Title: A digitization theory of the Weber-Fechner law

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Abstract

Since the publication of the Shannon's article about information theory, there have

been many attempts to apply information theory to the field of neuroscience.

Meanwhile, the Weber-Fechner law describes the relationship between the external

physical stimulus and the human response to that. It is not surprising that we assign

the amount of information to the response in that law. But no one has succeeded in

applying information theory directly to that law: the direct links between information

theory and that response in the Weber-Fechner law have not yet been found. The

proposed theory unveils a link between information theory and that response, and

differs subtly from the field such as neural coding that involves complex calculations

and models. Because our theory targets the Weber-Fechner law which is a

macroscopic phenomenon, this theory does not involve complex calculations. The

proposed theory is expected to mark a new era in the fields of sensory perception

research. This theory must be studied in parallel with the field of microscopic scale

such as neural coding.

Significance statement

Information theory of Shannon is originally a communication theory for communication engineers. Since the birth of information theory, this theory has affected diverse science fields, including neuroscience. A convergence of information theory and neuroscience is being proceeded. Information theory plays a big role in the research on stimulus and response. But past those researches involve complex calculations, because they target microworld such as neurons and cells. Our theory targets the Weber-Fechner law which is a macroscopic phenomenon. So our theory is simple. Our theory should be studied in parallel with the field of the microscopic scale. The author predicts that the new field of research will be created.

1   Introduction

1.1   Psychophysics of Weber and Fechner

In 1834, the German physiologist Ernst Heinrich Weber carried out an exciting experiment. Weber gradually increased the weight which the blindfolded man was holding. With doing so, Weber asked the subject when he first felt the increase of weight. Through this experiment, Weber found out that the least perceptible difference in weight was proportional to the starting value of the weight. For example, if a certain person barely feels the difference between 100 g and 104 g, then he barely feels the difference between 150 g and 156 g. In this case, $100 : 104 = 150 : 156$ holds. Weber's findings were elaborated in Weber (1851). Weber is considered one of the founders of experimental psychology.

Weber's empirical observations were expressed mathematically by Gustav Theodor Fechner who was a Weber's student (Fechner 1860). Fechner called his formulation the Weber's law. Fechner applied the result of the Weber experiment to the measurement of sensation (Ward 1876; Houstoun and Shearer 1930). The Weber-Fechner law established in the field of psychophysics attempts to mathematically describe the relationship between the magnitudes of a physical stimulus and the perceived intensities of the stimulus (Fechner 1860). In this fascinating law, physical stimuli are weight, sound frequency, etc. The Weber-Fechner law states that the relationship between stimulus and its perception is logarithmic. Therefore, if a

stimulus varies as a geometric progression, the corresponding perception varies as an

arithmetic progression. Bear in mind that the stimuli in the Weber-Fechner law have

physical units but that the responses in that law have no physical units. The best

example of Weber-Fechner law is that many musicians adopt the equal temperament

when they tune their musical instruments. Since the nineteenth century, musicians

have tuned virtually all instruments in equal temperament (Berg and Stork 2005).

Equal temperament in music theory is a tuning system in which the octave is divided

into 12 semitones of equal size; note here that the size does not indicate frequency,

but musical interval. In the equal temperament system, the musical instruments are

tuned so that the frequencies of all notes form a geometric sequence. Such a tuning

results in the fact that the musical intervals of every pair of two consecutive notes are

the same (Hopkin 1996; Berg and Stork 2005). This example needs to be explained in

more detail. Transposing or transposition in music means changing the key of a piece

of music. When the musical-instrument accompaniment is too high or too low for us

to sing to, we change the musical accompaniment into a lower or higher key. Even if

we transpose the accompaniment into another key, as if by magic, we don't feel

awkward. There is an essential reason that the accompaniment does not sound

awkward. This is because the musical instruments were tuned to the equal

temperaments. Such a tuning gives an equal perceived step size because pitch is

perceived as the logarithm of frequency. As another concrete example of Weber-

Fechner law, the response of the human ear to change in sound intensity is

logarithmic. Therefore, it is convenient to speak of the sound level defined as

97
$$(10 \text{ decibel}) \log_{10} \frac{\text{intensity of a sound wave}}{\text{standard threshold of hearing intensity}}$$

above the standard threshold of hearing intensity (Kinsler et al. 2000). The standard

threshold of hearing intensity is chosen as 10^{-12} W/m^2, which is near the lowest

limit of the human range of hearing (Kinsler et al. 2000; Halliday et al. 2014). Note

here that because the defining equation of the decibel scale has the same form as

