the **task**. Moreover, it would not have been possible to measure scores lower than this value with **any** great **accuracy**. For a displayed error of 1 mm the voltage output of the **servo-multipliers** which were used for squaring was about **1/10th** of a volt. After **integrating small** voltages of this order for half a minute the overall voltage is **also** contaminated by effects **such as** amplifier drift and **d.c.** bias, Thus it seemed logical to disregard scores apparently lower than **5%**. These were read as **5%**.

Experimental Results

Variations in performance scores with the number of task repetitions are shown in Figs. 2 to 8. Figs. 2 and 3 are the results obtained with single axis position and velocity controls. The crosses represent the scores in each of the individual runs. Scores were reduced to the 5% level after about 50 to 70 repetitions.

The scores obtained with single axis acceleration dynamics are shown in Fig. 4. The crosses in this figure are the mean scores of every 10 task repetitions. With acceleration controls the subjects required over 500 task repetitions to improve performance to the 5% level.

In Figs. 5, 6, 7 and 8, which illustrate the variations of scores in the two axes situations, the crosses are the mean value of 10 vertical scores and the circles the mean values of 10 horizontal scores. The number of task repetitions required to reach the 5% level was more than double those required for corresponding single axis situations.

The fitting of learning curves

The plots of performance scores as functions of the number of task repetitions can be fitted with exponential curves of the form:

$$\bar{S} = Ae^{-\frac{n}{B}} + C \quad \dots(I)$$

where $\bar{S} = \frac{\text{RMS error}}{\text{RMS input}} \times 10\%$

n = number of task repetitions

A, B, C = empirical constants.

Such curves are shown in Figs. 2 to 8 and the values of the appropriate constants are listed in Table I. These curves were obtained by plotting $\log(\bar{S}-C)$ against N and drawing a straight line through the points by eye. The constants A and B for the curve shown were obtained from the equation of the line. It would, of course, have been possible to obtain the best fitting line more rigorously by submitting the experimental points to a computer programme but this refinement was not considered necessary at this stage of the work. It may be adopted later. Due to the process of rounding off small scores to 5%, asymptotic convergence is ensured and the constant C in all cases is equal to 5. Although it may happen that ten or twenty repetitions will produce a score of 5% there is, nevertheless, still a finite probability that a score greater than 5% could be obtained in subsequent repetitions. Hence the convergence of the mean of the scores is, in fact, asymptotic and not absolute.

The constant A is a measure of the initial proficiency of the subject without practice. In effect the value:

$$\bar{S}_0 = A + C$$

is the expected value of the subject's score for n = 0. In a lengthy series of experimental tests there will inevitably be "transfer of training" effects. If a subject learns a task and the task is made slightly different by a change in

one of **the** independent variables the subject may initially perform a good deal better in the **new** task than he would have done had he not had the previous training. In the present tests both subjects initially performed better in two **axes** tasks than they **did in** single axis situations even though two axes **tasks** are considerably more difficult than single axis tasks, This resulted **from** improvements in the subjects' overall skill due to practice in the single axis tasks which were performed first, It is felt that the value of A has little relevance to the degree of task difficulty.

Consideration of the **values** of the constant B averaged for the **two** subjects (Table II) shows that this number at any rate places the six tasks **in** the order of difficulty which the two operators were agreed upon.

Moreover this number **is** evidently insensitive both to the order of testing end to the subjects controlling ability. It is, therefore, proposed to adopt the constant B as a measure of task difficulty. If this is done, and taking the **value** of B in single axis position control as unity, the relative difficulties of the six tasks are as shown in Table III.

Idealised learning

In a difficult tracking task a **subject's** error scores decrease very **slowly** with **practice**. Thus one of the difficulties in fitting learning curves is the fact that it **might** require an impractically large number of task repetitions to establish the trend of the **curve**. For example, the 1,500 or so task repetitions which were made by the subjects in the **two** axes acceleration control sine tracking task would be considered to be too many for a routine experiment in which **many** independent variables were to be tested.

Considerable saving of effort could be achieved if it were possible to predict a **subject's** ultimate **performance from** his early **performance**. If we assume that error scores in a tracking task will vary **with** task repetitions according to the law given by equation (I), then, by **taking** logarithms we see that:

$$\log(\bar{S}-C) = \log A - \frac{1}{B} n \quad \dots(2)$$

Thus if $\log(\bar{S}-C)$ is plotted against n we obtain a straight line of slope $-\frac{1}{B}$, where B is the index of task difficulty sought for. Hence it is only necessary to perform a few task repetitions to determine this slope with a reasonable degree of accuracy. In Fig. 9 the two axes acceleration scores of subject 1 have been plotted in the logarithmic form. The data points from the first few hundred task repetitions lie reasonably on a straight line. However, **when** this **line** is extrapolated we find that later scores do not lie on the line. The constant B obtained **from** the slope of the line in Fig. 9 has the value **263**, as compared with the **value** of **400** obtained by fitting **an** exponential curve to the whole 1,500 pairs of scores.

The initial and more rapid rate of learning will be referred to as "**idealised** learning". **When** a human operator first attempts to perform a difficult task it is reasonable to assume that he concentrates his whole attention on the task, thus using up all his spare mental capacity. He achieves a certain score, **S**, but he has been controlling very inefficiently by making a lot of imprecise control **movements**. After a short time he makes **fewer** mistakes and can achieve the performance score **S₁** with less than his maximum effort, which leaves him with an **amount** of spare mental **capacity**. He then exerts maximum effort **again** and the spare mental capacity enables him to decrease the error **slightly**.

Idealised/

Idealised learning is **thought** to be learning which occurs when the controller is continually using up **all** his spare mental capacity and always exerting maximum effort as his controlling ability improves. However, a controller learns at the idealised rate only during the initial practice period. **When** he had reduced **the tracking** errors to a level considerably smaller **than** his unpracticed level he *no* longer exerts **maximum** controlling effort. He prefers to continue controlling with increasing amounts of spare mental capacity rather than trying to improve his performance at a greater rate by exerting **maximum** effort. Actual learning curves **will** therefore **depart** further **and** further from idealised learning curves as practice progresses.

Nonetheless, the experimental results suggest **that values** of B for **idealised** learning curves may be in linear proportion to **values** of B for **actual** learning curves. Hence idealised **learning curves** may prove to be equally valid for comparative purposes. Table IV shows that the ratio $\frac{B_{2 \text{ axes}}}{B_{1 \text{ axis}}}$ for the acceleration control scores of subject 1 is approximately the same for both **idealised** learning **and** actual learning.

Conclusions

The following conclusions apply to the tracking of low frequency (~ 0.2 ops) sine waves. **Whether** they are valid for the tracking of more random input **signals** should be determined by further tests.

- (1) The **RMS** values of tracking errors computed during short controlling runs continued to decrease on the average as practice **was** extended.
- (2) In six tasks of **varying** difficulty, ranging from single axis position tracking to **two axes** acceleration tracking, the only foreseeable convergent limit to the scores was that imposed by the display threshold.
- (3) It is not practical to make comparisons **among** the different tasks on the basis of the magnitude of the scores when all improvement has ceased.
- (4) Scores **vary** with **task** repetitions according to **an** exponential law of the form

$$\bar{S} = A e^{\frac{-N}{B}} + C$$

where N = the number of task repetitions

S = expected value of score

A, B = empirical **constants**

C = lowest score of practical significance
(determined by **display** threshold or
limitations of **measuring** apparatus).

