# CMOS Ultralow Power Brain Signal Acquisition Front-Ends: Design and Human Testing

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Abstract—Two brain signal acquisition (BSA) front-ends incorporating two CMOS ultralow power, low-noise amplifier arrays and serializers operating in MOSFET weak inversion region are presented. To boost the amplifier's gain for a given current budget, cross-coupled-pair active load topology is used in the first stages of these two amplifiers. These two BSA front-ends are fabricated in 130 and 180 nm CMOS processes, occupying 5.45 mm<sup>2</sup> and 0.352 mm<sup>2</sup> of die areas, respectively (excluding pad rings). The CMOS 130-nm amplifier array is comprised of 64 elements, where each amplifier element consumes 0.216  $\mu$ W from 0.4 V supply, has input-referred noise voltage (IRNoise) of 2.19  $\mu V_{RMS}$  corresponding to a power efficiency factor (PEF) of 11.7, and occupies 0.044 mm<sup>2</sup> of die area. The CMOS 180 nm amplifier array employs 4 elements, where each element consumes 0.69  $\mu$ W from 0.6 V supply with IRNoise of 2.3  $\mu V_{RMS}$  (corresponding to a PEF of 31.3) and 0.051 mm<sup>2</sup> of die area. Noninvasive electroencephalographic and invasive electrocorticographic signals were recorded real time directly on able-bodied human subjects, showing feasibility of using these analog front-ends for future fully implantable BSA and brain- computer interface systems.

*Index Terms*—Analog front-end (AFE), CMOS, electrocorticography (ECoG), electroencephalogram (EEG), instrumentation amplifier (InAmp), noise efficiency factor (NEF), operational transconductance amplifier (OTA), power efficiency factor (PEF), ultra-low power (ULP), weak inversion (WI) region.

#### I. INTRODUCTION

T IS estimated that every year there are  $\sim$ 500,000 new cases of spinal cord injury (SCI) worldwide [1]. This condition

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substantially decreases independence and quality of life of those affected, and the resulting disability and comorbidities pose a significant economic burden on the individual as well as on society. Since there are no satisfactory means to restore motor function after SCI, novel approaches to address this problem are needed. Bypassing the damaged spinal cord by means of a brain-computer interface (BCI), which enables direct brain control of prostheses, constitutes one such approach. Non-invasive electroencephalogram (EEG)-based BCIs have the capacity to restore basic ambulation after SCI [2], [3], although their applicability is limited by the low information content (i.e., limited bandwidth and low spatial resolution) of EEG signals. Invasive BCIs, on the other hand, have enabled control of multi-degreeof-freedom robotic prostheses [4]. However, they utilize bulky and power-hungry general-purpose recording hardware, and rely on skull-protruding electronic components. Furthermore, these systems typically employ intracortically implanted microelectrode arrays, which can trigger foreign body responses such as inflammation and scarring, ultimately leading to failure of the system within months to few years [5]. These factors represent a serious obstacle to a widespread adoption of invasive BCI technology.

These problems may be addressed by developing a fully implantable BCI system that uses highly stable electrocorticogram (ECoG) signals [6]. Such a BCI system is envisioned to consist of ECoG electrodes, amplifiers, a processor, and a wireless module to control and communicate with output devices (e.g., prostheses), all implemented in a miniaturized form factor and operating in a low-power regime in order to facilitate permanent implantation. Since ECoG electrodes are placed above the arachnoid layer without breaching the neuronal tissue, ECoG signals have long-term stability [6], [7], while providing the spatiotemporal resolution necessary for high-performance BCI applications [8], [9]. In particular, studies have shown that the ECoG high- $\gamma$  frequency band (70–120 Hz) exhibits spatially localized amplitude modulation that is correlated with individual's physical movements [10], and this feature has been used to decode arm [11] and finger movements [12]. Chronic in vivo recording of ECoG signals has been used for neurological treatment. The Medtronic Activa PC+S system [13], [14], was used in patients having Parkinson disease with ECoG electrode strips implanted over the motor cortex and depth electrodes in the subthalamic nucleus [15]. Chronic recordings from these areas

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were used to study the association between gamma band oscillations and dyskinesia. The Activa PC+S system was also used for recording signals from ECoG electrode strips over the motor cortex of a patient with amyotrophic lateral sclerosis (ALS) to facilitate BCI-control of a virtual keyboard [16]. Finally, as shown in prior art, a fully implantable system eliminates the need for bulky skull-protruding components, often employed in the state-of-the-art invasive BCIs, as well as bulky recording hardware and external computers.

There has been extensive research on low-power amplifier and amplifier array designs for neural signal sensing applications, which vary substantially in frequency and dynamic range. For example, in [17], the authors present a folded-cascode operational transconductance amplifier (OTA) using current-splitting and current-scaling techniques with a cascaded 6th-order bandpass filter for detecting epileptic fast ripples between 250 and 500 Hz. The stack of 4 transistors and large degeneration resistors in this design increase the required supply voltage to accommodate sufficient output voltage headroom. In [18], a closed-loop neural recording amplifier has been developed that utilizes a T-network in its feedback path in order to achieve high input impedance and common-mode rejection ratio (CMRR) within a small chip area. The authors argued that the T-network in the feedback path is useful when the routing area overhead, crosstalk and input-referred noise (IRNoise) do not dominate the performance [18].

