DSP for Coherent Single-Carrier Receivers

Maxim Kuschnerov, Student Member, IEEE, Fabian N. Hauske, Member, IEEE, Kittipong Piyawanno, Bernhard Spinnler, Mohammad S. Alfiad, Student Member, IEEE, Antonio Napoli, and Berthold Lankl, Member, IEEE

Abstract—In this paper, we outline the design of signal processing (DSP) algorithms with blind estimation for 100-G coherent optical polarization-diversity receivers in single-carrier systems. As main degrading optical propagation effects, we considered chromatic dispersion (CD), polarization-mode dispersion (PMD), polarization-dependent loss (PDL), and cross-phase modulation (XPM). In the context of this work, we developed algorithms to increase the robustness of the single DSP receiver modules against the aforesaid propagation effects. In particular, we first present a new and fast algorithm to perform blind adaptive CD compensation through frequency-domain equalization. This low complexity equalizer component inherits a highly precise estimation of residual dispersion independent from previous or subsequent blocks. Next, we introduce an original dispersion-tolerant timing recovery and illustrate the derivation of blind polarization demultiplexing, capable to operate also in condition of high PDL. At last, we propose an XPM-mitigating carrier phase recovery as an extension of the standard Viterbi–Viterbi algorithm.

Index Terms—Blind adaptation, carrier phase recovery, coherent detection, digital receiver, equalization, fiber optic communication, timing recovery.

I. INTRODUCTION

DIGITAL SIGNAL PROCESSING (DSP) algorithms for blind coherent fiber optic receivers can be derived from their wireless communication counterparts. However, modifications and further developments are necessary and inevitable considering the special nature of the fiber channel. Although the algorithms have to be as simple as possible to enable high-speed processing, the channel impulse response that can arise in the fiber channel differs from typical wireless systems and can be significantly longer. Since the variations in the fiber channel are very slow compared to the symbol rate, adaptation speed is of less concern, so that blind adaptation can be used.

Chromatic dispersion (CD) is typically compensated for by a finite impulse response (FIR) butterfly structure [1]. In case of large accumulated CD in uncompensated links, the equalization of CD and PMD can be performed in two steps, first compensating for the static dispersion [2]. Here, the equalizer usually cannot be properly adapted using standard adaptation algorithms due to the severity of the distortion. In various investigations based on measured data, the filtering function for the CD compensation was either a priori known or no information on the blind adaptation scheme with an appropriate analysis was given [2].

Next, a correction of the timing frequency and phase offset is necessary before the FIR butterfly filter. Conventional timing recovery algorithms cited in [1] and [3] lack a high dispersion tolerance that is usually required in optically compensated systems, thus creating a possible bottleneck in the receiver design.

Contrary to wireless communication using polarization multiplexing [4], where the absolute signal polarization only slightly deviates from the original axes of the transmitter, the polarization evolution in typical fibers is arbitrary. Coherent receivers therefore require either a polarization control to align the signal and the local oscillator (LO), or polarization diversity reception, where the cross-polarization interference can be compensated for in the digital domain. Modifications to blind equalizer adaptation algorithms are required to achieve the demultiplexing of the two polarizations.

Finally, XPM has been identified as the major limitation in the multichannel transmission of quaternary phase-shift keying (QPSK) over legacy systems with ON–OFF keying (OOK) neighbors [6], [7]. XPM can be mitigated in the carrier phase recovery averaging over only a small number of taps due to the rapid XPM-induced phase fluctuations [8].

This paper presents solution concepts for all of the above mentioned problems of DSP algorithm design, emphasizing the subcomponent interaction, and is structured as follows. Section II provides an overview on the coherent optical receiver, while Section III introduces the frequency-domain equalizer as the first component of the digital receiver and presents a blind CD estimation algorithm. Sections IV and V present the timing recovery and the blind polarization demultiplexing, respectively. Section VI deals with the XPM-mitigating carrier phase recovery. Finally, Section VII draws the conclusions. The digital signal processing principles will be mainly demonstrated on single-carrier polarization-multiplexed (PolMux) QPSK.