- (5) The **constant** A depends upon transfer of training effects and the **subjects' initial** controlling ability. It does not give **any** valid information regarding comparative task difficulty,

- (6) The constant B is independent of transfer of training effects **and the subjects' initial** controlling ability and is a measure of relative task difficulty.
- (7) The concept of **idealised** learning as described in the present report leads to a rapid determination of relative task difficulty which does not require a **large** number of laborious task repetitions.
-

TABLE I/

CONTROL DYNAMICS	SUBJECT 1			SUBJECT 2		
	A	B	C	A	B	C
I-AXIS POSITION	15	11	5	20	20	5
" VELOCITY			5	25		5
" ACCELN	40	150	5	40	130	5
2-AXES POSITION		40			33	5
" VELOCITY	16	83	5	16	56	5
" ACCELN	22	400	5	22	500	5

TABLE I

CONTROL DYNAMICS	B average	CONTROL DYNAMICS	RELATIVE TASK DIFFICULTY
I-AXIS POSITION	15.5	1-AXIS POSITION	1.00
" VELOCITY	22.5	" VELOCITY	1.45
" ACCELN	150	" ACCELN	9.68
2-AXES POSITION	37.5	2-AXES POSITION	2.42
" VELOCITY	69.5	" VELOCITY	4.49
" ACCELN	450	" ACCELN	29.00

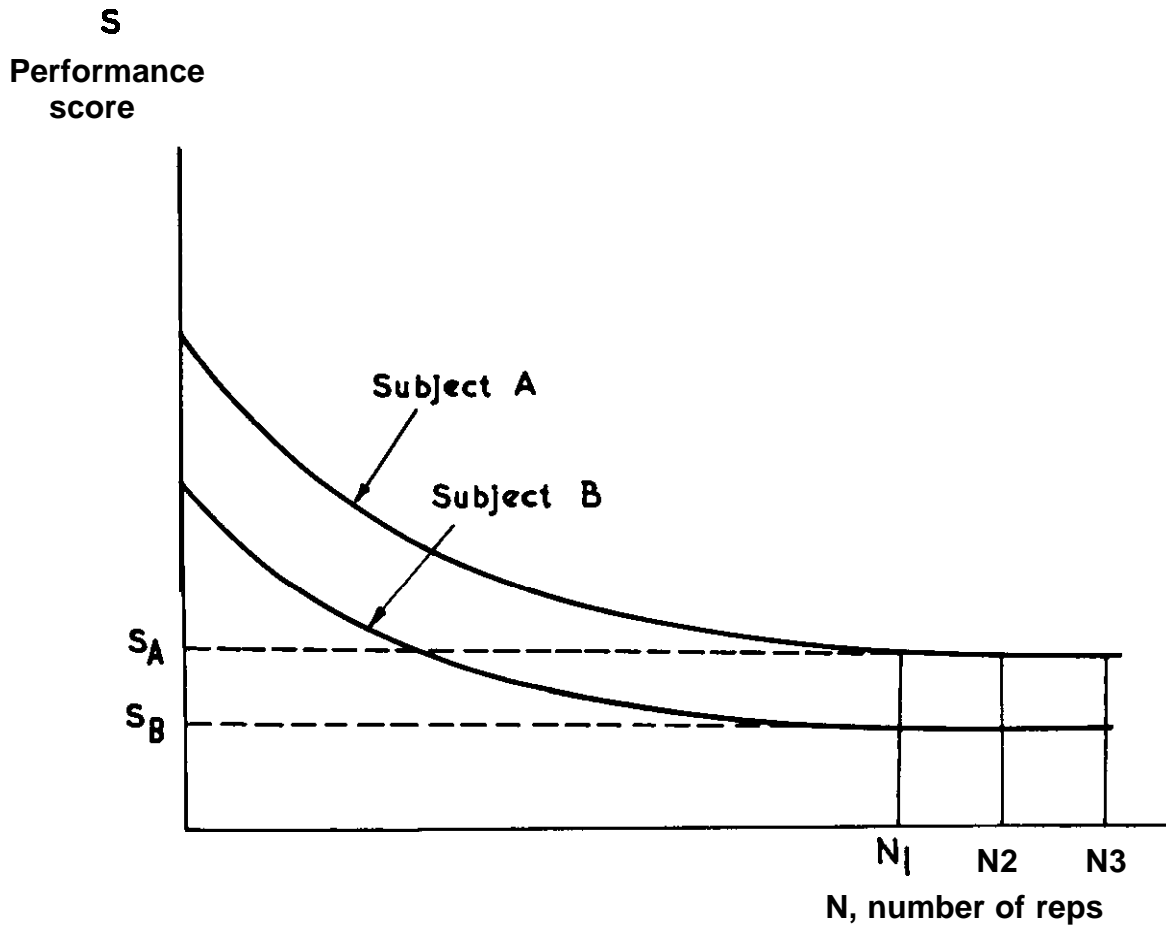
TABLE II

TABLE III

	1-AXIS ACCELN CONTROL. VALUE OF B FOR SUBJECT 1	2-AXES ACCELN CONTROL. VALUE OF B FOR SUBJECT 1	$\frac{B_{2-axes}}{B_{1-axis}}$
ACTUAL LEARNING	150	400	2.7
IDEALISED LEARNING	106	263	2.5