Most of the previously developed neural sensing amplifiers focus on EEG or single-unit recordings. Consequently, their designs are not optimal for use in other recording modalities, such as ECoG. Moreover, a few studies that exist with analog front-end (AFE) designs for ECoG recording lack in vivo experimental validation in humans. For example, [19] presents a 32-channel integrated circuit (IC) for ECoG recording, followed by in vivo measurements in a rat. The power consumption of this system is too high, making it unsuitable for human ECoG signal acquisition, especially in a fully implantable form. In [20], an ECoG/EEG IC has been introduced which records signals in 4 different sub-bands as opposed to simultaneously capturing the complete ECoG spectrum. This IC has been validated by comparing the measurements of a pre-recorded human ECoG signal with those generated by a model of the signal acquisition chain. This approach, however, does not accurately capture the IC's interface with the body (e.g., 60 Hz noise), which may significantly affect the performance. When tested in an awake monkey, the signals simultaneously measured by this IC and those acquired using a commercial system showed only modest correlations in  $\alpha$  (8–12 Hz) and high- $\gamma$  (70–120 Hz) bands. This can be explained by the dominating effects of flicker and thermal noise at these frequencies. On the other hand, the signals in the  $\beta$ (13-35 Hz) and low- $\gamma$  (35–70 Hz) bands were only qualitatively compared with no correlation coefficients reported. Recent work [21] reports on an AFE consuming 1.08  $\mu$ W of power per channel, which is achieved by narrowing the AFE bandwidth and filtering out the noise. A potential problem with this approach may lie in the high sensitivity of the designed  $G_m C$  filters to process variation. The proposed AFE has been tested in its ability to reproduce pre-recorded ECoG data and acquire ECoG signals

in vivo from an anesthetized monkey. However, human testing and direct comparison of recorded signals to those acquired with a commercial-grade system have not been performed. Finally, the work in [22] presents a 64-channel wireless micro ECoG recording system with the front-end achieving a powerefficiency factor (PEF) that is  $3 \times$  smaller than the state-of-theart. In vivo measurements from an anesthetized rodent show the power increase with respect to the pre-sedation state in  $\delta$  (1–4 Hz) and  $\theta$  (4–7 Hz) bands, but very little change in BCI-relevant frequency bands. Furthermore, none of the above systems were tested in a hospital environment, which is typically characterized by extremely hostile ambient noise and interference conditions. In summary, while the development of these architectures has been inspired by human BCI applications, their in vivo testing in humans and comparison to conventionally acquired ECoG signals are conspicuously missing.

This paper presents the design, experimental validation, and comparative study of two CMOS ultra-low power (ULP) amplifier array and serialization circuitries that constitute core building blocks of two brain signal acquisition (BSA) front-ends. These BSA front-ends can act as the basis for a future, fully implantable ECoG-based BCI system (Fig. 1(a)). The AFE IC will be housed within an enclosure, called the skull unit, to be surgically implanted into the skull [23]. Other building blocks required to develop a complete ECoG-based BCI, e.g., transceiver, power management unit and digital signal processor are intended to be placed in another unit away from the patient's brain. This approach imposes less health hazards for the patient as well as more practical system specifications for a portable, user-friendly solution. All circuits in this work are designed to operate in the weak inversion (WI) region to maximize power efficiency and minimize heat dissipation, while maintaining high gain and low noise operation. In vivo human measurements and objective validation against a commercial bioamplifier are done in (1) a human subject using non-invasive EEG cap, and (2) a human subject with subdurally implanted high-density ECoG grid.

The paper is organized as follows. Section II presents the proposed AFE for recording ECoG signals and identifies the criteria and required specifications of the building blocks for designing the system. Sections III and IV discuss the design and implementation of the two BSA front-ends, BSA I and BSA II, respectively. Section V illustrates the electrical and neural measurement results of both front-ends. Finally, Section VI presents concluding remarks and potential extensions of this work.

### **II. PROPOSED SYSTEM ARCHITECTURE**

Responsible for sensing and amplification of microvolt-level brain signals, the amplifier array IC is a critical building block of a BSA front-end. To be employed as a fully implantable device, the signal acquisition front-end should be small in size and consume micro-watt level of power. The system-level diagram of the proposed AFE is shown in Fig. 1(b) [24]. The AFE IC includes fully differential amplifiers, a serializer, and an output buffer all biased in the WI region. The outputs of the array are multiplexed in time to better facilitate input-output



Fig. 1. Proposed AFE: (a) A cross-sectional view of the envisioned fully implantable BSA circuit, enclosed within a skull unit module. The BSA circuit is connected to a subdurally implanted high-density (HD)-ECoG electrode grid that senses brain signals. (b) Block diagram of the structure showing the brain interface electrodes with their corresponding impedances and BSA comprised of an array of fully differential amplifiers, serializer and buffer.

cable management by reducing the number of wires. The non-overlapping clock generator within the serializer generates N-phase clock signals, each with 1/N duty cycle. Nonoverlapping clock signals ensure that only one amplifier is connected to the output buffer at a time during the channel switchover. This work presents two ULP BSA front-ends, BSA I and BSA II. BSA I provides symmetrical and complementary signal amplification paths to achieve energy-efficient low noise signal conditioning. BSA II is designed to achieve a high CMRR (i.e., better than 70 dB), thereby reducing the detrimental effect of power-line 60 Hz interference on the recorded signal.

Minimizing both noise and power dissipation imposes stringent design trade-off in an AFE for an implantable system, mandating meticulous considerations at every level of the design process. For example, at the device level, this notion implies that transistors should be designed to operate in a region which yields minimum power consumption for a given IRNoise imposed by minimum detectable ECoG signal power.