Weber-Fechner law, the perceived magnitude is proportional naturally to a number

obtained after we drop decibel in the sound level. As a third concrete example of

Weber-Fechner law, the response of the human eye to brightness of light is

logarithmic. The stellar magnitudes m_1 and m_2 for two stars are related to the

corresponding brightnesses b_1 and b_2 through the equation

107
$$m_2 - m_1 = \sqrt[5]{100} \log \frac{b_1}{b_2},$$

which results from the Pogson's proposal that a difference of five magnitudes should

be exactly defined as a brightness ratio of 100 to 1 (Pogson 1856; Burke-Gaffney

1963; Mayer 1986; Young 1990; Hearnshaw 1992; Schulman and Cox 1997). Note

here that the stellar magnitude corresponds to the response and hence has no physical

unit.

1.2   Information theory of Shannon

Claude Elwood Shannon defined mathematically the concept of 'the amount

of information' which is sometimes called 'the information content' (Shannon 1948).

Going one step further, he defined mathematically the concept of entropy (Shannon

1948). Due to information theory, we came to be able to quantify the amount of

information transmitted over a communication channel.

1.3   Background of digitization theory

Is it possible to digitize the Weber-Fechner law? Digitizing the Weber-

Fechner law in this article means the following: we define mathematically a new kind

of the information content from the Weber-Fechner law, and going one step further

we define mathematically a new kind of entropy. This digitization may be possible if

we can discover links between the information content formula of Shannon and the

Weber-Fechner formula. The author proposes a theory that we are able to digitize the

Weber-Fechner law. The digitization procedure is as follows. We will add one more

stimulus to the existing various stimuli applied to the Weber-Fechner law. This

addition procedure is based on the analogy between the Weber-Fechner law and the

information content. This additional stimulus and the corresponding perception are

extremely extraordinary. We will digitize the Weber-Fechner law by extending and

reinterpreting that law. Digitization of the Weber-Fechner law which is achieved by

merging information theory with the Weber-Fechner law will lead us to two new

concepts. Firstly, we define a new concept which the author calls 'the amount of

response information'. Ultimately, we end up defining another new concept which the

author calls the 'perception entropy'. The digitization theory of the Weber-Fechner

law treats external physical stimuli and responses, so what this theory says are the

results from a macroscopic perspective.

138   2   Weber-Fechner law and the information content

139   2.1   Derivation of Weber-Fechner law

140   Weber-Fechner law is derived as follows (Nutting 1908). In the Weber's

141   experiment, let S be weight at some instant and let $\mathrm{d}S$ be the differential increment in

142   weight. Let $\mathrm{d}R$ be the differential change in perception. The equation to express the

143   result of the Weber experiment is

144
$$\mathrm{d}R = k\frac{\mathrm{d}S}{S},$$

145   where k is a constant. (There is an important point in connection with k, which will

146   be mentioned in Section 4.) Integrating both sides of this equation yields

147
$$R = k\ln S + C,$$

148   in which case C is an integration constant. Let us suppose that S_0 is the threshold of

149   the stimulus below which a person does not perceive anything. Because $R = 0$

150   whenever $S = S_0$, C must be equal to $-k\ln S_0$. Therefore, we can finally obtain

151
$$R = k\ln\frac{S}{S_0} = K\log_2\frac{S}{S_0},$$

152 which is the mathematically expressed Weber-Fechner law. Let us note that R has no

153 physical unit. But the physical stimulus S has, of course, a physical unit.

2.2 Definition of the information content

155 Shannon defined the amount of information of an event of the occurrence

156 probability P as

$$I = -\log_2 P,$$

158 where the information content I is measured in bits (Shannon 1948). And he defined

159 the entropy of a discrete random variable as the average amount of information of the

160 random variable (Shannon 1948). In order to derive a new kind of the information

161 content and hence a new kind of the entropy, let's redescribe the above equation as

$$I = -\log_2 \frac{P}{P_0},$$

163 where as a matter of course $P_0 = 1$.

3 Derivation and interpretation of digitization theory

3.1 Special derivation

166 In the equation

$$R = K \log_2 \frac{S}{S_0},$$

168   for all practical purposes $S \geq S_0$. In the equation

169
$$I = -\log_2 \frac{P}{P_0},$$

for all practical purposes $P \leq P_0$. We can interpret $P_0 = 1$ as the threshold of P. Most

importantly, we can interpret I as a response to P; it would be nice for us to call P a

bizarre kind of stimulus. However, this bizarre stimulus fundamentally differs from

other ordinary physical stimuli in that it is simply a number which has no physical

unit. Very important another fact is that the higher P is, the lower I is while the

higher S is, the higher R is. But it is not essential that I is a strictly decreasing

function with respect to P while R is a strictly increasing function with respect to S.