TABLE IV

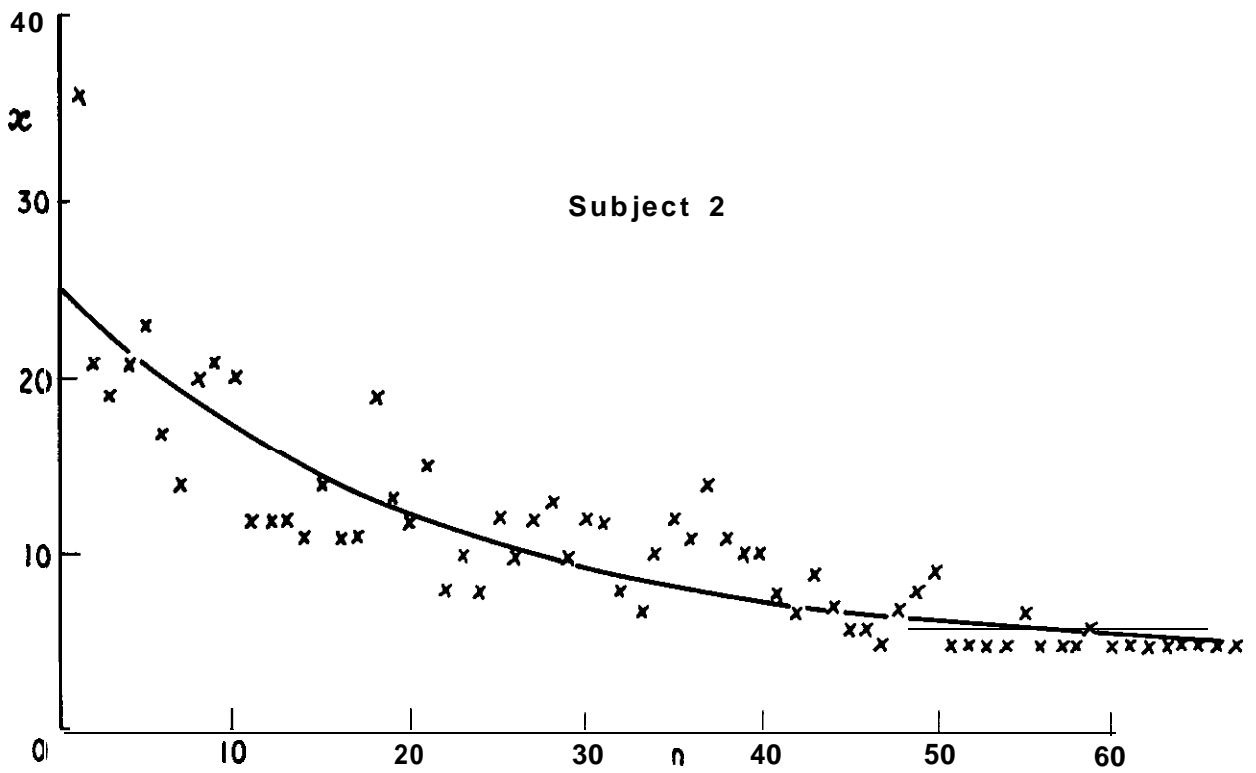
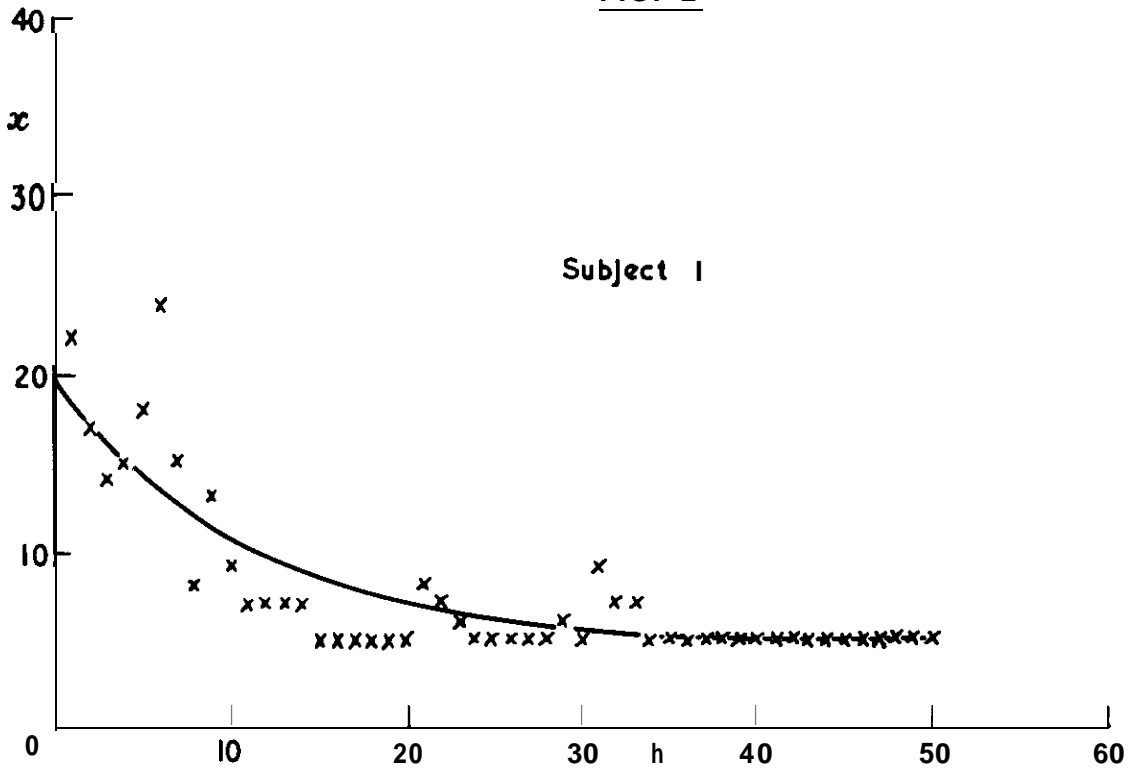
FIG 1



Hypothetical variations of performance scores with task repetition

for two subjects, A and B

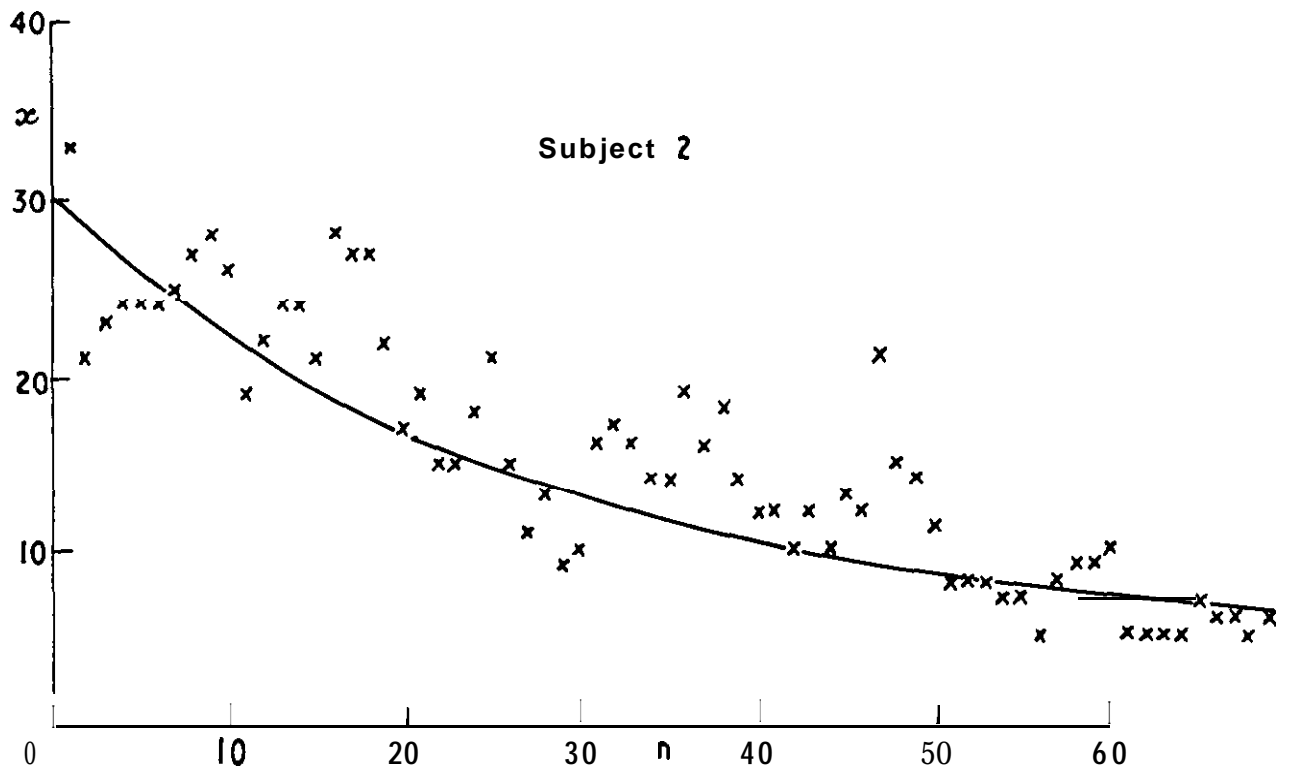
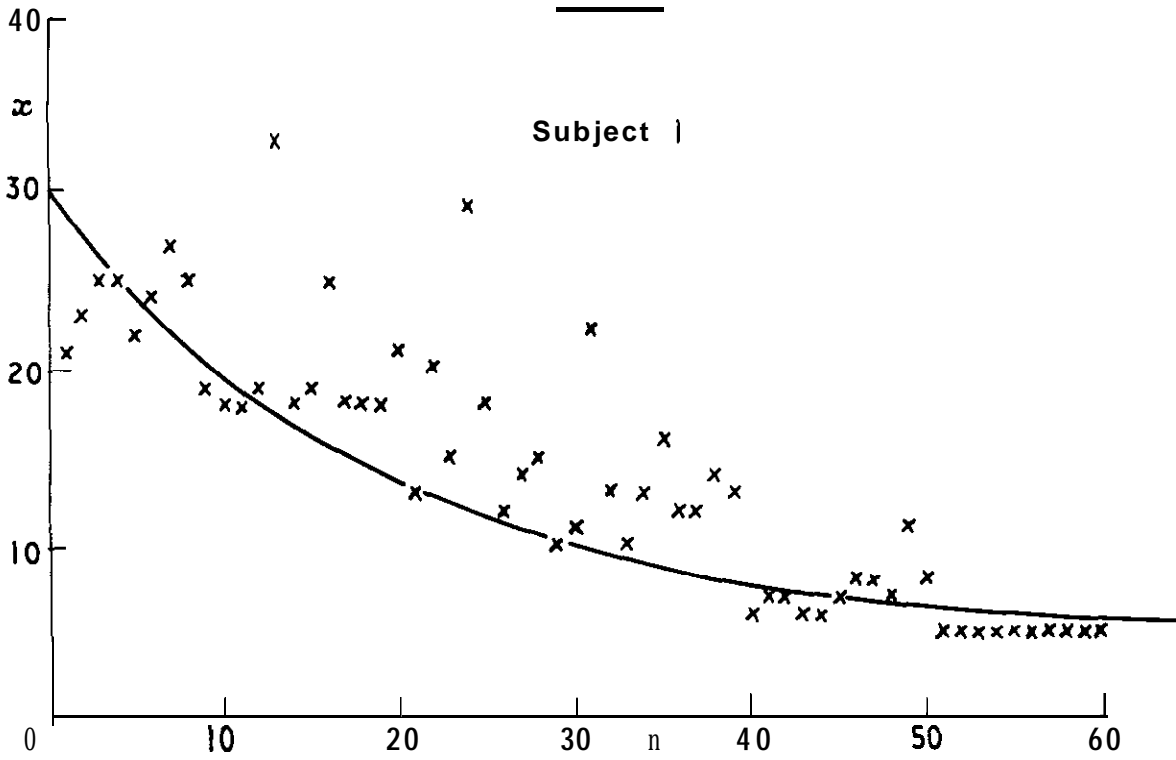
FIG. 2



