Alpha EEG Activity Depends on the Individual Dominant Rhythm Frequency

Olga Mikhailovna Bazanova PhD, DSci

Biofeedback Software Laboratory, State Research Institute of Molecular Biology and Biophysics, Siberian Branch, Russian Academy of Medical Sciences, Novosibirsk, Russia

Published online: 21 Nov 2012.
REVIEW ARTICLE

ALPHA EEG ACTIVITY DEPENDS ON THE INDIVIDUAL DOMINANT RHYTHM FREQUENCY

Olga Mikhailovna Bazanova

Biofeedback Software Laboratory, State Research Institute of Molecular Biology and Biophysics, Siberian Branch, Russian Academy of Medical Sciences, Novosibirsk, Russia

Exploring the role of the electroencephalographic (EEG) alpha oscillations in the processes of central and autonomic regulation of the cognitive and psychomotor performance generates considerable interest. However, until now there was no accurate definition of what alpha activity is, and which indices characterize it. This is a theoretical article that attempts to define EEG alpha-activity phenomena; define their physical, molecular, and morphological nature; and highlight the difference in its indices connected with different individual alpha frequencies and their role in optimal functioning. Specifically, this article examines individual alpha activity indices: (a) the individual alpha peak frequency; (b) activation magnitude measured by estimating the amount of alpha amplitude suppression in response to eyes-open and individual alpha band width; and (c) three alpha “autorhythmicity” indices: intraspindle amplitude variability, spindle length, and steepness. Throughout, the article provides a number of suggestions on how alpha activity indices are connected with the individual spectral distribution of the alpha frequency and why these individual alpha activity peculiarities are important for neurofeedback training effectiveness.

INTRODUCTION

Since the time of the eminent Russian physiologist Ivan Michailovich Sechenov, it was pointed out that the body is an amazing system with many complexities and a special potential for improving itself with feedback connections (Bernstein, 1967). Such self-improvement may target what is now called “peak performance” (R. N. Singer, 2002; Vernon, 2009) or “optimal functioning” (Keefe, 1978) and can be achieved with biofeedback technology (Bazanova, Memaya, & Shtark, 2009; Hanslmayr, Sauseng, Doppelmayr, Schabus, & Klimesch, 2005; Vernon & Gruzelier, 2008). The recent identification of specific brain signatures involved in optimal cognitive and psychomotor functioning demonstrates its association with general enhancement of the electroencephalographic (EEG) alpha activity (Bazanova & Aftanas, 2008; Bazanova, Gvozdev, Mursin, Verevkin, & Shtark, 2003; Hummel et al., 2004; Baumeister, Reinecke, Liesen, & Weiss, 2008). Thus, this article argues that EEG alpha activity reflects a certain type of self-control (top-down) process and that the properties of this process-related activity may vary among individuals.

Despite the key role of EEG alpha activity in optimal functioning—known since Berger’s time—it is not clear which qualities characterize it (e.g., increasing or decreasing alpha amplitude, left or right shift of spectral alpha activity...
peak frequency, or alpha synchronization or desynchronization). Therefore, this review aims to determine (a) what phenomena could be called “EEG alpha activity” and what brain mechanisms are underlying this, (b) which EEG parameters characterize EEG alpha activity, and (c) their intra- and interindividual variability.

From physics we know that wave activity is characterized by three main parameters: frequency, amplitude, and phase. So it is possible to define three main index groups of the alpha wave activity (Figure 1):

1. Frequency of the dominant posterior EEG rhythm (Niedermeyer, 1999; Nunez, Wingieier, & Silberstein, 2001), which is assessed as individual alpha peak frequency in an eyes-closed condition (IAPF; Angelakis, Stathopoulou, & Frymiare, 2007; Bazanova, 2011; Hooper, 2005).