The essentially important facts are that R and I with respect to S and P, respectively,

are logarithmic and that both S and P have the threshold values. From these

viewpoints, consequently, the author cannot choose but say that the occurrence

probability is an anomalous kind of stimulus. It would be appropriate that we name

the occurrence probability the 'mathematical stimulus' in that the occurrence

probability has no physical unit. This mathematical stimulus is under the government

of the Weber-Fechner law. By widening the range of applications of the Weber-

Fechner law, we come to be capable of combining information theory with the

Weber-Fechner law of psychophysics.

For simplicity, in current subsection we set K to 1 for all sensations. (We

make complete descriptions in Subsection 3.2. In that subsection we make no

assumptions about K.) Then, we can say that when $S = S_0$, the amount of perception

is 0 bits and that when $S = 2S_0$, the amount of perception is 1 bits. But there is one

thing to be careful about. In reality, different people have different S_0. So, different

people have different amount of response information for the identical S. In addition,

as the body of a person ages, S_0 changes. So, even the response information content

of a person changes with age for the identical S. Because of this fact, the response

information content is quite different from the information content of Shannon. In

defining equation of the information content of Shannon, P_0 is fixed at 1.

If the physical stimulus is the weight, the problem is simple. In order to define

a new kind of entropy, let's consider the sound wave from now on. Because a pure

tone which is a sound wave with a sinusoidal waveform is characterized by a

frequency and a displacement amplitude, considering only either frequency or

amplitude may be meaningless. Thus, we consider both simultaneously. For the sake

of argument let's consider a pure tone of frequency f and amplitude A. Physics

entropy and Shannon entropy are subject to the constraint that the total sum of

probabilities is 1. On the other hand, there is not such a constraint in the defining

equation of the new kind of entropy. The physical quantities A and f change much

more freely. Therefore, it is impossible that we define the new entropy in the same

form as the Shannon entropy. In order words, we must not define the new entropy as

207
$$\frac{A}{A_0} \log_2 \frac{A}{A_0} + \frac{f}{f_0} \log_2 \frac{f}{f_0},$$

208   which is of the form of Shannon entropy. Before we define the new entropy, we first

209   define the total amount of response as

$$\log_2 \frac{A}{A_0} + \log_2 \frac{f}{f_0} = \log_2 \frac{Af}{A_0 f_0},$$

211   where $\log_2 \frac{Af}{A_0 f_0}$ is measured in bits. In other words, the total amount of response is

212   defined so that it equals the sum of the amounts of each individual response.

213   Consequently, we define the new entropy as the arithmetic mean of responses being

214   considered; that is, new entropy is defined as

$$\left(\log_2 \frac{Af}{A_0 f_0}\right)/2.$$

The unit of this quantity is bits/response. Please note that in the current situation

there are two responses: one response due to the amplitude and the other due to the

frequency. The author calls this new quantity the 'perception entropy' or the

'percentropy'. The term 'percentropy' is a new compound word formed from

perception and entropy. The percentropy does not correspond perfectly to the

Shannon entropy. Nonetheless, it is useful to define the concept of percentropy

because percentropy is linked directly with physics energy. We are easily able to

understand such a fact. As an example, consider two pure tones. One tone has

amplitude A_1 and frequency f_1. The other tone has amplitude A_2 and frequency f_2.

The percentropy of the former tone is $\left(\log_2 \frac{A_1 f_1}{A_0 f_0}\right)/2$. The percentropy of the latter

tone is $\left(\log_2 \frac{A_2 f_2}{A_0 f_0}\right)/2$. The equality $A_1 f_1 = A_2 f_2$ is a necessary and sufficient

condition for the fact that the percentropies of the two tones are the same. And

$A_1 f_1 = A_2 f_2$ if and only if $(A_1 f_1)^2 = (A_2 f_2)^2$. We can therefore immediately

conclude that the equality $A_1 f_1 = A_2 f_2$ is a necessary and sufficient condition for the

fact that the energies of the two tones are the same. Let us now consider two

compound tones. One compound tone consists of a pure tone which has amplitude

A_{11} and frequency f_{11} and a pure tone which has amplitude A_{12} and frequency f_{12}.