Learning curves.

One-axis position control

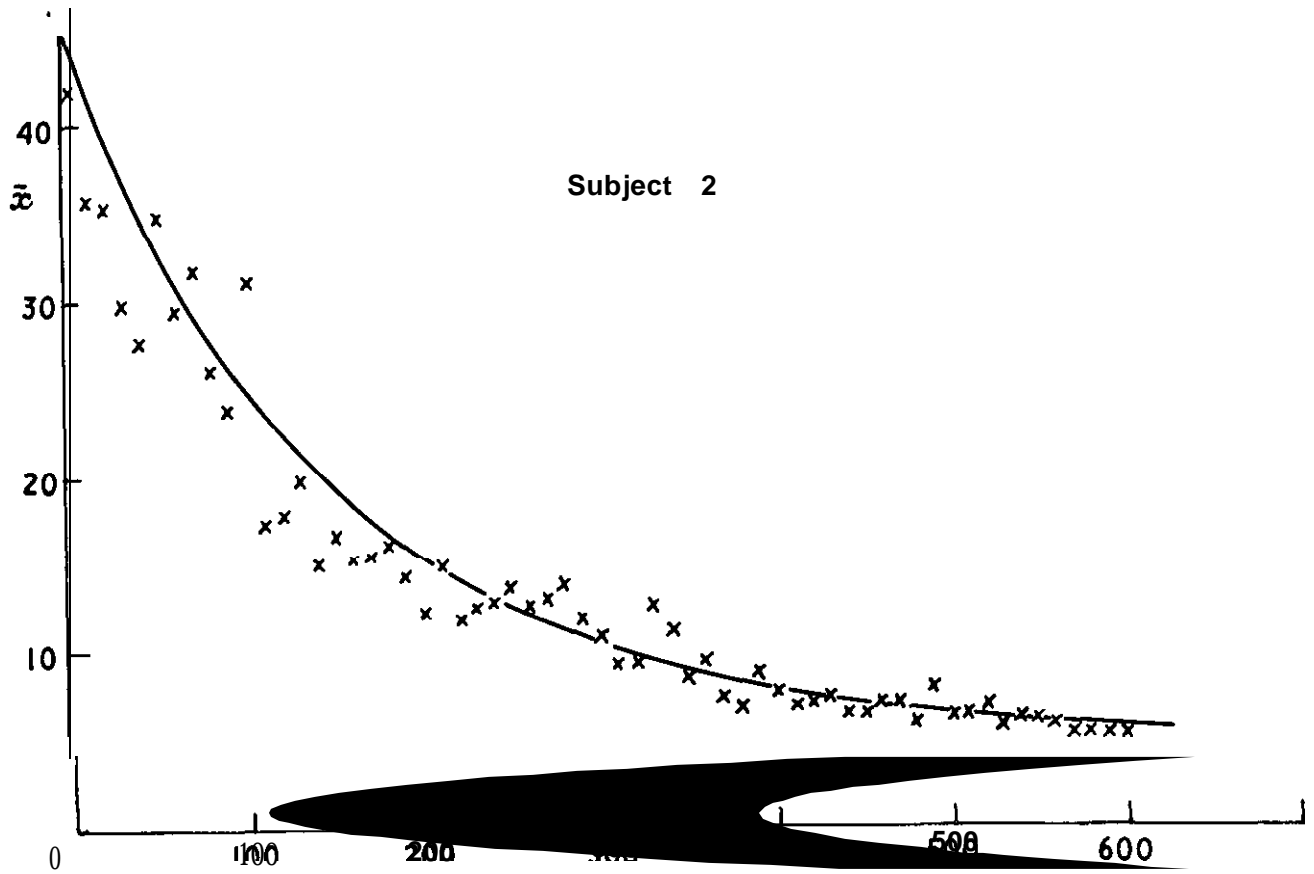
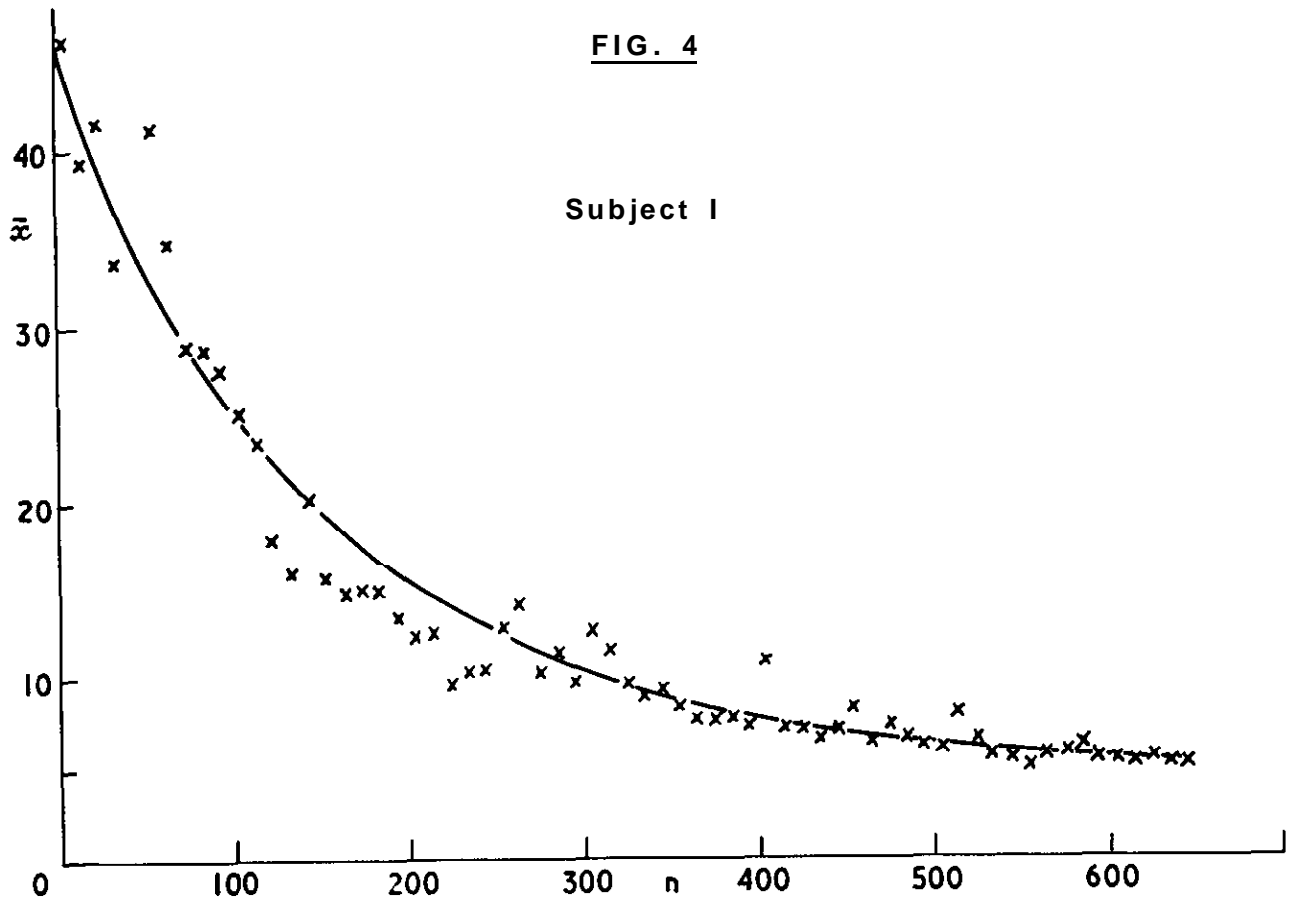
FIG 3