2. Responsiveness/activation, which is assessed by alpha amplitude suppression in response to opening of the eyes (Barry, Clarke, Johnstone, Magee, & Rushby, 2007; Del Percio et al., 2011; Kirshfieldt, 2005; Nunez et al., 2001), and width of the frequency range of amplitude suppression, or individual alpha bandwidth (Bazanova & Aftanas, 2006, 2008).

3. “Autorhythmicity” (or bursting or spindle forming process), which is a result of phase resetting (Ivanitsky & Lebedev, 2007; Thatcher, North, & Biver, 2008; Timofeev et al., 2002).

THE IAPF

Many empirical studies present evidence that IAPF varies intraindividentally as a function of age—it increases through childhood until pubertal age (Bazanova, 2008; Stroganova, Orekhova, & Posikera, 1999) and decreases after 40 years of age (Clark et al., 2004). A few studies have shown the hormonal influence on IAPF—it increases with progesterone activity enhancement during menstrual cycle in women (Baker & Colrain, 2010; Bazanova & Mernaya, 2008; Creutzfeldt et al., 1976; Solis-Ortiz, Guevara, & Corsi-Cabrera, 2004), and in conditions when cortisol blood level increases (Becker, Creutzfeldt, Schwibbe, & Wuttke, 1980; Tops, van Peer, Wester, Wijers, & Korf, 2006). Moreover, IAPF varies depending on

![Figure 1](Image)

**FIGURE 1.** Alpha activity indices. Note. EC = eyes closed – bold line spectra; EO = eyes open – dotted line spectra. (Color figure available online.)
the state individuals are during resting (baseline) conditions, and their personal involvement in a cognitive performance task (Hummel et al., 2004; Klimesch, Sauseng, & Hanslmayr, 2007; Ng & Raveendran, 2007). Good task performance is related with higher IAPF (Angelakis, Lubar, & Stathopoulou, 2004; Klimesch, Schimke, & Pfurtscheller, 1993), and neurofeedback training (NFT) directed on increasing IAPF may be used to improve cognitive performance (Angelakis et al., 2007). A decrease in IAPF is related to a drop in performance (Klimesch et al., 1993) or to fatigue (Ng & Raveendran, 2007).

IAPF doesn’t always increase after NFT to enhance the upper alpha power. In a recent study investigating the enhancement of the upper alpha power with NFT and its impact on the peak performance of musical achievement, we showed that IAPF increased only in students with a resting baseline alpha frequency lower than 10 Hz (LAF subjects). On the contrary, students with original IAPF higher than 10 Hz (HAF) did not change or even decreased their IAPF in a post-NFT resting condition (Bazanova et al., 2009). Hence, it could be proposed that differences in alpha peak frequency reflect different mechanisms of brain activation and alpha wave generation.

In a study with twins, Gavrish and Malykh (1994) proposed that the nature of different alpha-frequency components is heritable. Other genetic studies confirm the high heritability of the resting IAPF (Anohkin, Muller, Lindenberger, Heath, & Myers, 2006; Smit, Wright, Hansell, Geffen, & Martin, 2006). It is now evident that IAPF reflects individual genetic influences on the underlying neural mechanisms of the alpha waves generation (Steriade, 1999; Steriade & Timofeev, 2003).

Evidence from animal models suggests that alpha wave frequency is a result of the local cortical network tuning, which depends on Ca2+ T-channels activity (Lopes da Silva, 1991; Steriade, 1999; Steriade & Timofeev, 2003). It was shown that blocking of the Ca2+ oscillations (an intracellular signal, rather than an EEG one) is associated with blocking of the spindle wave refractory period, which lies in the origin of the 6 to 10 Hz oscillations (Lüthi & McCormick, 1998). So frequency of the Ca2+ oscillations determines the dominant brain frequency (Lüthi & McCormick, 1998). These experimental results support the hypothesis by Michael Livanov and John Eccles about the inhibitory role of delaying the rhythmic waves generated in the thalamus for brain and mind self-regulation (Eccles, 1994; Livanov, 1984; Livanov & Dumenko, 1987).