It is well-known that the MOS transistors in the WI region achieve maximum  $g_m/I_{\rm DS}$ -ratio, resulting in the highest power efficiency at the cost of lower operation maximum bandwidth [25], [26]. Fig. 2 demonstrates  $g_m/I_{\rm DS}$  and  $log_{10}(I_{\rm DS})$  variations with respect to  $V_{\rm GS}$  for the two technologies given the same transistor sizes and bias conditions. Referring to Fig. 2(a), a higher subthreshold leakage current and a higher slope are observed in the weak-inversion region for the 130 nm process compared to the 180 nm process. A higher slope corresponds to a larger  $g_m$  for the same bias current. This feature translates to a better power efficiency (Fig. 2(b)) and noise performance for transistors designed in this specific 130 nm process. It is noteworthy that the  $g_m/I_{\rm DS}$ -plot for the 130 nm CMOS process does not show the expected flat region in the deep subthreshold region. This is because BSIM4 device model was adopted for this process by the foundry. On the other hand, the 180 nm



Fig. 2. (a) Drain-source current  $(I_{\rm DS})$  vs. gate-source voltage  $(V_{\rm GS})$  for the two technologies. (b)  $g_m/I_{\rm DS}$  vs.  $V_{\rm GS}$  for the two technologies. W/L = 20  $\mu$ m/2  $\mu$ m with 10 fingers,  $V_{\rm DS} = 1$ V for both transistors and body temperature of 37 is considered for simulation.

process employed PSP device model, which can predict the device behavior in deep subthreshold region more accurately.

ECoG signals typically have an amplitude of around 50–100  $\mu$ V [27], with  $\beta$  and high- $\gamma$  bands typically providing the most informative features for BCI applications [10]. The IRNoise of the AFE should be kept lower than the noise floor of the ECoG electrodes. Our recorded measurements using a commercial BCI signal acquisition equipment showed that the RMS noise floor, integrated over a frequency range of 8–200 Hz, is typically less than 10  $\mu$ V<sub>RMS</sub>, which is in compliance with the data reported in literature, e.g., [28]. Low noise operation is of particular interest for high- $\gamma$  band, because the ECoG signal power becomes weaker with frequency [29].

The CMRR and power-supply rejection ratios (PSRR) should be large to attenuate the effect of environmental noise sources (e.g., 60 Hz power-line noise). Assuming an IRNoise level of 2  $\mu$ V<sub>RMS</sub> in the presence of common-mode interference with 10 mV<sub>RMS</sub>, a nominal 34 dB attenuation (i.e., 74 dB CMRR) is needed so that the output noise and interference voltage magnitudes are equal. In addition, the amplifier should show a high input impedance to lower the effect of common-mode



Fig. 3. Complementary input structure of the OTA used in Amplifier I.

interference. This attribute is especially important for multichannel recordings since the impedance mismatch between electrodes  $(Z_{e,1}, ..., Z_{e,N})$  as well as the mismatch between the impedance seen from the common reference input (parallel combination of  $Z_{in,1}, ..., Z_{in,N}$  in Fig. 1) and  $Z_{in,k}$   $(1 \le k \le N)$  reduces the overall CMRR. Subdural electrodes' impedance have been reported (as well as measured) to be about 1 k $\Omega$  [6], thus the input impedance at the frequency of interest should be  $\gg$ 1 M $\Omega$  [30]. Moreover, large DC offset associated with neural recording electrodes should be eliminated so as to minimize distortion or avoid saturation of the amplifier. Furthermore, electrical shielding and DC isolation are needed between the IC and implanted electrodes. Finally, the crosstalk in a multi-channel system should be mitigated to avoid contamination of the overall information recorded from different channels.

# III. BSA I: AN ARRAY OF 64 AMPLIFIER I CIRCUITS AND A SERIALIZER

BSA I incorporates 64 units of Amplifier I and a serializer, as shown in Fig. 1(b). Fig. 3 shows the general block diagram of the OTA used in the Amplifier I, composed of complementary NMOS-PMOS input stage. Intuitively, the signal is amplified by the transconductance gain of the input transistor pairs and subsequently applied to the current gain stage in each of the top and bottom branches  $(A_{I,N} \text{ and } A_{I,P})$ . Upon flowing through the load impedance  $Z_L$ , the summing current will generate the output voltage. Fig. 4(a) shows the top-level topology of Amplifier I employing an OTA with an RC feedback network. The AC-coupled input provides DC rejection between the recording electrodes and the OTA input, thus providing a layer of electrical safety and isolation between the patient's brain and the amplifier. Fig. 4(b) depicts the transistor-level schematic of the OTA utilized in Amplifier I, including common-mode feedback (CMFB) circuitry (in gray) [23]. The OTA device sizes and aspect-ratios together with operating points of the individual devices are presented in Table I. NMOS and PMOS transistors' body connections are tied to the ground and supply rails, respectively. The minimum headroom for a single transistor biased in the WI region is  $\sim 4U_T$  (where  $U_T \approx 26$  mV at room temperature) [31]. As a result, the OTA is biased at 0.4 V supply



Fig. 4. (a) Amplifier I comprising a closed-loop amplifier with capacitive feedback and its die microphotograph, and (b) the schematic of the complementary NMOS-PMOS OTA [23].

TABLE I Amplifier I Device Sizes and Operating Points

Devices	W/L ( $\mu$ m/ $\mu$ m)	$I_D$ (nA)	$g_m / I_D (V^{-1})$		
$M_{1a} - M_{1b}$	53.5/1.35	130	34		
$M_{2a} - M_{2b}$	50/10	65	25		
$M_{3a} - M_{3b}$	50/10	65	25		
$M_{4a} - M_{4b}$	3.7/32	1.5	25		
$M_{5a} - M_{5b}$	140/1.2	138	29		
$M_{6a} - M_{6b}$	15/30	69	28		
$M_{7a} - M_{7b}$	15/30	69	28		
$M_{8a} - M_{8b}$	0.4/40	1.5	26		
$M_9$	80/0.36	89	34		
$M_{10}$	80/0.36	170	35		
$M_{11}$	120/0.13	277	25		

to mitigate large process variations resulting from WI operation, while achieving low power and low noise. The first stage employs a complementary NMOS-PMOS differential configuration with a complementary active load comprising parallel combination of diode-connected transistors and a cross-coupled pair [23], [32]. Cross-coupled pair and diode-connected transistors are identically sized as shown in Table I, thereby having the same transconductance. The effective output resistance of the input stage is thus increased from  $\frac{r_{o3}}{1+g_{m3}r_{o3}}||r_{o1}|$  (in the absence of cross-coupled pair load) to  $r_{o1}||r_{o2}||r_{o3}$ , where  $r_{o1}$ ,  $r_{o2}$  and  $r_{o3}$  are the drain resistance of  $M_1$ ,  $M_2$  and  $M_3$ , respectively. The active-load devices are sized in a way that no instability or latch-up happens due to the process variation. The size of output transistors  $M_4$  and  $M_8$  are chosen to exhibit large drain resistance and low current consumption at the output stage.

