

1. introduction There are two reasons why deep learning is used.

서식 있음: 글꼴 색: 텍스트 1

With its recent development ~~of deep learning~~, deep learning ~~it~~ has shown great achievements ~~in when applied to other areas outside the computer science domain~~. Therefore, ~~it deep learning is has~~ also been applied to physical layer communication research such as channel coding and ~~MIMO(Multi Input Multi Output)~~ antenna technology. ~~There are two reasons why deep learning is used to communication system. Deep learning is has recently been applied to communication research for the following two main reasons: Performance improvement, Reduced complexity.~~ The first reason is to apply deep learning technology with decoding to achieve better performance. ~~First, the primary goal of communication is that the receiver must receive the message accurately, but noise is added when the a message passes through the a channel, possibly so that preventing the receiver may not from receiving the message. To improve this, there is a process called encoding and decoding in the classical communication domain.~~ ~~There~~ The first reason is to apply deep learning technology with decoding to achieve better performance. ~~Second is, a p~~Popular decoding method, known as the Belief Propagation Algorithm (BPA), ~~or the other also~~ named the Sum Product Algorithm (SPA). This algorithm ~~,~~ has good performance, but it ~~consists of is composed of~~ many multiplication operations. Therefore, the longer the length of the message used as an input value, the more complicated calculation is. At that time, -second reason to use deep learning is reducing complexity. ~~To improve this complexity solve this problem, there is a solution called the min-sum algorithm (MSA). In -this min-sum algorithm, the complexity problem was improved, but performance loss degradation occurred. To properly adjust this trade-off relationship of between performance and complexity, there are two algorithms:- 1.First, a normalized min-sum algorithm (NMSA) that multiplies correction factor value, which is a constant value, from the check node update process. 2.-Second, an offset min-sum algorithm(OMSA) that adds or subtracts correction factor value from check node update process. By incorporating deep learning into existing communication systems, correction factor is optimized. As a result, Complexity is improved compared to BPA, and performance is improved compared to MSA. The reason for use of deep learning these systems is that it wants to improve performance can be further improved, more than before by incorporating deep learning into existing communication system.~~

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메모 포함[NP1]: Is this an example?

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<related work>

Recently, many researchers have been actively researching methods to incorporate

deep learning into channel coding. A formative study, by Nachmani, channel-coding's prior researcher, used deep learning in the decoding process, and by By setting different weight values at the edges connecting check nodes(CN) and variable nodes(VN), [1] improved performance was improved by reducing the effect of small cycles in tanner graphs during the decoding process and is very meaningful that used deep learning for the first time. [1] In 2012, deep learning applied to PA, Lugosch applied to OMS, but OMS Deep Learning. The OMS algorithm was used to obtain OMS's optimized correction factor value. Particularly, OMS is an algorithm consisting of addition and subtraction, not multiplication, so it is a suitable method for an algorithm to hardware because of its low complexity. This algorithm is called a neural offset min-sum (NOMS). In the case of [3], which conducted a similar research where a neural normalized min-sum (NNMS) was proposed using an optimized correction factor by-through deep learning. And Alternatively, Wang suggested another way method. In order to improve the complexity problem, he used a sharing shared method that uses the same correction factor value for each iteration, unlike recent one studies that used different correction factor values for each iteration and node. This algorithm is called a Shared Neural NMS (SNNMS).

Aforementioned researches mentioned used deep learning for to optimizing correction factor. However, there is a study focusing on refining the deep learning architecture for this application. Deep Learning has several architectures such as Deep Neural Network (DNN), Convolutional Neural Network (CNN), and Recurrent Neural Networks (RNN) and so on. In [4], it is a seminal research work using 'RNN'. It is called a 'circular neural network', and This RNN is a deep learning architecture that utilizes past data for learning through concept of a recurrent. In other words, it is an algorithm that utilizes not only current inputs but also past data for learning. Remarkably, And, this research is remarkable because it was first study to incorporate RNN into the decoding process and showed similar performance to previous prior studies using fewer parameters. However, this 'RNN' based method has two limitations. To be specific, input vectors are entered sequentially to enable sequential data processing, but 'parallelization operation' is not possible. And The derivative value of tanh, activation function of RNN, is used in this case. However, there is a disadvantage that back propagation information is rarely transferred because a vanishing gradient occurs.

with. And proposed method used the concept of relaxation [5]. Then, LSTM is a special case of 'RNN' and purpose of using relaxation concepts. And The purpose of the proposed method is to determine how much previous data to use previous data. And it is noteworthy that this method optimized decoder relaxation factor through deep learning rather than previous method, brute force simulation. In order to show excellent performance excellence off from our proposed method, simulations were conducted in BCH code, which is a high density parity check code (HDPC) and low density parity check code (LDPC) with different lengths. And S-emiconductor

메모 포함[NP2]: Instead of just listing, compare these works. What was done better, or worse or was improved upon.

메모 포함[NP3]: Use a reporting verb instead of "did a study"

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메모 포함[NP4R3]: E.g. Lugosch examined the application of an OMS algorithm on a

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