The other compound tone consists of a pure tone which has amplitude A_{21} and

frequency f_{21} and a pure tone which has amplitude A_{22} and frequency f_{22}. The

percentropy of the former compound tone is

236
$$\left(\log_2 \frac{A_{11}}{A_0} + \log_2 \frac{f_{11}}{f_0} + \log_2 \frac{A_{12}}{A_0} + \log_2 \frac{f_{12}}{f_0}\right)/4 = \left(\log_2 \frac{A_{11} f_{11} A_{12} f_{12}}{A_0{}^2 f_0{}^2}\right)/4.$$

237   The percentropy of the latter compound tone is

238
$$\left(\log_2 \frac{A_{21}}{A_0} + \log_2 \frac{f_{21}}{f_0} + \log_2 \frac{A_{22}}{A_0} + \log_2 \frac{f_{22}}{f_0}\right)/4 = \left(\log_2 \frac{A_{21} f_{21} A_{22} f_{22}}{A_0{}^2 f_0{}^2}\right)/4.$$

The equality $A_{11} f_{11} A_{12} f_{12} = A_{21} f_{21} A_{22} f_{22}$ is a necessary and sufficient condition for

the fact that the percentropies of the two compound tones are the same. And

$A_{11} f_{11} A_{12} f_{12} = A_{21} f_{21} A_{22} f_{22}$ if and only if $(A_{11} f_{11})^2 (A_{12} f_{12})^2 =$

$(A_{21} f_{21})^2 (A_{22} f_{22})^2$. Therefore, in this compound-tones situation, the equality of the

two percentropies is a necessary and sufficient condition for the fact that the products

of energies of the two pure tones composing each compound tone are the same.

245   **3.2   General derivation**

246   Now we do not assume that Ks are 1 for all sensations. Because Ks are simply

247   constants, we just have to define the response information content for a single

248   stimulus as

249
$$K \log_2 \frac{S}{S_0},$$

250   where $K \log_2 \frac{S}{S_0}$ is measured in bits. Let's consider a pure tone of frequency f and

251   amplitude A. Let K_A and K_f be the corresponding constants. We first define the total

252   amount of response as

253
$$K_A \log_2 \frac{A}{A_0} + K_f \log_2 \frac{f}{f_0} = \log_2 \left(\frac{A}{A_0}\right)^{K_A} \left(\frac{f}{f_0}\right)^{K_f},$$

254   where $\log_2 \left(\frac{A}{A_0}\right)^{K_A} \left(\frac{f}{f_0}\right)^{K_f}$ is measured in bits. In other words, the total amount of

255   response is defined so that it equals the sum of the amounts of each individual

256   response. And then we define the new entropy as the arithmetic mean of responses

257   being considered; that is, new entropy is defined as

258
$$\left\{\log_2 \left(\frac{A}{A_0}\right)^{K_A} \left(\frac{f}{f_0}\right)^{K_f}\right\}/2.$$

259 The unit of this quantity is bits/response. Because the amplitude and the frequency

260 change independently of each other, we took the arithmetic mean. Notice that the

261 percentropy is not linked directly with physics energy unless K_A equals K_f.

3.3 Utilization and interpretation

According to the proposed theory, it comes to be possible that we utilize

percentropy (or the total amount of response information) for quantifying the

sensitivities of the sensory organs of the humans; at this time the sensitivity is a

function of the stimuli and their thresholds. This quantification allows us to compare

the sensitivities of the sensory organs of people. It is best to explain an example. For

simplicity, we set $K = 1$ for all sensations. Let us consider the case of the hearing.

There are two people. For a person X, $A_0 = 33$ pm and $f_0 = 20$ Hz. For another

person Y, $A_0 = 22$ pm and $f_0 = 60$ Hz. Let us suppose that two people hear a sound

wave with an amplitude of 330 pm and a frequency of 1000 Hz. Then, the

percentropy of the hearing organ of X is

$$\left(\log_2 \frac{Af}{A_0 f_0}\right)/2 = \left(\log_2 \frac{330 \times 1000}{33 \times 20}\right)/2 \approx 4.483 \text{ bits/response,}$$

274 and the percentropy of the hearing organ of Y is

$$\left(\log_2 \frac{Af}{A_0 f_0}\right)/2 = \left(\log_2 \frac{330 \times 1000}{22 \times 60}\right)/2 \approx 3.983 \text{ bits/response.}$$

Thus the hearing organ of X is superior to that of Y. In fact, as we can very easily

understand, we need only $A_0 f_0$ in comparing superiority of the hearing organs of

people. In conclusion, we can say that the information-theoretical measure of the

superiority of the hearing organs of people is $A_0 f_0$. Of course, we can easily devise

$A_0 f_0$ as a measure of the superiority, without any knowledge appearing in this article.