The capacitance ratio  $C_1/C_2$  ( $C_1 = 20$  pF and  $C_2 = 200$  fF) defines the closed-loop gain with high accuracy so long as the open-loop gain is sufficiently high. High output impedance of the OTA imposes a high impedance load for the feedback and next stage circuits. Pseudo-resistors realized by transistors  $M_A$ and  $M_B$  (as in [33]) provide large equivalent resistance R of few  $G\Omega$ , self-bias the input stage of the OTA without consuming any additional power for closing the loop, and set the lower 3-dB cutoff frequency ( $f_L = (2\pi R C_2)^{-1}$ ). Compared to pseudo-resistor used in [34], this implementation provides a wider linear range of operation. Assuming all transistors are identically matched (to simplify the analysis), the IRNoise power spectral density (PSD) of the open-loop OTA ( $\overline{V_{in,OTA}^2}$ ) is calculated to be:

$$\overline{V_{\text{in,OTA}}^2(f)} = \frac{4kT\gamma}{g_{m1}} \left(1 + \frac{2g_{m2}}{g_{m1}}\right) + \frac{K_{p,1/f}}{C_{ox}(WL)_1} \times \frac{1}{f} \left[1 + 2\frac{K_{n,1/f}}{K_{p,1/f}} \frac{(WL)_1}{(WL)_2} \left(\frac{g_{m2}}{g_{m1}}\right)^2\right] \quad (1)$$

where k is the Boltzmann constant,  $\gamma$ ,  $K_{p,1/f}$  and  $K_{n,1/f}$  are technology-dependent parameters, f is frequency,  $C_{ox}$  is the gate oxide capacitor, and T is the temperature.  $\gamma$ , the excess thermal noise factor, is slightly lower in the WI region than in the strong inversion (SI) [25]. Note that the complementary structure used in this OTA doubles the overall  $G_m$ . Flicker noise and mismatch effects are slightly attenuated by large input transistors and symmetrical circuit layout. In addition, dynamic compensation techniques such as chopper stabilization and autozeroing are commonly used to reduce the effect of amplifier offset and flicker noise [35]-[39]. However, these techniques require switches with low on-resistance to accommodate highly linear operation for autozeroing techniques and low residual input-referred offset voltage for chopping techniques. Thus, a high-swing on-chip clock needs to be generated at the expense of high power consumption. Therefore, we have not used these compensation techniques in the current design. The IRNoise of Amplifier I in Fig. 4(a),  $\overline{V_{in,tot}^2}$ , is calculated to be:

$$\overline{V_{\text{in,tot}}^2(f)} = \left(4kTR + \overline{V_{\text{in,OTA}}^2(f)}\right) \left(\frac{f_L}{G_c f}\right)^2 + \overline{V_{\text{in,OTA}}^2(f)} \left(\frac{C_1 + C_2 + C_{in}}{C_1}\right)^2$$
(2)

where  $G_c$  is the midband closed-loop gain defined by  $C_1/C_2$ and  $C_{in}$  is the equivalent input capacitance seen from the input of the OTA.

Sizing of the input transistors is critical due to existing tradeoff between  $\overline{V_{in,OTA}^2}$  and  $\overline{V_{in,tot}^2}$ . More precisely, large input transistors with low flicker noise will reduce  $\overline{V_{in,OTA}^2}$ . On the other hand, a larger device size leads to larger input capacitance,  $C_{in}$ , which adversely affects the system sensitivity. Another



Fig. 5. BSA II: (a) overall topology, including 4 Amplifier II circuits and one InAmp (b) InAmp implementation.

point to consider is that  $C_{\rm in}$  shunts the gate of the input transistor to ground, causing a capacitive voltage division between  $C_1$ ,  $C_2$  and  $C_{\rm in}$ . This, in turn, lowers the differential loop-gain, thereby preventing the closed-loop gain to be accurately defined. Moreover, as  $f_L$  decreases, the thermal noise contribution of the pseudo-resistors to  $\overline{V_{\rm in,tot}^2}$  is reduced, while the flicker noise contribution of the OTA to  $\overline{V_{\rm in,tot}^2}$  is increased.

The serializer in Fig. 1(b) is clocked at 64 kHz and is composed of a custom-designed 6-bit synchronous binary counter, a 6-to-64 decoder and  $2 \times 64$  complementary pass-gate switches for selecting the amplifier channels. A reset signal puts the circuit in an initial state (channel 64) and the clock signal selects the channels sequentially [23].

Section V presents the measurement results of the BSA I, which was fabricated in a 130 nm CMOS process [23].