Among different alpha frequency measurements (individual alpha peak frequency, mean and center of gravity), IAPF is preferential for studying between individual differences (Hooper, 2005). As shown by Greg Hooper (2005), the contrasts in upper and lower alpha bands may be explained in terms of the variability or distribution of the peak alpha frequency itself. However, the definition of appropriate frequency bands and the choice of recording reference limit the interpretability of quantitative EEG. The analysis of fixed frequency bands could therefore blur the real alpha peak, masking the age- or function-related differences. To study the endophenotypic qualities of the alpha frequency, we asked, “In which experimental condition and where topographically is IAPF mostly stable and reproducible?” We then analyzed the “test–retest” EEGs from 96 male subjects aged 26 to 40 recorded twice over 14 to 15 days. EEG was recorded in resting eyes-closed and eyes-open conditions and analyzed in a fixed standard 8–12 Hz frequency band, as well as in an individually determined alpha frequency band. It appeared that the intraindividual correlation coefficient was strongest in posterior brain areas, in the eyes-closed condition, and assessed with individually determined alpha band. The weakest intraindividual correlation coefficient was calculated in anterior areas, in the eyes open condition, and using the fixed standard 8–12 Hz alpha range (Bazanova, 2011). We did not find the laterality influence on the IAPF mean. Bodenmann and coauthors showed, too, that genotype-dependent differences in IAPF were independent from EEG recording location (Bodenmann et al., 2009). So it could be
concluded that alpha generation in healthy individuals has a universal origin.

What is the functional role of the human alpha rhythm frequency? An eminent Russian psychophysiologist, Artur Lebedev (1994, 2006), has proposed that “cyclical oscillations in the alpha rhythm determine the capacity and speed of working memory. The higher the frequency the greater the capacity and the speed of memory” (Lebedev, 1994, p. 257). Then, Klimesch et al., (1993) argued that thalamo-cortical feedback loops oscillating within the alpha frequency range allow searching and identification of encoded information. He speculated that faster oscillating feedback loops would correspond to faster access to encoded information. These theories are supported by the results of the following studies. Klimesch et al. (1993) found that the alpha peak frequency of good working memory performers lies about 1 Hz higher than that of poor working memory performers. We had shown that reading is better in musicians with higher, rather than lower, IAPF (Bazanova et al., 2009). The same results were revealed by the functional Val158Met polymorphism of the catechol-O-methyltransferase (COMT) gene, where LAF-Val/VAL homozygote subjects demonstrated a lower score in a counting task than HAF-Met/MET homozygote young men (Bodenmann et al., 2009).

With the question “Are smarter brains running faster?” Posthuma, Neale, Boomsma, and de Geus (2001) concluded that both peak frequency and the dimensions of IQ were highly heritable (range = 66–83%). But, in this study, a large part of the genetic variance in alpha peak frequency as well as in working memory and processing speed was due to nonadditive factors. Moreover, there was no evidence of a genetic correlation between alpha peak frequency and any of the four WAIS dimensions. So there must be additional EEG indices predicting cognitive ability.

In our experiment using the median of posterior IAPF from a resting eyes-closed condition, 96 healthy male subjects were divided into two groups: those with low (LAF, IAPF < 10 Hz) and those with high (HAF, IAPF ≥10 Hz) alpha frequency. It appeared that LAF and HAF subjects differed in psychometric strategies for achieving success in nonverbal creative tasks. LAF subjects emphasized originality, whereas HAF emphasized fluency in reaching the same performance score in the Torrance test (Bazanova & Aftanas, 2008). However, the subjects with highest and lowest IAPF level showed the highest originality score in a nonverbal creativity task (see Figure 2).

We proposed that different behavior strategies observed in LAF and HAF subjects could be due to (a) genetic factors for individual patterns of spontaneous waking EEG and (b) different neurophysiological mechanisms of brain activation in low and high alpha frequency ranges (Kirshfeldt, 2005).