Learning curves

One-axis velocity control

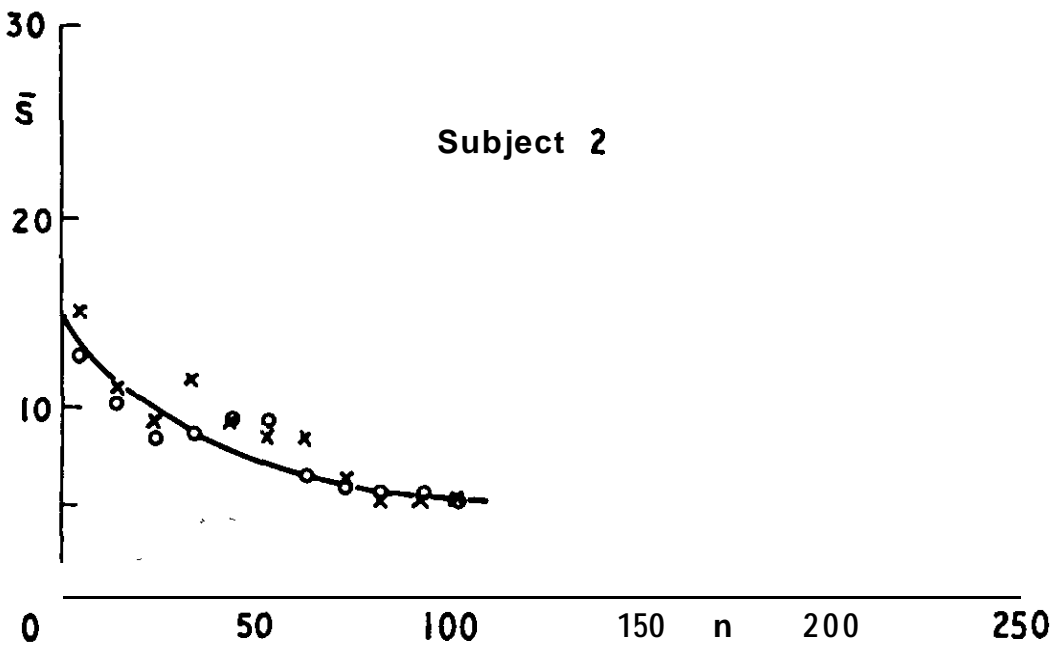
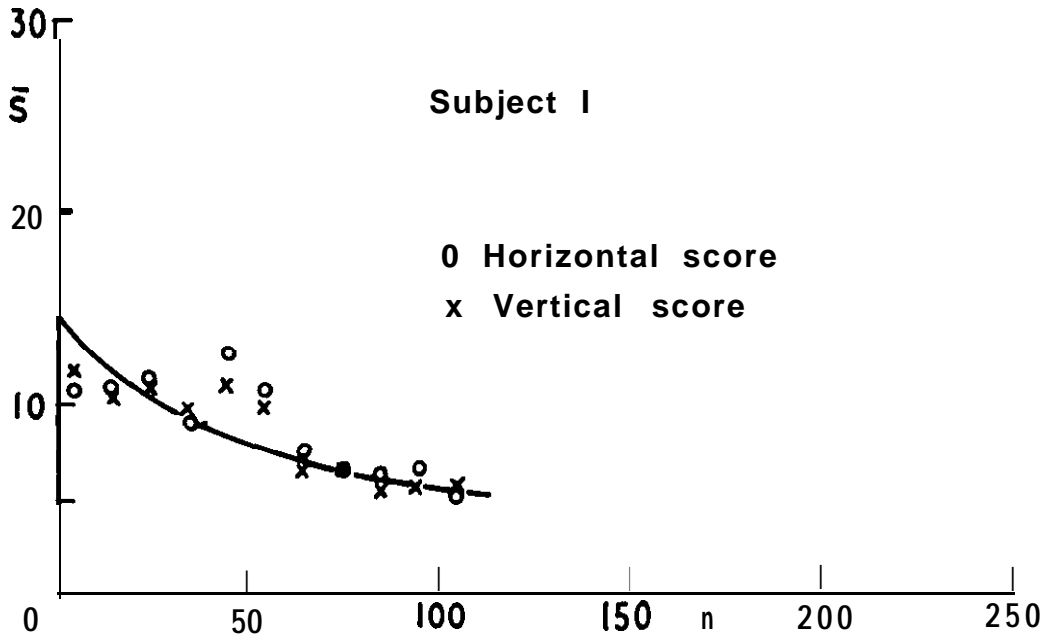
FIG. 4



Learning curves.