# IV. BSA II: AN ARRAY OF 4 AMPLIFIER II CIRCUITS, A SERIALIZER, AND AN INSTRUMENTATION AMPLIFIER

The existence of two signal paths in Amplifier I leads to a degradation in CMRR ( $\approx 60 \text{ dB}$ ). To further elaborate, suppose that the only existing mismatch is the one between each of the input pairs in Fig. 3 (i.e.,  $\Delta g_{m_N}$  and  $\Delta g_{m_P}$ ). This mismatch directly contributes to the common-mode to differential-mode gain for Amplifier I, which is derived as follows:

$$A_{\rm cm-dm} \approx \frac{\Delta g_{m_N} \times A_{I_P} Z_{\rm out}}{(g_{m_{N1}} + g_{m_{N2}}) Z_{S_N}} + \frac{\Delta g_{m_P} \times A_{I_N} Z_{\rm out}}{(g_{m_{P1}} + g_{m_{P2}}) Z_{S_P}}$$
(3)

where  $Z_{out}$ ,  $Z_{S_N}$  and  $Z_{S_P}$  are output impedances of Amplifier I,  $I_{S_N}$  and  $I_{S_P}$ , respectively. It is inferred from (3) that the CMRR of Amplifier I can statistically be degraded by a factor of 2 compared to an amplifier with a single path from the input to the output. A high CMRR is important in brain signal amplifiers due to the presence of a strong 60 Hz power-line noise in the amplification band. If not eliminated, major degradation in the output signal-to-noise ratio (SNR) will be seen. To further improve this feature, Fig. 5 introduces the block diagram of BSA II, which is composed of an array of 4 Amplifier II circuits, a serializer, and an instrumentation amplifier (InAmp). Similar to Amplifier I, Amplifier II is realized as a fully differential RC feedback circuit incorporating 200 fF feedback and 18 pF input



Fig. 6. (a) OTA schematic used in Amplifier II. (b) Op-amp schematic used in InAmp.

AC-coupled capacitors. Matching accuracy of the feedback capacitor limits the achievable CMRR. For instance, it is readily shown that for closed-loop gain of 100 and 10% mismatch of the feedback capacitor, CMRR is lower than 60 dB. The open-loop OTA within Amplifier II employs PMOS input differential-pair with NMOS cross-coupled active loads, as shown in Fig. 6(a). Having one signal path from the input to the output relaxes the mismatch considerations present in complementary signal paths used in the first design.

The IRNoise of the open-loop OTA is calculated to be:

$$\overline{V_{\text{in,OTA}}^2(f)} = \frac{8kT\gamma}{g_{m1}} \left(1 + \frac{2g_{m2}}{g_{m1}}\right) + \frac{2K_{p,1/f}}{C_{ox}(WL)_1} \times \frac{1}{f} \left[1 + 2\frac{K_{n,1/f}}{K_{p,1/f}}\frac{(WL)_1}{(WL)_2} \left(\frac{g_{m2}}{g_{m1}}\right)^2\right]$$
(4)

Assuming a single-pole frequency response, it is readily proven that the noise efficiency factor (NEF [40]) reaches a lower-limit of  $2\sqrt{n\gamma}$  (where *n* denotes the subthreshold slope factor [31]) for both OTAs used in Amplifiers I and II if no dynamic compensation techniques are employed. The use of the same closed-loop architecture as in Amplifier I indicates that the IRNoise of Amplifier II is also expressed by (2).

The InAmp, after the serializer, provides further amplification and buffering to the output. It is commonly known that isolated resistive feedback circuitry ( $R_1$  and  $R_2$ ) provides flexibility in the design of an InAmp and its constituent open-loop op-amps with no concern of loading on preceding circuits [41]. In addition, any variation in  $R_1$  is widely known to only contribute to the differential gain variation and will not increase commonmode to differential-mode gain ( $A_{cm-dm}$ ) [42], [43]. Therefore, the CMRR is not degraded. As for the contribution of the mismatch between the  $R_2$  resistors ( $R_{2\Delta}=R_2 + \Delta R$ ) on CMRR, the InAmp's  $A_{cm-dm}$  induced by this mismatch is derived first:

$$A_{\rm cm-dm} = \frac{E_{\rm CM} - 1}{1 + \frac{R_{2\Delta}}{R_1 A_{\rm dm2}} - \frac{R_{2\Delta}}{A_{\rm dm1} R_1} E_{\rm CM} + \frac{1}{A_{\rm dm2}}}$$
(5)

where:

$$E_{\rm CM} = \frac{1 + \frac{1}{A_{\rm dm2}} \left(1 + \frac{R_2 + R_{2\Delta}}{R_1}\right)}{1 + \frac{1}{A_{\rm dm1}} \left(1 + \frac{R_2 + R_{2\Delta}}{R_1}\right)}$$
(6)



Devices	W/L ( $\mu$ m/ $\mu$ m)	$I_D$ (nA)	$g_m / I_D (V^{-1})$
$M_{1a} - M_{1b}$	152/0.18	510	27
$M_{2a} - M_{2b}$	12.8/20	255	20.7
$M_{3a} - M_{3b}$	12.8/20	255	20.7
$M_{4a}$ - $M_{4b}$	0.8/25	27	16.6
$M_{5a} - M_{5b}$	16/12	67	22.9
$M_6$	192/1	1020	27

In deriving (5), the open-loop gains of the op-amps,  $A_{dm1}$  and  $A_{dm2}$ , are assumed to be finite, while each op-amp exhibits negligible differential to common-mode gain. The common-mode gain  $A_{cm}$  of the InAmp is almost unity. If followed by a high-CMRR amplification stage, the contribution of  $A_{cm}$  on CMRR will be negligible. On the other hand, to reduce the impact of  $A_{cm-dm}$  on CMRR, the op-amps need to exhibit large open-loop gain. Large open-loop gain significantly reduces the contribution of  $R_2$  mismatch on the CMRR. Ideally, if the op-amps are perfectly matched ( $A_{dm1} = A_{dm2}$ ),  $A_{cm-dm}$  would become zero regardless of  $\Delta R$  value.

Amplifier II and the InAmp are DC-coupled, eliminating the need for large coupling capacitors. Considering a 39-dB gain for the OTA, the expected differential input amplitude of the InAmp is less than 9 mV, which falls within the input common-mode range of InAmp (0 to  $V_{\rm DD} - 2V_{\rm DS,sat}$  where  $V_{\rm DS,sat}$  is the drain-source saturation voltage).