THE RESPONSIVENESS/ACTIVATION MAGNITUDE

Vaez Mousavi and colleagues conceptualized two aspects of the energetic state, “arousal” and “activation” (Vaez Mousavi, Barry, Rushby, & Clarke, 2007). “Arousal” has been defined as the energetic state at any particular time, and task-related “activation” as the task-related change in state from resting baseline to the task.
situation. Accordingly, EEG data activation magnitude indicates the amount of alpha amplitude suppression in response to eyes opening (Barry et al., 2007). Recent investigations showed that EEG measures of visual and cognitive activation are a promising avenue of study in the search for putative endophenotypes for individual cognitive strategy type (Ivanitsky, Ivanitsky, & Sysoeva, 2009; Loo & Smalley, 2008) or for ADHD (Loo et al., 2010).

Doppelmayr and colleagues found a significant interaction of task difficulty and activation in the upper alpha band, indicating both a weaker activation for the highly intelligent subjects during the easy tasks and a significant increase from easy to difficult tasks for this group only (Doppelmayr, Klimesch, Hodlmoser, Sauseng, & Gruber, 2005). Recently, Del Percio and coauthors tested the hypothesis that, compared with nonathletes, elite athletes are characterized by a reduction of reactivity of EEG alpha rhythms (about 8–12 Hz) to eyes opening in the condition of resting state, as a possible index of spatially selective cortical activation (i.e., “neural efficiency”; Del Percio et al., 2011). It could be proposed that activation magnitude variability is dependent on cognitive task difficulty and professional experience and is particularly associated with reaction time (Vaez Mousavi et al., 2007) and psychomotor task performance (Bazanova, Kondratenko, Kondratenko, Mernaya, & Zhimulev, 2007). Moreover, the amount of alpha suppression in response to eyes opening depends on the neurohumoral status (Bazanova & Mernaya, 2008; Mantanus, Ansseau, Legros, & Timsit-Berthier, 1988) and even the direction of eye movement (V. Kaiser, Clemens, Leeb, Neuper, & Pfurtscheller, 2009).

A decreased amount of alpha suppression in response to eyes opening was demonstrated in spinal-cord-injured individuals (Thuraisingham, Tran, Boord, & Craig, 2007) in comparison with healthy participants, and in patients with attention deficit (Barry et al., 2003). Increased activation magnitude is a result of upper alpha stimulating neurofeedback training (Bazanova et al., 2009; Schmelkina, 1999). Of interest, reactivity to eyes opening increases from 3 to 20 years of age and decreases after 40 years of age in HAF but doesn’t change in LAF healthy male subjects (Bazanova, 2008).

Based on simulations of the EEG, Pfurtscheller and Lopes da Silva (1999) hypothesized that with an increasing number of interconnecting neurons and therewith an increasing number of coherently activated neurons, the amplitude increases and the frequency decreases. Furthermore, they support that EEG amplitude is negatively correlated with frequency not only between frequency bands, but also within frequency subcomponents (Pfurtscheller & Lopes da Silva, 1999).