Fig. 1. Polarization-diversity coherent receiver front-end.
although they can as well be applied to higher order modulation formats.

II. BLIND COHERENT RECEIVER DESIGN

A typical coherent receiver front-end employing polarization diversity is shown in Fig. 1 [1]. After a polarization-beam splitter, the two orthogonal polarizations are mixed with a local oscillator in two 90° optical hybrids and sampled in an analog-to-digital converter (ADC) after the automatic gain control (AGC). Depending on the distortion-tolerance requirements, the ADC can sample with one or two samples per symbol. While a twofold oversampling makes it possible to fully equalize the signal, symbol-rate sampling can be employed in lower complexity systems with a limited dispersion tolerance [9]. The local oscillator can be controlled via low-bandwidth feedback from the receiver to achieve a rough frequency alignment, while a precise digital carrier frequency acquisition continuously tracks the remaining frequency and phase offset in the electrical domain. Further digital signal processing is required in order to compensate for linear distortion, mitigate nonlinear transmission effects, and demultiplex both polarizations.

In the following, the signals in the four real-valued tributaries will be described in complex vector notation form leading to

$$\mathbf{p}[k] = \begin{pmatrix} p_x[k] \\ p_y[k] \end{pmatrix} = \begin{pmatrix} x_{Re}[k] + j x_{Im}[k] \\ y_{Re}[k] + j y_{Im}[k] \end{pmatrix}. \quad (1)$$

Here, \( p_i[k], i = x, y \), is the complex symbol vector at the time instant \( k \) consisting of one sample \( p_x[k] \) for Nyquist rate sampling or two samples \( \{p_{x1}[k], p_{x2}[k] \} \) for twofold oversampling. A typical blind receiver block diagram is shown in Fig. 2.

The receiver employs frequency-domain equalization (FDE) to compensate for the bulk of accumulated chromatic dispersion. Then, the signal is retimed, demultiplexed, and equalized in a butterfly FIR filter. Finally, the constellation diagram is derotated in the carrier frequency and phase recovery. Throughout this paper, a 35-GHz, second-order optical Gauss filter and a 19.6-GHz fifth-order electrical Bessel filter were assumed.

III. DISPERSION COMPENSATION

A. Filter Design

Coherent transmission systems can be operated either on optically compensated links with dispersion-compensating fiber (DCF), or over fully uncompensated links with no DCFs in place. In the latter case, system installation can be made more cost-efficient and tolerant to nonlinearities [10]. If dispersion compensation is performed in front of the timing recovery and the other processing blocks in the receiver, the subsequent algorithms can remain identical for compensated and uncompensated transmission with similar residual dispersion requirements.

Dispersion can either be compensated using time-domain (TDE) [2] or frequency-domain equalization [11]. Fig. 3 shows the equalizer structures for TDE and FDE equalization. The filter choice between TDE and FDE mainly depends on the maximum dispersion in the channel and the resulting filter length. Fig. 4 compares the equalization complexity for 43- and 112-Gb/s PolMux-QPSK receivers [11]. Here, a FIR butterfly with five taps (43 Gb/s) and nine taps (112 Gb/s) was assumed, in order to compensate for PMD and low-pass filtering. The filter length is set to a maximum of 1-dB required OSNR penalty for the given chromatic dispersion. While the CD compensating filters were assumed nonadaptive for the computation, the time-domain butterfly filter was adapted using the least-mean square algorithm (LMS) [12]. Fig. 4 shows that FDE achieves a break-
even over TDE starting at \( \sim 11,000 \) ps/nm in 40 G systems and at \( \sim 2000 \) ps/nm in 100-G systems.

Another possibility for dispersion compensation is infinite impulse response (IIR) filtering, where the required tap number is lower as for FIR filters [13]. However, the inherent feedback of IIR equalizers makes this approach virtually impossible to implement in high-speed applications with parallelized signal processing. In the following, dispersion estimation will be demonstrated on the example of FDE, although it can as well be applied to time-domain filters.

### B. Dispersion Estimation

The receiver must be able to estimate the chromatic dispersion at system startup and continuously monitor it during transmission. Ideally, the estimation is performed in the dispersion-compensating device, independently of the follow-up DSP, thus increasing the startup speed of the receiver.