The key point here is that the measure $A_0 f_0$ is justified by the digitization theory of

the Weber-Fechner law; i.e., $A_0 f_0$ is valid from the information-theoretical

perspective. Please note that we have set $K = 1$ regardless of the type of sensation. If

we had not set so, a more complex quantity other than $A_0 f_0$ would have been

obtained as the measure of the superiority.

Even if we do not assume that $K = 1$ for all sensations, we can still compare

the sensitivity of the sensory organs of people. But the measure of the superiority is

no longer $A_0 f_0$. If we assume that the values of K_As for people are all identical and

that the values of K_fs for people are all identical, then the measure of the superiority

is $A_0^{K_A} f_0^{K_f}$.

So these are good. The question is then what 4.483 bits/response and

3.983 bits/response mean; equivalently, what 8.966 bits and 7.966 bits mean. It is

clear that the response information content is proportional to the duration of the

external stimulus. And, probably, the duration of the response to that stimulus is

quantized. In order words, the duration of the response is always some integral

multiple of elementary duration. And, probably, 8.966 bits and 7.966 bits is the

information content of elementary duration. According to this hypothesis, the

information content during some time interval is equal to the information content

during elementary duration multiplied by some integer. This hypothesis should be

verified through other works.

4 Future challenges

The stimuli and responses that have been mentioned so far were about the

external physical stimuli and responses to those; the stimuli and responses that have

been mentioned so far have not included the stimuli and responses that neural coding

handles. In order for the digitization theory of the Weber-Fechner law to be firmly

established, we need to elucidate the relationships between microscopic and

macroscopic theories on stimuli and responses.

Meanwhile, past information-theoretical researches on stimuli and responses

have mainly focused on microanalysis (MacKay and McCulloch 1952; Rapoport and

Horvath 1960; Werner and Mountcastle 1965; Stein 1967; Borst and Theunissen

1999; Dimitrov and Miller 2001; Koechlin and Summerfield 2007; Ikeda and Manton

2009). Those researches involve complex models and calculations. Because the

Weber-Fechner law is a macroscopic phenomenon which is a manifestation of

numerous microscopic outcomes, the digitization theory of the Weber-Fechner law

involves far less complex calculations. In the future, macroanalyses as well as

microanalyses should be done together. Future researches will have to focus on

elucidating the relationships between microscopic and macroscopic theories.

In particular, there is an important point in connection with K. K in

psychophysics is originally an experimentally determined proportionality constant.

But, K for each sense in our theory must be determined by comparing microscopic

works and our theory. Additionally, author guesses that regardless of the type of

sense, an expression for K includes the Boltzmann constant or the Boltzmann-like

constant which serves to bridge the macroworld and the microworld. Besides the

values of Ks, we must find the value of the elementary duration of response for each

sense. Once Ks and the elementary durations are all determined, the field of research

on stimuli and responses is expected to develop rapidly.

5   Conclusions

So as to achieve our goal, the links between the Weber-Fechner formula and

the information content formula must be found. These links are the key idea behind

the digitization theory of the Weber-Fechner law. A link presented in this article is an

undiscovered link. Further insight into the defining equation of the information

content is to recognize that we can redescribe the defining equation using the

threshold of probability. Consequently, the definition of the information content for

physiological response was chosen to be in analogy with the definition of the

information content which was given by Shannon. Based on the analogy between the

defining equation of the information content and the formula describing the Weber-

Fechner law, the author was able to theorize about the digitization of the Weber-

Fechner law. On the one hand, the response information content has some features of

conventional information theory. On the other hand, the response information content

has the different features from conventional information theory. We also defined the

perception entropy which means the average amount of the human response to

external physical stimuli; the unit of the perception entropy is bits/response. The

perception entropy is very information-theoretic, in the sense that its definition

originates in the combination of information theory and the Weber-Fechner law.

The digitization theory of the Weber-Fechner law has the advantage of being

powerful. The reason is that the Weber-Fechner law is a macroscopic phenomenon,

and hence the calculations in this theory are straightforward. In order for our theory to

be firmly established, we need to elucidate the relationships between microscopic and

macroscopic theories on stimuli and responses. If the author's theory turns out to be

of great usefulness, new field of research will be created. Then, that new field will

find applications in many diverse fields such as physiology and medicine. It will be

good for us to name this new field 'perceptional information theory' or 'perceptual

information theory'.

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