Measuring ERPs from the visual cortex after the presentation of checkerboard stimuli, Mazaheri and Jensen (2006) found that alpha (M = 10 Hz) oscillations preserve their phase relationship after visual stimuli with respect to before the stimuli, while theta (M = 6.6 Hz) phase-resetting oscillations are responsible for visually evoked responses. They concluded that alpha activity plays a modulatory role in perception rather than generating visually evoked responses (Mazaheri & Jensen, 2006). In EEG-fMRI coupled investigations, Laufs and coauthors (2006) showed that spontaneous reductions in alpha amplitude over occipital areas is connected to fronto-parietal fMRI activation when (a) theta power is low and beta power is high and (b) PAF is high, while it is connected to occipital fMRI activation when beta power is low and theta power is high and PAF is low and more variable. These different interrelations between the amount of alpha amplitude suppression (desynchronization) and other EEG indices in low- and high-frequency ranges is consistent with the idea that alpha desynchronization serves an integrative role through a corticocortical “gating” (Kirshfeldt, 2005; Tenke & Kayser, 2005). This supports the idea that the neurophysiological mechanisms by which activation is generated are different in LAF and HAF subjects (Bazanova, Jafarova, Mazhirina, Mernaya, & Shtark, 2008), which is what was theoretically supported in previous investigations (Kirshfeldt, 2005; Mazaheri & Jensen, 2006).
Until now, the alpha frequency range boundaries were defined at 8–12/13 Hz on the basis of general agreement, without a theoretical basis and without respect to functional features of alpha rhythm reactivity to visual stimulation (D. A. Kaiser, 2001; Thatcher et al., 2008; Arns, Gunkelman, Breteler, & Spronk, 2008). Meanwhile, the individual alpha bandwidth (IABW) is dependent on (a) the age—it enlarges from 3 until 20 years (Bazanova, 2008); (b) neurohumoral status in women—in high estrogen and progesterone stages, alpha band is wider than in lower hormonal phases during the menstrual cycle (Bazanova & Mernaya, 2008); and (c) gender—women have a higher low alpha band boundary frequency than men (Bazanova & Mernaya, 2008; Carrier, Land, Buyse, Kupfer, & Monk, 2001). Moreover, individual alpha frequency bandwidth could vary in accordance with brain activation (Bazanova & Aftanas, 2008; Kaiser, 2001; see Figure 3).

For example, IABW is wider in highly skilled professional musicians than in nonmusicians (Bazanova et al., 2003), and IABW is positively correlated with the overall Torrance creativity coefficient and flexibility in creative task performance (Bazanova & Aftanas, 2008), academic achievement (Bazanova & Aftanas, 2006), and biofeedback training efficiency (Bazanova et al., 2007).

Research by D. A. Kaiser (2001) and Bazanova and Aftanas (2010) supported the importance of accounting for individual alpha band range in neurofeedback, showing that neurofeedback training applied in individual EEG frequency ranges was much more efficient than neurofeedback training of standard EEG frequency ranges (SNFT; Figure 4). The SNFT protocol of decreasing theta/beta ratio not only showed no psychometric improvement but also aggravated the clinical condition of the ADHD patient (Bazanova & Aftanas, 2010). Moreover, we have shown that the enhancement of the individual alpha bandwidth is one of the main indicators of success in both neurofeedback and other modalities of biofeedback training (Bazanova et al., 2008; Bazanova et al.,

![FIGURE 3. EEG spectral power in the parietal-occipital areas in the eyes-closed (white) and eyes-open (gray) conditions in healthy subjects. Note. Spectral power decreases by more than 20% from baseline within individual alpha ranges (black); (a) and (b) examples of different alpha-bandwidths. The abscissa shows frequency in Hz; the ordinate shows spectral power at the frequencies indicated in $\mu$V$^2$; IABW = individual alpha bandwidth; IAPF = individual alpha peak frequency.](image)
2007), indicating that the ability for enhancing self-control could be associated with the individual alpha band broadening (see Figure 4).

Activation magnitude measured as amount of alpha amplitude suppression and bandwidth during desynchronization raises many questions impossible to answer with simple spectral analysis. For example, it is not clear (a) whether the change of total power of particular brain alpha oscillations results from a change in the number of occurrences per minute rather than the change of the average oscillations amplitude, and (b) whether a change in the total power of alpha oscillations relates to the whole analyzed signal or rather to a small portion (Kaplan, Borisov, Shishkin, & Ermolaev, 2002). Thus, regardless of how powerful or statistically significant the different estimations of averaged EEG effects may be, it is difficult to make meaningful interpretations if the estimations are not matched to the EEG piecewise stationary structure (Kaplan et al., 2002).