Fig. 6(a) and (b) show the transistor-level schematics of the OTA used in Amplifier II and the op-amp used in InAmp, respectively. Both amplifiers use similar topology while the devices are sized according to the performance specs needed from each circuit, namely, low noise and high transconductance for the OTA (high driving power and high voltage gain for the op-amp). Tables II and III show device sizes and operating points for the OTA and the op-amp, respectively. All transistors are biased in WI to maximize power efficiency. To achieve a maximum ECoG bandwidth of 200 Hz in the OTA and avoid out-of-band noise accumulation, a large 48 pF capacitor  $C_L$  is placed differentially at the output. The input transistors operate in deep WI to maximize their  $g_m/I_D$ -ratio so as to reduce the IRNoise

TABLE III Op-Amp Device Sizes and Operating Points

Devices	W/L ( $\mu$ m/ $\mu$ m)	$I_D$ (nA)	$g_m/I_D (V^{-1})$		
$M_{1a} - M_{1b}$	100/0.18	80	29		
$M_{2a} - M_{2b}$	1/5	40	21		
$M_{3a} - M_{3b}$	1/5	40	21		
$M_{4a} - M_{4b}$	8/5	320	21		
$M_{5a} - M_{5b}$	32/4	320	21		
$M_6$	200/1	160	28		



Fig. 7. CMFB circuit used in Amplifier II.

contributions of active-load devices  $(M_{2a} - M_{2b})$  and  $M_{3a} - M_{3b}$ ). PMOS transistors are used in the input differential pair to have a lower flicker noise. Furthermore, the use of a PMOS input pair for the op-amp makes common-mode levels of the OTA output and the op-amp input compatible, thereby making it possible to DC-couple the two. DC-coupling eliminates the need for large decoupling capacitors as well as biasing circuitry of the op-amp inputs. The OTA bandwidth and stability are determined by its output stage where the dominant pole is located. On the other hand, the op-amp's dominant pole is located at its first stage's output node, as its output stage should provide high current drive capability. The op-amp is thus Miller-compensated and its bandwidth is chosen to be  $\approx 800$  Hz in order to accommodate 4 recording channels.

Fig. 7 shows the proposed CMFB circuit to set the output common-mode voltage of the OTA. The drain currents of transistors  $M_{4a}$  and  $M_{4b}$  are steered to ground or to transistor  $M_6$ depending on common-mode level of  $V_{in}$ .  $M_{3a}$  and  $M_{3b}$  mirror  $M_6$ , sinking current from the OTA's output stage, thereby adjusting the OTA common-mode level. Note that the input and the output of the CMFB are physically connected together. The CMFB output currents are expressed as:

$$I_{3a} = I_{3b} = \frac{W_3}{W_6} \times g_{m1,2} \times \left(\frac{V_{in+} + V_{in-}}{2} - V_{\text{ref}}\right)$$
(7)

where  $g_{m1,2}$  denotes the transconductance of  $M_{1a} - M_{1b}$  and  $M_{2a} - M_{2b}$ ,  $W_3$  is channel width of  $M_{3a} - M_{3b}$ , and  $W_6$  is  $M_6$  width. Input transistors  $(M_{1a} - M_{1b}$  and  $M_{2a} - M_{2b})$  should remain in saturation region for proper operation of the CMFB. Having few millivolts swing at the OTA's output ensures that no transistor leaves saturation. Transistors  $M_{3a}$  and  $M_{3b}$  are designed to have long channel length, with negligible loading effect on the OTA. Their parasitic capacitances are absorbed in



Fig. 8. Non-overlapping clock signals applied to serializer switches.

the OTA's load capacitor. The simulation of the CMFB shows a current consumption of 24 nA and a common-mode phase margin of at least 35°.

The circuitry for the serializer logic used in BSA II is presented in Fig. 8. This logic contrives i) non-overlapping clock signals for time-multiplexing, and ii) a gray-coding scheme for a 2-bit binary counter to eliminate race conditions. The serializer clock signal's duty-cycle produces temporal spacing between clocks applied to the serializer switches (Fig. 8). A Gray-code converter is used to convert binary code to Gray code such that the counter exhibits no race condition, which could otherwise result in sparks in the 2-to-4 decoder in Fig. 8. A T-network switch is used for channel selection in this serializer to provide large input-output isolation and minimize the effects of charge-injection and clock-feedthrough.

## V. MEASUREMENTS

The functionality of BSAs I and II was verified by electrical and *in vivo* measurements. The EEG test verified the functionality of Amplifiers I and II to detect weak signals in the presence of environmental noise. BSA II was further tested in a hospital environment on a patient who underwent ECoG grid implantation over the motor cortex area. It was experimentally shown that BSA II was capable of recording signals with high output SNR and comparable performance with respect to a commercial EEG acquisition unit, while consuming orders of magnitude less power.

### A. Electrical Measurements

Amplifiers I and II were fabricated in 130 nm and 180 nm CMOS processes, occupying 0.044 mm<sup>2</sup> / 0.052 mm<sup>2</sup> die areas, and consuming 0.216  $\mu$ W / 0.69  $\mu$ W from 0.4 V / 0.6 V externally provided supply voltages, respectively. Figs. 9 and 10 show die microphotographs of BSA I and BSA II frontend circuits. The first chip (BSA I) occupies 5.45 mm<sup>2</sup>, and the second chip (BSA II) occupies 0.352 mm<sup>2</sup> (excluding pad rings). The pad ring incorporates a 2 kV HBM ESD protection circuitry with negligible leakage current. BSA I prototype uses an off-chip buffer to drive commercial signal acquisition unit (MP150 with 12-bit ADC, Biopac Systems Inc., Goleta, CA) [23]. The overall amplification gain for the two AFEs have been measured using Agilent 33250A waveform generator and SMA attenuators, each providing 39 / 58 dB voltage gain and IRNoise



Fig. 9. Die microphotograph of BSA I with 64-channel amplifier array and serializer [23].



Fig. 10. Die microphotograph of the BSA II with 4-channel amplifier array and serializer.