If the dispersion is to be estimated, one can distinguish between data-aided and blind estimation algorithms. The pros and cons for each are listed in Table I.

In addition to these arguments, one has to add that there is only one unknown parameter in the second-order approximation of the chromatic dispersion transfer function, which can be analytically described by

\[
G(I, \omega) = \exp \left( -j D \frac{\lambda^2}{2 \pi c^2} \frac{\omega^2}{2} I \right) \quad \text{(2)}
\]

where \( I \) is the transmission distance, \( D \) is the dispersion parameter, \( \lambda \) is the carrier wavelength, \( c \) is the speed of light, and \( \omega \) is the angular frequency. Clearly, the equation, which describes a parabolic phase transfer function, has only one degree of freedom, so that a fully adaptive estimation is not justified under these conditions.

In the following, a blind estimation algorithm first introduced in [14] will be presented that restricts the equalizer solutions to the inverse of the dispersion transfer function from (2), thus stabilizing the equalizer convergence. The presented minimum-search algorithm is nondaed aided and works prior to the timing recovery. Given a minimum and a maximum dispersion in the channel, predefined CD-values are loaded in the filter as shown on Fig. 5.

The proposed minimum search algorithm uses an error criterion that was derived from Godard’s CMA [15]. It employs a cost function working on the twofold oversampled signal given by

\[
J[k] = \sum_{i=1}^{2} \sum_{m=1}^{2} \left| q_{i,m}[k]^2 - R_{i,m}[b] \right|
\]

where \( R_{i,m}[b] \) are the normalization constants of the odd and the even samples for \( i = x, y \) and \( b \) is index of the fast Fourier transform (FFT)-block. In order to account for the changing timing phase due to the timing frequency offset between the transmitter and receiver oscillators, the normalization constant \( R_{i,m}[b] \) is determined separately for every FFT signal block. They are computed from the mean power for the odd and even samples after equalization given by

\[
\eta_{i,m}[b] = \frac{1}{N_{eq}} \sum_{k=0}^{N_{eq}-1} \left| q_{i,m}[k + \delta] \right|^2 \quad \text{(4)}
\]

where \( N_{eq} \) is the number of symbols equalized per each FFT. Here, a simple case differentiation of the power ratio \( \bar{q}_{i,2}[b]/\bar{q}_{i,1}[b] \) for each FFT-block is sufficient to give optimum results regardless of return-to-zero (RZ) or nonreturn-to-zero (NRZ) pulse shaping. It is given in Table II.
In the following, the parameters have been optimized using RZ-pulse shaping with the constraint of a minimum estimation error. They are set to $\xi = 1.25$, $R_{\text{in}} = 0.6$, $R_{b} = 1.5$, and $R_{c} = 2$ with the signal power in each polarization normalized to half the total signal power.

For every value of chromatic dispersion, the total cost function is given by

$$J[CD] = \frac{1}{N_{\text{max}}} \sum_{i=1}^{2} \sum_{\delta=0}^{N_{\text{max}}-1} J_i[k + \delta]$$

(5)

where $W[CD] \in [-\text{CD}_{\text{max}} : \delta\text{CD} : -\text{CD}_{\text{min}}]$, with $N_{\text{max}}$ symbols used for the cost function of each CD value. The minimum of $J[CD]$ indicates the correct value of chromatic dispersion. After a rough scan with a large $\delta\text{CD}$, the dispersion values around the minimum can be scanned at a higher resolution improving the estimation precision. An exemplary estimation of the optimum filter value with a rough scan is given in Fig. 6.

The main limiting parameter for the proposed algorithm is PMD, since this effect is not modeled in the transfer function $W[CD]$, deteriorating the estimation. Fig. 7 shows the standard deviation of the estimator versus the mean DGD for 112-Gb/s polarization multiplexed QPSK and 16 quadrature amplitude modulation (QAM). Here, a maximum dispersion of 48,000 ps/nm was computed from the 3-dB penalty limit of the required OSNR at BER = $1 \times 10^{-3}$ for an FFT size of 1024 symbols, where half of the samples were used for overlap. In general, the overlap size should be larger than the length of the channel impulse response. An average over eight FFT blocks was used for the cost function computation at each CD value.