To overcome such limitations of conventional spectral analysis based on averaging procedures and to reveal dynamic and temporal characteristics of alpha activity, an entire set of individual short-term stationary EEG segments may be obtained (Brodsky et al., 1999; Towers & Allen, 2009). Nonstationary phenomena are present in the EEG, usually in the form of transient events, such as alternation of relatively homogenous intervals (bursting segments) with different statistical features (e.g., with different amplitude or variance; Lopes da Silva, 1991). The idea that alpha oscillations have a spindle-like form only during sleep (Niedermeyer, 1999) has been contradicted by Kellaway (2003), who described the so-called lambda waves (8–13 Hz), which may be identified using simple procedures. The lambda wave is believed to represent occipital lobe activity in a person actively reading or scanning a room. Often, the subjects have their eyes open and are looking carefully

**FIGURE 4.** Spectral power (A) and psychometric test performance (B) of a boy with attention deficit disorder in baseline condition (BC), after 10 neurofeedback sessions decreasing the theta/beta ratio in standard EEG ranges (SNFT: theta = 4–8 Hz and beta = 13–18 Hz) and in individual EEG ranges (INFT: theta = 3–5 Hz, beta = 10–14 Hz). Note. A: Gray areas denote spectral power in individual alpha band. Spectra power in baseline condition = dotted line; numbers with arrows = individual alpha peak frequency. B: Black bars indicate number of words read, and white bars indicate time to perform the task.
at the ceiling tiles. The technician will have the patient reproduce the activities that they felt caused the waves to appear in the first place (Kellaway, 2003). It was proposed that the physiological basis of sleep spindles is probably very similar to lambda and alpha waves. To determine if the brain activation state would modulate the composition of alpha spatial microstates (spindles), Cantero and colleagues (1999) used spatial segmentation methods to show that (a) the mean duration of alpha spindles is longer in relaxed wakefulness than in drowsy periods and REM sleep, and (b) the number of different amplitude values are more abundant in drowsiness than in other brain states.

Thus, overall, brain activation mechanisms could become clearer if we consider the third important EEG alpha activity feature—spindle-ability of alpha oscillation.

**AUTORHYTHMICITY**

The spindle-form or autorhythmicity of alpha oscillations is a product of the dynamics of neuronal assemblies in the underlying cortical activity (Dorokhov, 2003; Lehmann, Strik, Henggeller, & Koukkou, 1994; Livanov & Dumenko, 1987; W. Singer et al., 1997). Starting from Livanov’s (1984) studies, it has been shown that spindle oscillations are essential for memory formation (Gais, Plihal, Wagner, & Born, 2000; Lebedev, 1994, 2006), and associated with short- and mid-term synaptic plasticity (Steriade & Timofeev, 2003). A probable molecular mechanism of these phenomena was proposed by Destexhe and Sejnowski (2003), who suggested that spindling may activate the protein kinase—a molecular “gate”—thus opening the door for gene expression and allowing long-term changes to take place following subsequent inputs.

Average amplitude within a segment indicates the volume of the neuronal population. The more neurons recruited into an assembly through local synchronization of their activity, the higher will be the oscillation amplitude of the corresponding assembly (Brodsky et al., 1999; Livanov & Dumenko, 1987; Lopes da Silva, 1991). The assumption that alpha amplitude reflects inhibition may, at first glance, appear contradictory to the idea that alpha plays an active role in information processing, but the idea is that inhibition is an important factor that controls the exact timing of an oscillation. Thus, inhibition helps to establish a highly selective activation pattern (Klimesch et al., 2007).

Average spindle duration represents the functional lifespan of the neuronal population or the duration of operations produced by such a population (Kaplan et al., 2002; Maltseva & Masloboev, 1997). It has been shown that longer spindles indicate a more relaxed state (Huupponen et al., 2008). The spindle lifetime is correlated with fluency in cognitive task performance (Bazanova & Aftanas, 2008), and efficiency of a single biofeedback training session (Bazanova et al., 2007; Bazanova et al., 2009). In addition, alpha-spindles are longer in highly skilled musicians than in amateurs (see Figure 5; Bazanova et al., 2009).