Fig. 11. Measured and simulated Amplifier I gain and noise responses. Note that the sharp peaks were due to 60 Hz harmonics on the unshielded cables [23].

of ~2.19 / 2.3  $\mu$ V<sub>RMS</sub> across 12–190 Hz / 2–175 Hz of operation bandwidth, respectively (Figs. 11 and 12). Without explicit calibration scheme, the lower-cutoff frequency is not well controlled across process corners. In this work, this frequency was chosen to be smaller than the 8-Hz corner frequency of  $\alpha$ -band with negligible effect on noise performance. Simulations show that this lower-cutoff frequency varies from 2 to 10 Hz across process corners.

The 60 Hz interference and its harmonics were removed from the noise plot and calculations in Fig. 12. Linearity and noise measurements were done using Agilent E4448a spectrum analyzer. A low noise off-the-shelf instrumentation amplifier (AD620) was used to boost the noise level and drive the spectrum analyzer. The calculated dynamic range of the Amplifier I at 37 Hz for ~1% Total Harmonic Distortion (THD) was 58 dB. The Amplifier II harmonics for 0.2 mV input voltage at 47 Hz



Fig. 12. Measured and simulated gain and noise responses for a single channel of BSA II.

(which is 2 times higher than the expected neural signal amplitude) was lower than the measured noise floor, indicating linear operation. For 150 mV<sub>pp</sub> signal at 60 Hz, Amplifier I / II exhibits a CMRR greater than 60 dB / 74 dB and a PSRR greater than 58 dB / 70 dB, respectively. Table IV provides the performance summary of the designed amplifiers and comparison with prior art from academia and industry.

## B. Human Neurological Measurements

This study was approved by the Institutional Review Boards of the University of California, Irvine and the Rancho Los Amigos National Rehabilitation Center, and is considered nonsignificant risk. Three human subjects provided informed consent to participate. The chip was powered by a current-limited (2 mA) supply source. The hospital instruments were disconnected to avoid creation of any unwanted electrical loop. The AC-coupled connection between the electrodes and the amplifiers provides DC isolation.

1) EEG: For two healthy subjects (male, 26 and 27 years old), the impedances of electrodes AFz, Cz, Pz, and Oz in the 10/10 EEG system [44] were reduced to  $< 3 \text{ k}\Omega$  using conductive gel. Measurements were performed on one of the subjects using Amplifier I, as follows. EEG from Cz, Pz, and Oz (all referenced to AFz) was recorded at 2353.2 Hz per channel using a data acquisition system (Biopac MP150). This sampling rate corresponds to a sampling period of  $\sim 100 \ \mu\text{s}$  per channel. The subject was provided verbal cues to alternate between eye opening/closing every 10 s. As a representative example, Fig. 13 shows prominent changes ( $\sim 10 \text{ dB}$ ) in the power of the occipital posterior dominant  $\alpha$  rhythm at channel Oz in both the time series and the time-frequency spectrogram during this task. This is consistent with classic neurophysiological findings [45].

Measurements were performed on the second subject using Amplifier II, as follows. EEG from channel Oz (referenced to AFz) was split to Amplifier II as well as to a commercial bioamplifier (Biopac EEG100C) and sampled at 50 kHz. The output from both systems was downsampled to 2 kHz and filtered into the 8–35 Hz frequencies in software (see Fig. 14). The two signals exhibited a Pearson correlation coefficient of 0.89, and their envelope powers exhibited a correlation of 0.93. In addition, we recorded multiplexed EEG from electrodes AFz, Cz, Pz, and Oz (all referenced to AFz) using Amplifier II and the results after de-multiplexing in software are shown in Fig. 15. KARIMI-BIDHENDI et al.: CMOS ULTRALOW POWER BRAIN SIGNAL ACQUISITION FRONT-ENDS: DESIGN AND HUMAN TESTING

	JSSC 2011 [17]	JETCAS 2011 [19]	TBCAS 2014 [46]	JSSC 2015 [22]	TBCAS 2016 [21]	RHD2000 Intan [47]	Amplifier I	Amplifier II
Power $(\mu W)^{\dagger}$	2.44	56	23-46	1.4	1.08	$7.4^{9}$	0.216	0.69
Supply (V)	2.8	3.3	+/- 0.6	0.5	1	3.3	0.4	0.6
Gain (dB)	39.4	60	Variable	30	40	45.7	39	$39^{1}$
Bandwidth (Hz)	0.36–1.3 k	300	1–15 k	1-500	0.5-150	0.02-1 k <sup>10</sup>	12-190	2-175
IRNoise $(\mu V_{RMS})^7$	3.07	0.5	$13^{2}$	1.23	$N/A^3$	2.4	2.19	2.3
NEF	3.09	4.4	7	3.7	$4.52^{5}$	$4.37^{9}$	4.65	7.22
PEF	$26.7^{6}$	$64^{6}$	$58.5^{6}$	6.9	$20.43^{5}$	63 <sup>9</sup>	11.7	31.3
PSRR (dB)	> 80	69	39-93	67	68	75	58	70
CMRR (dB)	> 66	51	51-97	88	82	82	60	74
Area (mm <sup>2</sup> )	0.13	$86^{8}$	$12.19^{8}$	0.025	0.085	N/A	0.044	0.052
% THD at								
mV <sub>DD</sub> Input	1% (10)	1% (2.6)	$1\% (4.5)^4$	0.4% (1)	1% (4)	0.8% (10)	1% (5)	<1% (0.2)
Technology	$0.6 \ \mu \mathrm{m}$	$0.35 \ \mu \mathrm{m}$	$0.13~\mu{ m m}$	65 nm	65 nm	N/A	$0.13~\mu{ m m}$	$0.18 \ \mu \mathrm{m}$