Although the computation of $R_{\text{in},m}[b]$ was optimized for RZ-QPSK, NRZ-QPSK as well as 16-QAM also perform similarly. With varying FFT sizes, the estimation performance remains virtually the same, if the identical number of symbols $N_{\text{max}}$ is used for the computation of $J[CD]$, and $N_{\text{overlap}}$ and $\text{CD}_{\text{max}}$ are appropriately scaled with the FFT size.

The required tap number of the subsequent FIR butterfly filter is governed by the mean DGD and the maximum residual CD after the dispersion compensating block. The proposed algorithm avoids an additional FFT for feedback that is inevitable in standard adaptive FDE [16]. Furthermore, it is not significantly affected by nonlinearities as demonstrated by measurements in [17]. After the dispersion estimation at the startup of the transmission, tracking is possible estimating the residual CD from the taps of the FIR butterfly equalizer [18].

### IV. TIMING RECOVERY

The task of the timing recovery is to correct for the timing phase and frequency offset between the transmitter and receiver clocks. The timing error can be corrected in a feedforward structure, e.g., using the square timing recovery [19], where an interpolation to four samples per symbol is required. If two samples per symbols are used, an early–late phase error indicator can be derived using, e.g., the Gardner algorithm [20]. The phase error is then typically fed into a second-order phase-locked loop (PLL). The correction of the timing phase is performed either in the digital domain via an interpolator, as shown in Fig. 2, or directly in the ADC.

Fig. 8 shows a general block diagram of a timing recovery. After prefiltering the signal, the timing phase error is detected using the known algorithms and then typically averaged in a postfilter.

Depending on the residual estimation error of the electrical CD compensation or on the residual dispersion in optically compensated systems, a certain dispersion tolerance of timing recovery algorithms can be desired. Conventional timing recovery algorithms rely on the filtering of the spectral line at the symbol rate after the squaring of the signal. With increasing dispersion, the spectrum clock line disappears as shown in Fig. 9, so that no reliable timing phase error can be generated.

Since chromatic dispersion has an all-pass character, theoretically, a prefilter can be designed that perfectly recovers the spectral line. However, such a filter would have to be adaptive, introducing further complexity to the receiver. In [21] a simple, suboptimal prefilter was introduced that increases the tolerance.
of the timing recovery up to arbitrary values of chromatic dispersion. The filter is given by

$$h_{\text{pre},m}[n] = (-1)^n$$

(6)

where \( n = 0, \ldots, N_{\text{filt}} - 1 \) is the filter length index, \( N_{\text{filt}} \) is the number of filtered symbols, and \( m = 1, 2 \) for twofold oversampling. The resulting filter is a band-pass around the Nyquist frequency. Increasing the filter length \( N_{\text{filt}} \), the band-pass of the filter can be decreased, thus lowering interfering beat terms at the symbol rate after signal squaring. The filtering requires only adders and can be performed separately on the odd and even samples of the signal with filter length \( N_{\text{filt}}^a = N_{\text{filt}}/2 \) to further reduce complexity, which will be assumed in the following.

The transfer function value is shown in Fig. 10 for several realizations of \( h_{\text{pre}} \).

The minimum required length for the timing recovery prefilter is defined by the signal spread due to chromatic dispersion, similar to FIR filter length requirements for equalization. The dispersion tolerance of the timing recovery will be demonstrated using the timing detector presented in [21]. The performance of the histogram timing recovery can be verified against the modified Cramer–Rao bound (MCRB), which is the lower bound on the normalized estimation variance and given by

$$\frac{\sigma^2}{T_s^2} = \frac{1}{8\pi^2 L \xi T_s N_0}$$

(7)

for pulse-amplitude modulation, where \( T_s \) is the symbol duration, \( L \) is the averaging length of the postfilter, and \( G(f) \) is the spectrum of the signal pulse shape. Fig. 11 shows the estimator variance versus chromatic dispersion for several prefilter lengths.