The shortest alpha segments belong to HAF subjects with the highest individual alpha peak frequencies and LAF subjects with the lowest individual alpha peak frequencies (Figure 5; Bazanova et al., 2008). The longest spindles belong to persons with an average, or approximately average, 10 Hz individual alpha peak

![FIGURE 5. Results of regression analysis of the relationship between alpha spindle length and individual alpha peak frequency in low (LAF n = 48) and high (HAF n = 48) alpha frequency male subjects.](image)
frequency. Multiple regression analysis showed that spindle lifespan is positively related to individual alpha peak frequency in LAF subjects and negatively related in HAF subjects (Figure 5; Bazanova et al., 2008). These data suggest the different neuronal mechanism for spindle formation between LAF and HAF subjects. The membrane bistability might play an important role in different patterns of spindles displayed by thalamocortical neurons. Actually, as shown by Fuentealba and colleagues, the reticular neurons display membrane bistability, as indicated by two discrete electrical potential modes, with differential responsiveness to cortical inputs (Fuentealba, Timofeev, Bazhenov, Sejnowski, & Steriade, 2005). In addition, in vivo (Steriade & Llinas, 1988; Steriade & Timofeev, 2003) and in vitro (Bal & McCormick, 1996) intracellular studies have revealed at least two different patterns during spontaneously occurring spindles, which may be related to the actions exerted by nonbistable and bistable neurons, respectively. Indeed, nonbistable neurons fired stronger bursts with higher intraburst frequencies, which are assumed to generate IPSPs, ~7–10 Hz. By contrast, IPSPs with lower amplitudes and higher frequency are likely to be mainly generated by single action potentials, as they occur during the depolarizing plateau in bistable cells (Fuentealba et al., 2005). If we assume that longer spindles of stable brain activity imply less information to process (as reflected by a higher stability of the brain generator), whereas shorter segments imply a higher number of brain microstates caused by more different steps of information processing, then it is possible to suggest that the intrasegment alpha amplitude variability could be indexing a phasic event (Oprisan, Prinz, & Canavier, 2004). Probably, alpha-bursts are associated with a brain microstate change (such as sleep spindles), as demonstrated by its phasic intrusion in a desynchronized background of brain activity (Cantero, Atienza, Salas, & Gomez, 1999). It is important to know that intraspinde amplitude variability decreases in coma or stupor (Brenner, 2005) but increases during cognitive loading (Kaplan & Borisov, 2003) and in relation to the ability for self-control in neurofeedback training (18–20 sessions; Bazanova et al., 2008). Therefore, amplitude variability, which is associated with phase resetting intensity (Oprisan et al., 2004), reflects engagement of cognitive control mechanisms (Hanlsmayr et al., 2007; Lebedev, 1994; Livano & Dumenko, 1987). Moreover, in children, the phase resetting index and intrasegment amplitude variability generally increases as a function of age (Bazanova, 2008; Thatcher et al., 2008). This probably reflects the well-known fact that ability for self-control develops with age (Mischel, 2004; Orekhova, Stroganova, Posikera, & Malikh, 2003). Therefore, alpha activity indices, including individual alpha peak frequency, amount of alpha suppression and individual alpha bandwidth, and segmental parameters of alpha spindle characterize the degree of cortical ability for activation, neuronal plasticity, excitability, and inhibition. In this way these parameters could characterize self-regulation ability.