TABLE IV COMPARISON AND PERFORMANCE SUMMARY OF AFES

<sup>†</sup>Power dissipation includes only the front-end amplifier unless otherwise stated. <sup>1</sup>58 dB for the whole AFE. <sup>2</sup>7.3  $\mu$ V<sub>RMS</sub> for 300–15 kHz. 13 for 1–300 Hz  $\mu$ V<sub>RMS</sub><sup>3</sup>112 nV/ $\sqrt{Hz}$  at 150 Hz. <sup>4</sup> Including ADC. <sup>5</sup> Reported for a defined single frequency. Actual value is higher. <sup>6</sup> Value calculated from reported results. <sup>7</sup> Integrated over the bandwidth. <sup>8</sup> Total chip area. <sup>9</sup> Estimated over 1 kHz bandwidth. <sup>10</sup> Bandwidth tunable from 0.02–20 kHz



Fig. 13. Amplifier I EEG time series (top) and spectrogram (bottom) from channel Oz with 10 dB increase in the posterior dominant alpha rhythm (8–12 Hz) amplitude when the subject closed his eyes (arrow). The subject closed his eyes at 10 and 32 s and opened again at 20 and 42 s [23].

As physiologically expected, electrodes Oz and Pz exhibit larger amplitudes of the occipital posterior dominant  $\alpha$  rhythm during the eyes-closed state.

2) ECoG: One subject (43-year-old male) undergoing ECoG implantation for epilepsy surgery evaluation participated in the study. This subject had an  $8 \times 8$  grid (Ad-Tech, Racine, WI) of 2 mm-diameter electrodes (4-mm center-center spacing) implanted over the primary motor cortex. Fig. 16 shows the locations of the implanted electrodes (derived by co-registering a CT scan and MRI of the head, as in [48]). The subject completed his epilepsy monitoring procedure and was awaiting ECoG grid removal the next day. Hence, the hospital EEG system was disconnected at the time of measurement. ECoG signals were simultaneously routed to Amplifier II and a commercial EEG100C bioamplifier using unshielded cables, as shown in Fig. 17. Negligible loading effect and source impedance mismatch from EEG100C (2 M $\Omega$  input impedance) on Amplifier II is expected due to a relatively small electrodes' impedance



Fig. 14. Top: PSD of the BSA II (red) and commercial (black) bioamplifier from 30 s of EEG data. Bottom: EEG  $\alpha/\beta$ -band (8–35 Hz) time-series data from channel Oz (referenced to AFz) using the BSA II (red) and commercial (black) bioamplifiers. The subject was instructed to alternate between eyes-open (white background) and eyes-closed (blue background).

(<1 kΩ). ECoG electrodes' impedance is reported to be stable over time [6], eliminating the need for constant monitoring of its value. The output from both amplifiers was recorded at 25 kHz by the MP150 system for 30 s. Note that the subject was asleep during this time and did not participate in any associated behavioral task for further verification of the amplifier array. The resulting signals were then downsampled to 2 kHz in software for further processing. The correlation coefficient between the signals from BSA II and the EEG100C was 0.99 from 8–35 Hz (covering the  $\alpha$  and  $\beta$  bands), 0.94 from 35–70 Hz (low- $\gamma$  band), and 0.72 from 70–120 Hz (high- $\gamma$  band). Moreover, the correlation between each system's envelope power in  $\alpha/\beta$ , low- $\gamma$ , and high- $\gamma$  bands was 0.99, 0.99, and 0.89, respectively. This slight decrement in the high- $\gamma$  band correlation between the



Fig. 15. BSA II EEG  $\alpha/\beta$ -band (8–35 Hz) time-series data (top) and spectrogram (bottom) from channel AFz, Cz, Pz, and Oz (all referenced to AFz) as the subject was instructed to alternate between eyes-open (white background) and eyes-closed (blue background). The channel-multiplexed data from the custom designed IC were demultiplexed in software.



Fig. 16. MRI of the patient with implanted ECoG grid over the left motor cortex. Electrodes 28 and 24 were used as the reference and ground, respectively.



Fig. 17. In vivo ECoG measurement setup.

bioamplifiers is expected since the signal power decreases with frequency and approaches the Amplifier II's noise floor. A software notch filter was applied on the signal from 57 to 63 Hz before calculating the correlations. A representative PSD of the recorded signals across the  $\alpha$ ,  $\beta$  and  $\gamma$  bands (8–120 Hz) and 10-s output time-series of BSA II and its commercial counterpart are shown in Fig. 18, demonstrating qualitative similarities between the two. The peaks at 60 Hz for custom and commercial



Fig. 18. Top: PSD of the BSA II (red) and commercial (black) bioamplifier from 30 s of ECoG data. Middle: Filtered (8–120 Hz) time-series data from the implanted ECoG grid with the BSa II (red) and commercial (black) bioamplifier. Bottom: Zoomed-in view of the recorded ECoG.

PSDs are caused by limited CMRR of the signal chain as well as the coupled power line interference to the unshielded interface between the analog output and the external ADC.

#### VI. CONCLUSION AND FUTURE WORK

Two brain signal acquisition front-ends designed in the WI region were presented. Fabricated in 130 nm and 180 nm CMOS processes, each amplifier within the arrays consumes 0.216 / 0.69  $\mu$ W, respectively (not including buffer and InAmp). Measured IRNoise across the bandwidth was 2.19 / 2.3  $\mu V_{RMS}$ corresponding to NEF of 4.65 / 7.22 and PEF of 11.7 / 31.3 [49]. Objective comparison of human in vivo EEG and ECoG measurements acquired by our custom IC and a commercial bioamplifier demonstrated that our BSAs were able to record these neural signals reliably. This suggests that the circuit architecture presented in this work can serve as the basis for a highly miniaturized and ultra-low power brain signal acquisition unit for a future fully implantable BCI system. Future work will focus on further reducing the susceptibility of the front-end to environmental noise, e.g., including an on-chip analog-to-digital converter, and incorporating the capability of large interference rejection at low supply voltages in the presence of a sensory feedback stimulation circuitry.

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