In a back-to-back configuration with zero dispersion and no prefiltering, the estimator performance is very close to the MCRB. The maximum dispersion that can be tolerated in this case is around 300 ps/nm. The dispersion tolerance requirement in optically compensated systems is usually higher, and in the case that the previously presented electrical dispersion compensation was used, the estimation precision can deteriorate for high PMD. In order to remove the timing recovery bottleneck, the prefilter can be introduced, clearly increasing the dispersion tolerance at the cost of a back-to-back degradation. A similar performance as in back-to-back configurations without prefiltering can be achieved, if the averaging length \( L \) of the filter is increased, thus lowering the overall response bandwidth.

V. BLIND POLARIZATION DEMULTIPLEXING

Polarization multiplexing in the fiber optic channel is a principle that has been used in radio transmission since the 1980s [4]. The optic channel is a special case of the more general multiple-input–multiple-output (MIMO) systems, widely applied in space-division multiplexing. Polarization demultiplexing is one of the most interesting problems of coherent receiver design.

The CMA is a popular choice for approximate channel acquisition that can be followed by the LMS for tracking purposes. If no known training symbols are available at the receiver, translating the CMA from a single-input–single-output (SISO) channel to a \( 2 \times 2 \) MIMO equalizer [26], as shown in Fig. 3, results in the possibility of the degenerate one-to-many output problem.
In [2] and [22], the PMD-matrix constraint was imposed on the equalizer filter taps, using the unitary properties of the fiber channel transfer function given by

\[ \mathbf{H}(\omega) = G(I, \omega)\mathbf{U}(\omega) = G(I, \omega) \cdot \begin{pmatrix} u(\omega) & v(\omega) \\ -v^*(\omega) & u^*(\omega) \end{pmatrix} \]  

with \( \mathbf{U}(\omega) \) being the PMD transfer matrix with \( \det(\mathbf{U}(\omega)) = 1 \). Although this approach works well as long as the matrix is unitary, once significant PDL is present in the channel, the transfer function loses its previously described characterization and the probability of a degenerate equalizer solution increases as shown in Fig. 12, with the identical polarization appearing at both outputs of the equalizer. Here, one can distinguish between a best and a worst case in terms of performance, depending on the input polarization into the PDL element at the beginning of the transmission [5]. For a given channel realization, the equalizer converged using CMA/LMS with the constraint of (9). In the degenerate case of the same polarization at both equalizer outputs, a convergence failure was counted. Since the degenerate convergence is a random process that depends on the transmitted signal, noise, distortion, as well as the update speed of the equalizer, a convergence failure probability was evaluated over a sufficient number of trials.

Another solution for the demultiplexing problem have been demonstrated for one-tap filters only [23], or have not been fully commented upon [5]. In [24], a semiblind approach is described, where one polarization is adapted blindly and the other one afterwards with the help of the extracted training symbols from the first polarization, presumably using higher layer information bits. Introducing training symbols in all polarizations might seem as the most elegant solution [25]. However, it comes at the cost of signal overhead, so that a fully blind approach might be preferred for 100-G QPSK systems.

The problem of blind source separation based on the CMA has been widely covered in wireless literature [27], [28]. While the CMA is able to achieve optimal equalization using second-order statistics in combination with oversampling, MIMO systems require higher order statistics [29], and independent-component analysis (ICA) in order to achieve blind source separation, as introduced by [30] for fiber optic systems and one-tap filtering. While [30] uses an ICA cost function different from the CMA or LMS, the source separation method described in [28] has the advantage that it can be added on top of the existing CMA/LMS algorithms, thus leading to virtually identical convergence and tracking properties in most cases as standard CMA algorithms. It is given by

\[ J[k] = E \left( \sum_{i=1}^{2} (|z_i[k]|^2 - R_2)^2 + 2 \sum_{L,m=1,L \neq m} \sum_{\xi=\xi_1}^{\xi_2} |\rho_{lm}[\xi]|^2 \right) \]  

where \( E \) is the expectation value. Here \( \rho_{lm}[\xi] \) is the cross-correlation function between polarization \( l \) and \( m \) defined as