According to Klimesch’s “time inhibition” theory, the active role of alpha waves is seen in a mechanism that may also underlie the functional role of other oscillations (Klimesch et al., 2007), whereas synchronization in the alpha frequency range helps neurons in distributed networks to effectively activate common target cells (Basar, 2006; Klimesch et al., 2007). This alpha-frequency dependent mechanism plays an important role in the top-down control of cortical activation. Mechanisms giving rise to alpha amplitude desynchronization also reflect the dependence of brain activation on the alpha frequency range. Klimesch has proposed that upper-frequency alpha oscillations are related to top-down processes in a complex sensory-motor system that controls the access to and manipulation of stored information (Klimesch et al., 2007). For example, when a task requires that certain types of processes be performed with stored information (e.g., stored information must be kept in mind, must not be retrieved or must be manipulated such as with highly skilled musicians during musical performance), individual alpha-activity increases. As with overall brain activity,
top-down control is not a unitary phenomenon. Recently, Ben-Simon and associates (Ben-Simon, Podlipsky, Arieli, Zhdanov, & Hendler, 2008) combined fMRI with EEG in a study that proposed two parallel patterns of alpha modulations and underpin their anatomical basis in the human brain. These findings suggest that the human alpha rhythm represents at least two simultaneously occurring processes that characterize the “resting brain”; one is related to expected change in sensory information, whereas the other is endogenous and independent of stimulus change. Although twin studies have long shown that heritability of EEG oscillations is substantial (Enoch et al., 2008; Smit et al., 2006), very little is known about the genes underlying distinct EEG traits. Early linkage analyses identified a genetic locus on the distal part of chromosome 20q to modulate alpha power which authors named “alpha-activity” (Anokhin et al., 1992). It is likely, however, that multiple genes contribute to “alpha phenotypes,” and a few candidate genes were indeed found to affect alpha oscillations. Recent genetic polymorphism studies indicated that the gene on chromosome 5q13-14 of corticotrophin releasing hormone-binding protein (CRH-BP) modulates alpha power in isolated Plains American Indians and Caucasians (Enoch et al., 2008). Moreover, a functional variation in exon 7 of the gene on chromosome 6 encoding the human GABA B receptor (GABABR1) also influences EEG voltage in the alpha range (Winterer et al., 2003). Finally, more recent data of Bodenmann et al. (2009) demonstrate that mechanisms involving COMT contribute to interindividual differences in alpha oscillation frequency, which are functionally related to executive performance (Bodenmann et al., 2009). Recent data by Aftanas et al. (2008) proposed that LAF and HAF subjects have different mechanisms for the modulatory influences of the balance of reinforcement system activities on autonomic vascular regulatory processes. Thus, the data just presented emphasize the genetically different EEG patterns and associated behavioral strategies dependent on alpha frequency.

In conclusion, studying the alpha EEG activity should not be limited only to alpha amplitude intra- and interindividual variability but should also include two other important physical characteristics: frequency and phase of alpha oscillations. Moreover, the evidence that alpha oscillations play an active role in cognitive processing and self-regulation is strongly related with the genotypic and phenotypic difference between low and high frequencies of the alpha activity. Moreover, the neuronal activation strategies for increasing IAPF as a result of biofeedback training are different according to the individual alpha frequency. Several factors may be common in the generation of different types of oscillations. Nonetheless, the exact mechanisms for generating an oscillation may differ widely between different frequency waves such as low- and high-frequency alpha, depending on individual network properties, cell types, cell physiology, hormone level, blood feeding, and so on. The role of such biological factors as immunological, neurohumoral, and activity of heart-vessel and breathing systems in the generation and formation alpha oscillations is still unknown. Thus, for a better understanding of the functional role of alpha activity indices in peak performance achievement through Biofeedback or BCI technologies, the investigation of cortico-visceral interplay in conjunction with the inherited alpha frequency EEG types may be of crucial importance.

REFERENCES


overview and an example of peak alpha frequency training for cognitive enhancement in the elderly. *Clinical Neuropsychologist*, 21, 110–129.


Hanslmayr, S., Klimesch, W., Sauseng, P., Gruber, W., Doppelmayr, M., Freunberger, R., ... Birbaumer, N. (2007). Alpha phase
reset contributes to the generation of ERPs. *Cerebral Cortex, 17*, 1–8.