One-axis acceleration control

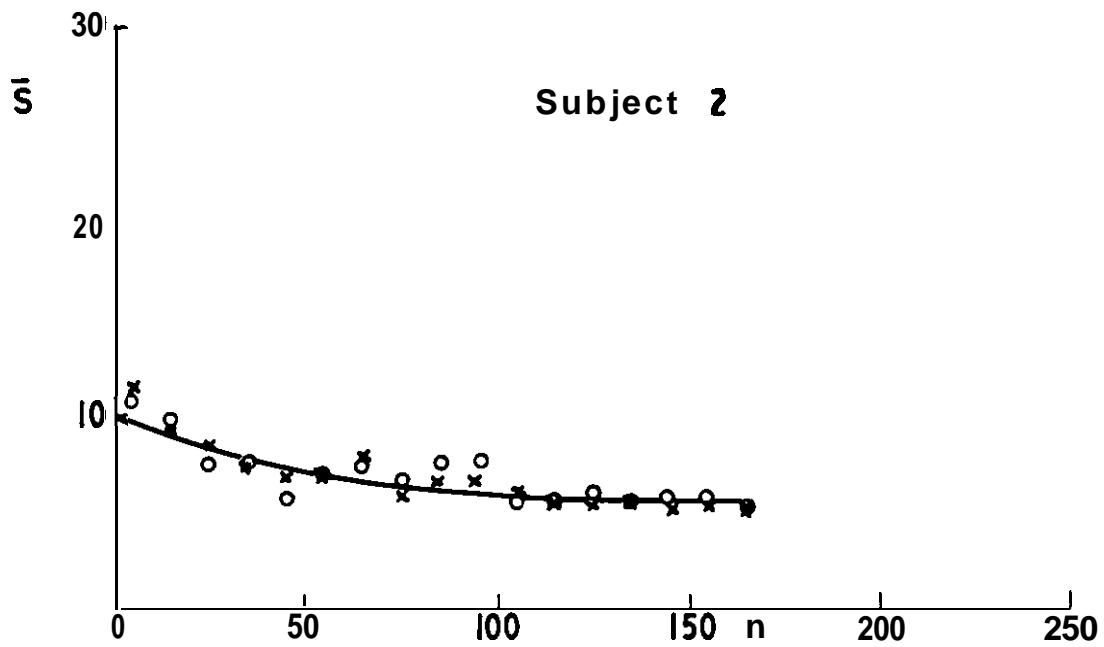
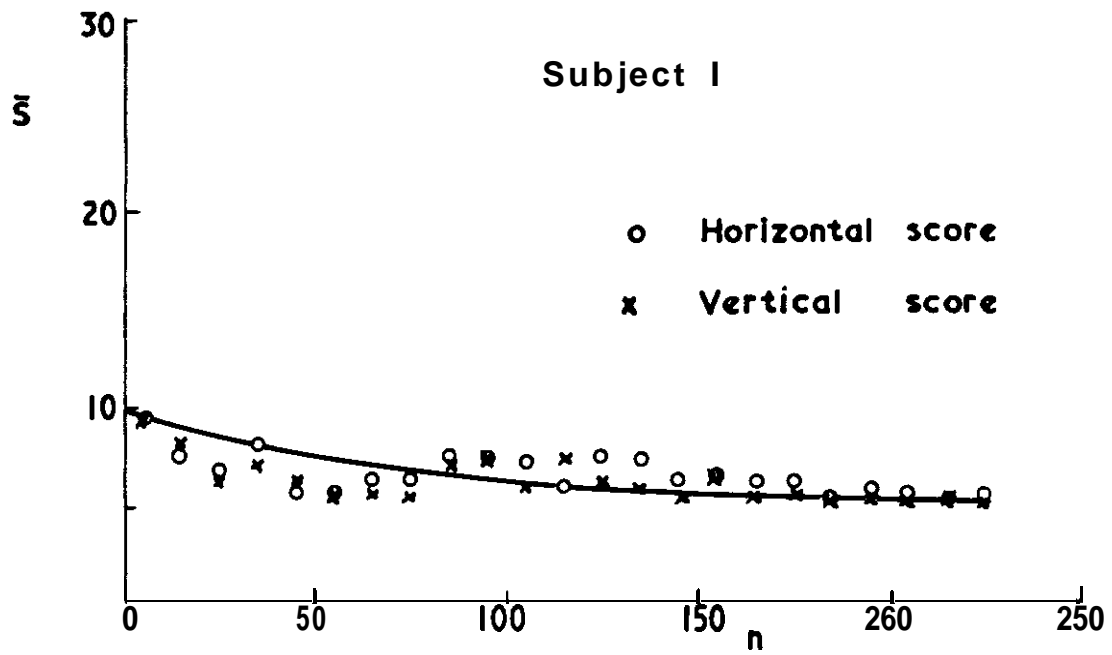
FIG. 5



Learning curves.

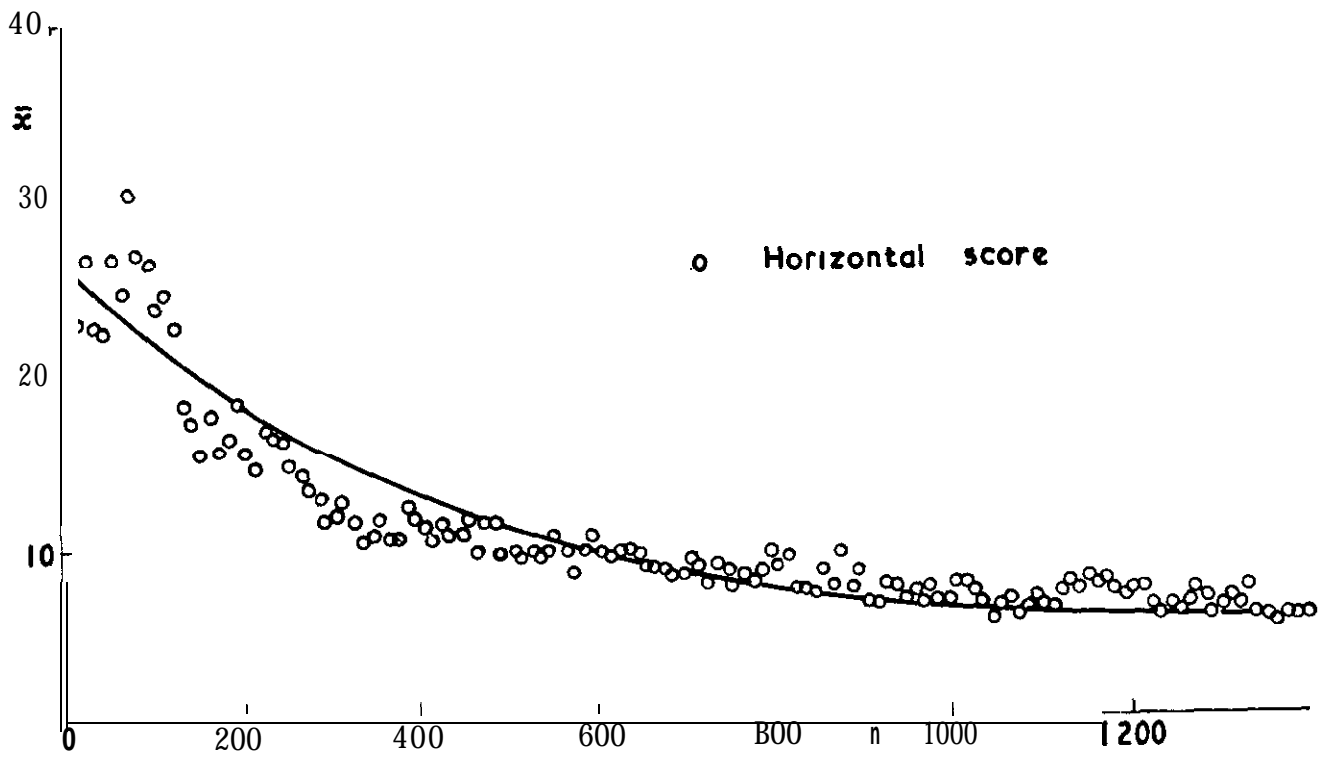
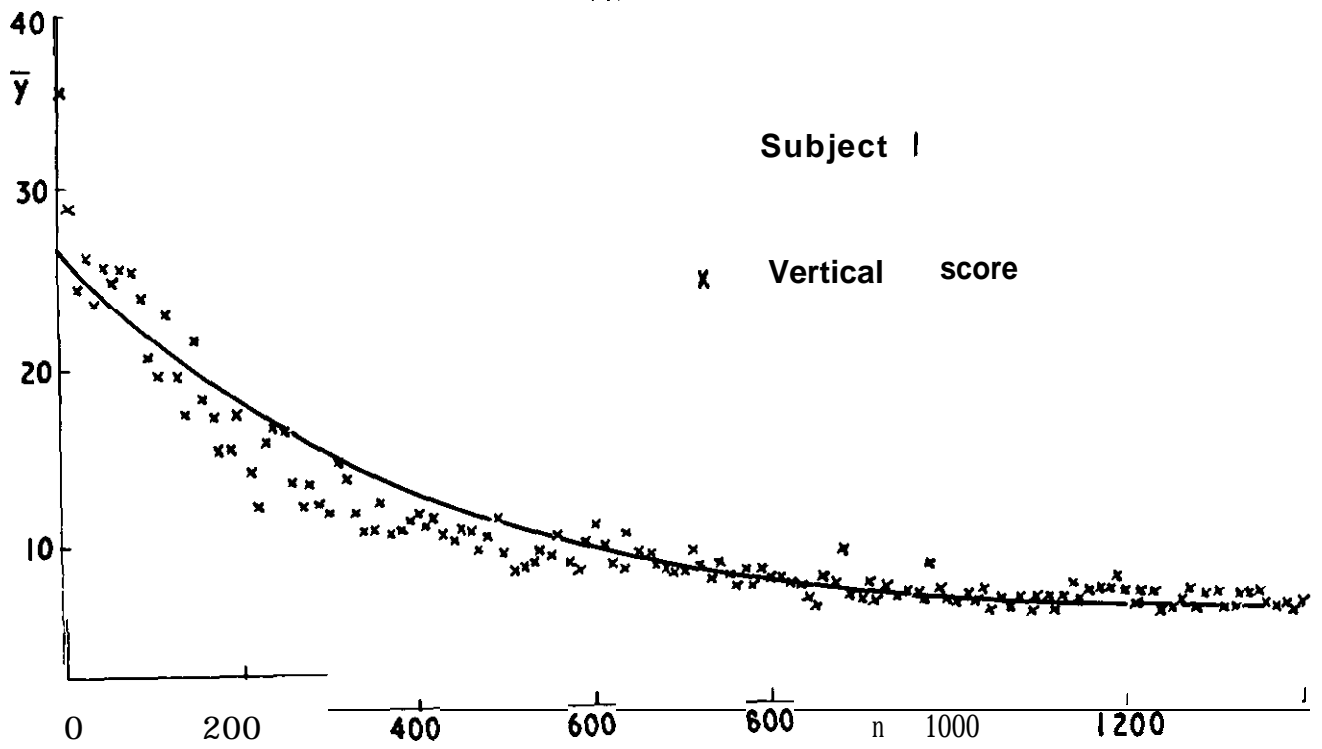
Two -OXES position control.