\[ \rho_{lm}[\xi] = E(z_l[k]z^*_m[k-\xi]) \]  

and \( \xi_1 \) and \( \xi_2 \) are integers that depend on the channel delay spread. By minimizing the cross-correlation between the channels, both sources can be separated in presence of PDL. Fig. 13 evaluates the performance of the algorithm versus PDL for simulations of 112-Gb/s PolMux-QPSK using a 15-tap T2 fractionally spaced FIR butterfly with worst case distortions of CD = 1000 ps/nm and 30-ps mean DGD (full-order PMD). Even in this case, where the channel memory length begins to exceed the number of equalizer taps, the theoretical performance can approximately be achieved for PDL ranging from 0 to 10 dB.

After initial convergence with the CMA, faster convergence can be reached by switching to the LMS, especially if significant PDL is present. However, the LMS algorithm is decision directed, requiring a prior correction of the carrier phase and frequency offset.

VI. CARRIER PHASE RECOVERY

The mixing with the local oscillator introduces a frequency and phase offset, identical in both polarizations, leading to a...
rotating constellation diagram. Feedback or feedforward algorithms can be used to compensate for the LO phase offset [31]. While the LO-phase recovery should have a rather low bandwidth in order to minimize noise influence, a second high-bandwidth component can be employed to mitigate for nonlinearities, if SPM and XPM are the main limiting effects over amplified spontaneous emission (ASE) and the LO phase noise. If the coherent polarization multiplexed channel is transmitted in overlay to legacy OOK channels, the resulting cross-phase modulation (XPM) can be effectively compensated using the phase recovery [8]. Here, the $n$th power Viterbi-and-Viterbi (V&V) algorithm was used for phase offset estimation [32] by reducing the averaging length for the feedforward structure down to 2–5 symbols. The estimator is given by

$$\hat{\theta}[k] = \frac{1}{\delta} \arg \left\{ \sum_{\delta=-\delta_1}^{\delta} a_\delta (z[k + \delta])^{\delta} \right\}$$

(12)

where $a_\delta$ is a weighting factor, and $\delta_1$ have to be chosen according to the required tap number. In AWGN systems, $a_\delta$ can be optimized depending on the signal-to-noise ratio [33]. However, in the case of OOK-neighbor limited transmission, an optimization of $a_\delta$ did not yield significant improvements for the low tap numbers in contrast to $a_\delta = 1, \forall \delta$. Inevitable cycle slips were mitigated using differential precoding of the phase-modulated signal.

Standard implementations of the V&V algorithm, as used in [8], process each polarization independently and do not make use of the phase correlation between the two polarizations. The joint-polarization (JP) algorithm [34] is an extension of the standard V&V algorithm, making use of the correlated carrier phase information in both polarizations. In case that the carrier in both polarizations is synchronous, e.g., when using the LMS algorithm for tracking in the equalizer, the phase offset information in the two polarizations can be simply superpositioned, as shown in Fig. 14.

Note that the feedforward structure requires phase unwrapping, introducing feedback [31]. If, however, the equalization is performed using the CMA only, the carriers in the two polarizations are typically asynchronous, requiring feedback as shown in Fig. 15.

Fig. 16 shows the performance of the JP phase recovery compared to V&V for the measurements as described in [8] with a 43-Gb/s PolMux-NRZ-QPSK channel and nine copolarized 10.7-Gb/s NRZ-OOK neighbors at arbitrary orientation to the QPSK channel on the 50-GHz ITU grid. Timing recovery and equalization were performed using methods introduced before. The LO phase and frequency offset was corrected using a PLL as outlined in Fig. 2 for LMS adaptation as in Fig. 14. CMA-only adaptation with a phase recovery as in Fig. 15 led to identical BER. The phase was averaged over three taps in each polarization, giving the best overall performance.
The performance gain is significant for input powers up to $-4 \text{ dBm}$ per channel and saturates due to high XPM at $-2 \text{ dBm}$ for the given transmission distances. Fig. 17 shows the influence of XPM on the optimum coupling factor $\chi$ that decreases for higher input power due to different nonlinear phase shifts in both polarizations [36].