FIG. 6



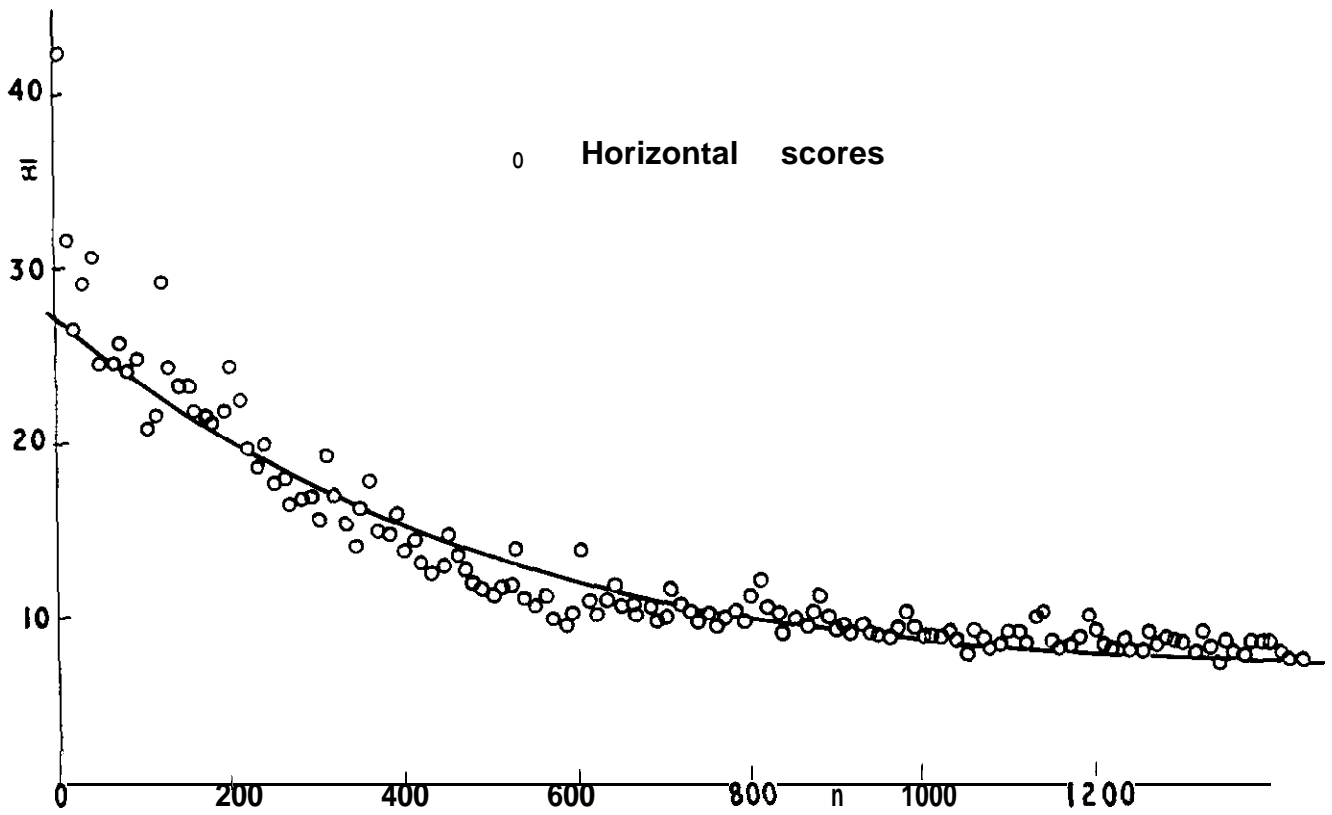
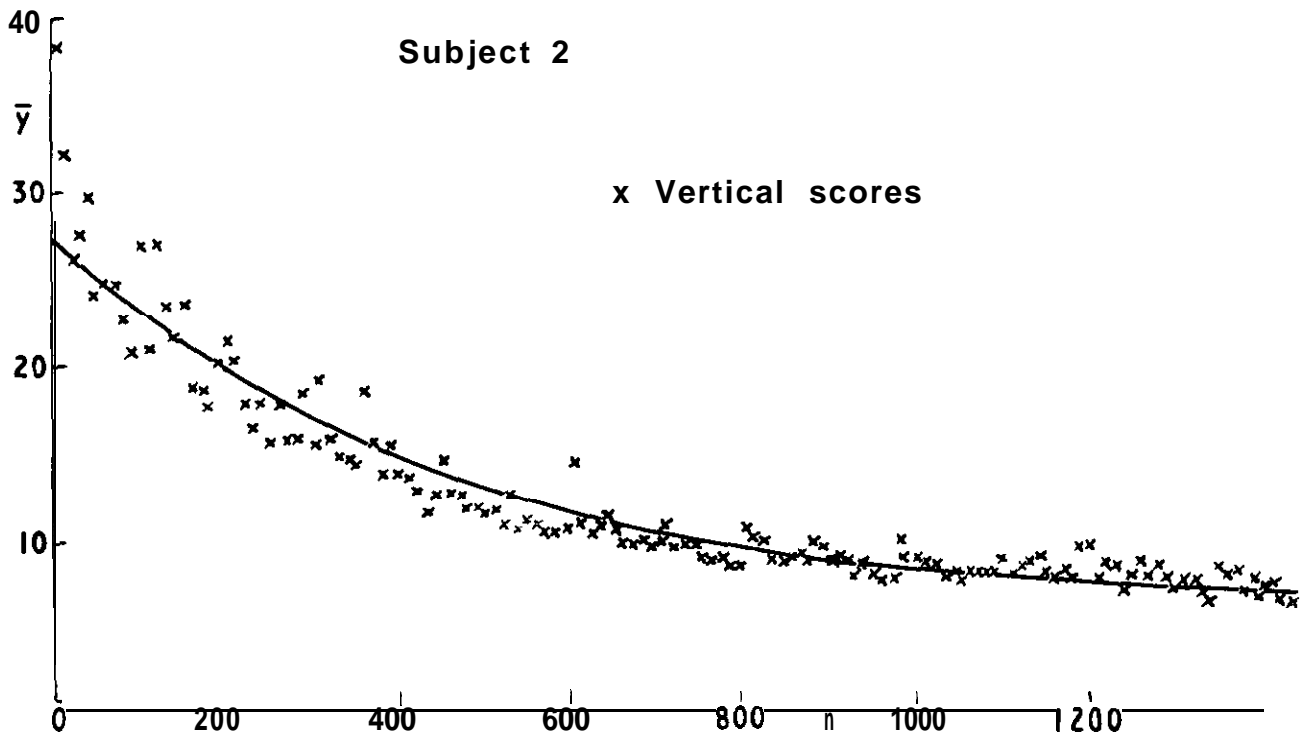
Learning curves. Two -axes velocity control

FIG.7



Learning curves Two- axes acceleration control.

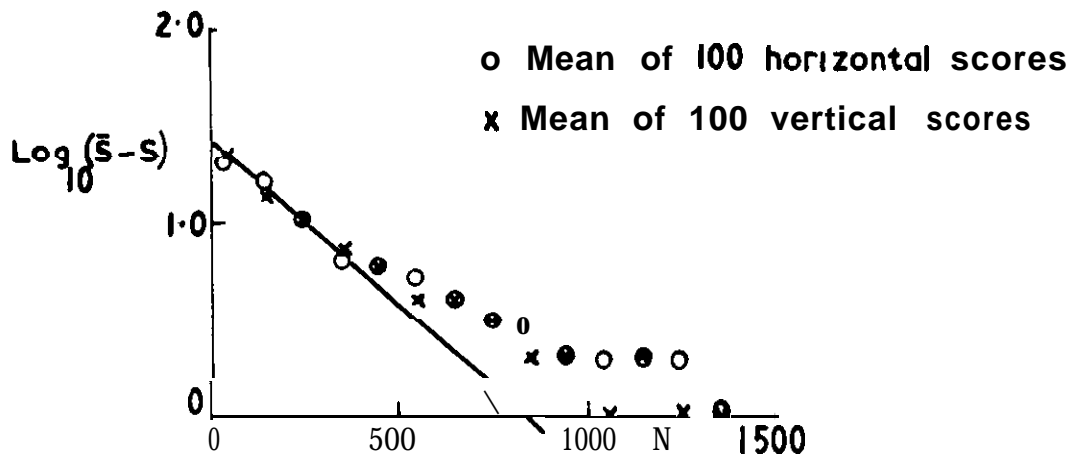
FIG. 8



Learning curves.

Two-axes acceleration control

FIG.9



Idealised learning curve - Subject 1.

2 -axes acceleration control

A.R.C. C.P.1030

December, 1967

R. C. Hornby and R. Wilson.

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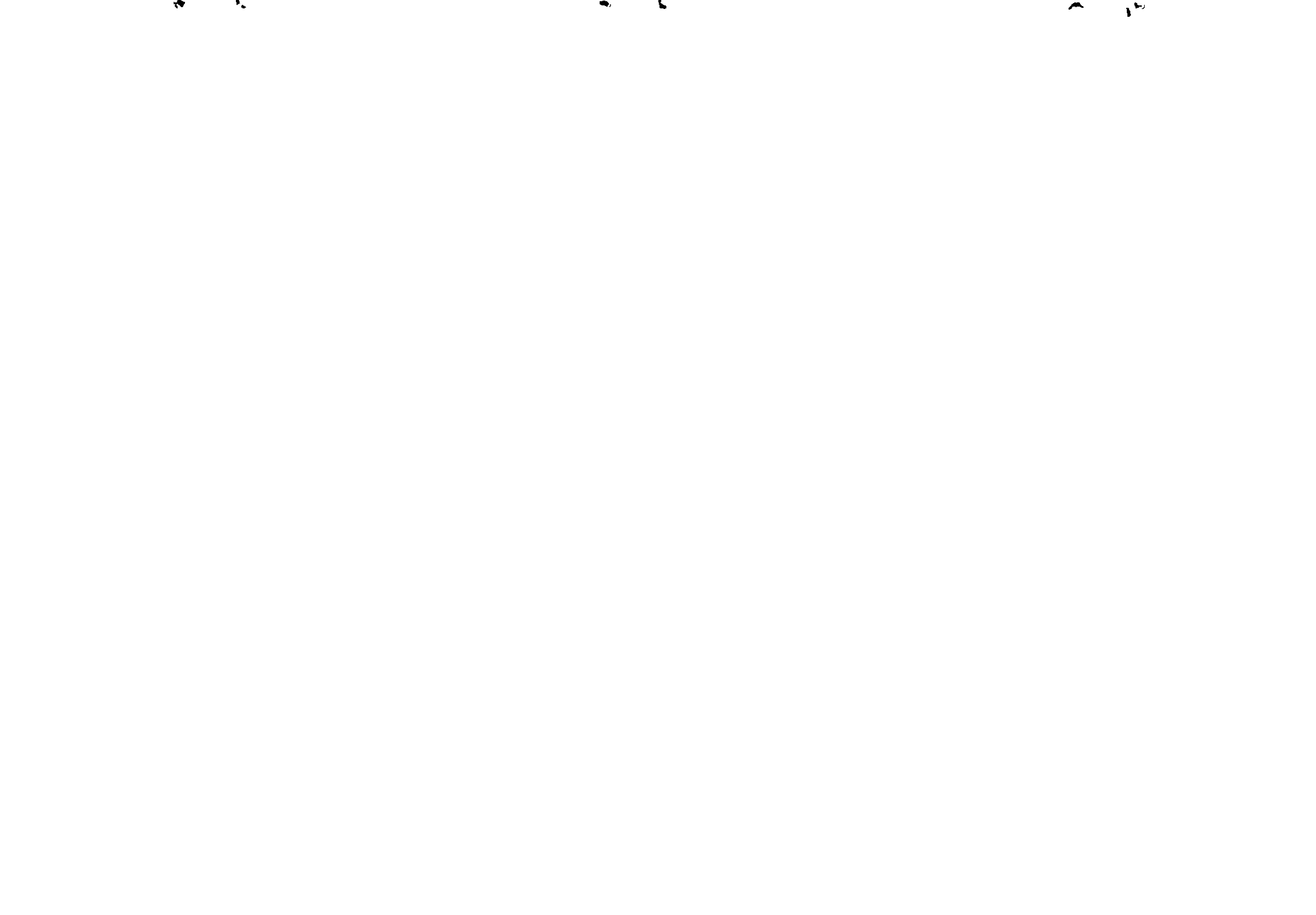
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