The correlation of the phases in the two polarizations decreases with high XPM, and a near-optimum performance can be reached with a constant $\chi \approx 0.8$ both for CMA and LMS implementations [34]. These improvements lead to a significant increase of the ASE noise tolerance, as demonstrated in [7].

VII. CONCLUSION

This contribution outlined the design of the key DSP algorithms for fiber optic coherent receivers. A blind stable dispersion estimation algorithm for arbitrarily large dispersion was presented in combination with frequency-domain equalization. The following timing recovery with prefiltering can be designed to tolerate any residual dispersion requirements, which is not possible using conventional algorithms, thus removing a critical bottleneck in receiver design. After the correction of the timing phase and frequency, the proposed equalization cost function can be used to derive polarization demultiplexing tolerant to PDL under worst case linear distortions and avoiding degenerate equalizer solutions. Finally, a JP version of the V&V algorithm for the carrier phase recovery was presented, further increasing the nonlinear tolerance in legacy channels with OOK-neighbors.

ACKNOWLEDGMENT

The authors would like to thank D. van den Borne, T. Wuth, E.-D. Schmidt, T. Duthel, C. Fludger, F. Bondoux, and E. Gourdon for all the valuable discussions and their contribution to this work.

REFERENCES


His research focuses on digital signal processing and its application in optical transmission systems.

Fabian N. Hauske (M’09) was born in Munich, Germany, in 1977. He received the Dipl.-Ing. degree in electrical engineering from the Technical University of Munich, Munich, Germany, in 2003. From 2000 to 2001, he was a visiting student at the University of Nottingham, U.K. Since 2003, he has been with the Institute for Communications Engineering, University of the Federal Armed Forces, Munich, working towards the Dr.-Ing. degree in collaboration with Nokia Siemens Networks GmbH & Co. KG.

In 2003, he carried out research at the Fujitsu Laboratories, Ltd., Kawasaki, Japan, within the scope of his diploma thesis. His fields of interest cover digital signal processing algorithms for equalization and optical performance monitoring for next generation optical transmission systems.

Mohammad S. Alfiad (SM’05) was born in Zarqa, Jordan, in 1982. He received the M.Sc. in broad band telecommunication (cum laude) from Eindhoven University of Technology, Eindhoven, The Netherlands, in 2007. His M.Sc. thesis research was conducted at Siemens AG, Munich, Germany. Currently, he is working towards the Ph.D. at the Eindhoven University of Technology in cooperation with Nokia Siemens Networks, Munich, Germany.

His research focuses on digital signal processing and 100-Gb/s Ethernet for long-haul fiber-optic transmission systems.

Antonio Napoli was born in Cossato, Italy, in 1974. He received the M.S. degree in electronics engineering and the Ph.D. degree with a dissertation on electronic equalization for advanced modulation formats from Politecnico di Torino, Torino, Italy, in 2002 and 2006, respectively.

During his studies, he was a visiting student at the Technical University Wien, Universitat Politecnica de Catalunya, and University College London. In 2006, he joined the R&D of Siemens which later merged into Nokia Siemens Networks, where he is working on digital signal processing algorithms for optical communications systems. His interests cover the areas of digital signal processing, advanced modulation formats, forward error correction, and equalization.

Berthold Lankl (M’82) was born in Tirschenreuth, Bavaria, in 1959. He received the Dipl.-Ing. degree in electrical engineering from the Technical University of Munich, Munich, Germany, in 1983 and the Dr.-Ing. degree from the University of the Federal Armed Forces, Munich, Germany, in 1989.

From 1983 to 2001, he worked at the Transmission Systems Division of Siemens AG, Munich, Germany, particularly in the development of radio relay systems. From 2001 to 2003, he was responsible for a research department in the field of optical communications. Since October 2003, he has been a Full Professor of Communications Engineering at the University of the Federal Armed Forces, Munich, Germany. Dr. Lankl is a member of Verband der Elektrotechnik, Elektronik und Informationstechnik e.V. (VDE) and Informationstechnische Gesellschaft im VDE (ITG).