

LDPC 부호의 Degree Distribution 설계 방법

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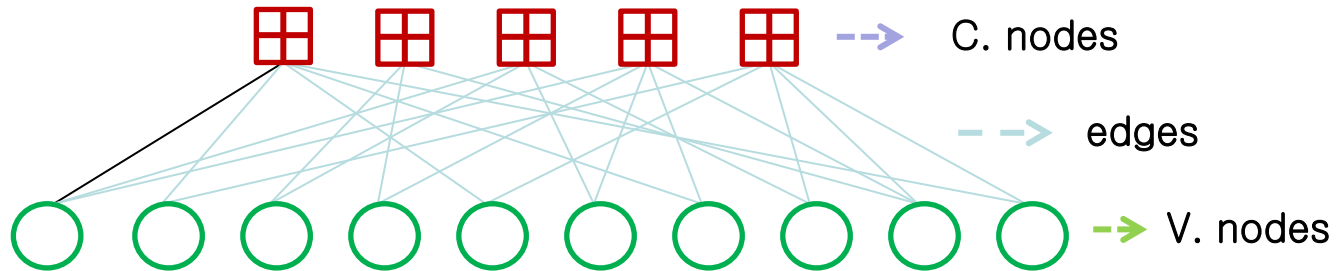
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연세대학교

부호 및 암호 연구실

송민규, 박진수, 송홍엽

LDPC Codes



$$\lambda(x) = \sum_{i=2}^{d_v} \lambda_i x^{i-1} \quad \rho(x) = \sum_{i=2}^{d_c} \rho_i x^{i-1} \quad \text{Code rate } R = \frac{K}{N} = 1 - \frac{\sum_{i=2}^{d_c} \frac{\rho_i}{i}}{\sum_{i=2}^{d_v} \frac{\lambda_i}{i}}$$

λ_i : fraction of edges which are connected to degree i variable nodes
 ρ_i : fraction of edges which are connected to degree i check nodes

Ensemble Optimize Problems

There are two objects

- Optimize code rate R for given threshold
- Threshold s^* for given code rate → **practically important!**

Optimize s^*

Minimize s^*

subject to

$$\sum_{i=1}^{d_c} \rho_i = 1$$

$$\sum_{i=1}^{d_v} \lambda_i = 1$$

$$p_e \rightarrow 0 \text{ as } l \rightarrow \infty$$

by tuning

$$\rho(x), \lambda(x)$$

for given

$$R, d_v, d_c$$

- Richardson 01

Optimize R

Maximize R

subject to

$$\sum_{i=1}^{d_c} \rho_i = 1$$

$$\sum_{i=1}^{d_v} \lambda_i = 1$$

$$p_e \rightarrow 0 \text{ as } l \rightarrow \infty$$

by tuning

$$\rho(x), \lambda(x)$$

for given

$$s^*, d_v, d_c$$

3 - Sae-Yong Chung 01

Simplex Method and Differential Evolution

- George Dantzig 47

- Price 97

	Simplex method	Differential evolution
Type	Optimization algorithm for linear programming (LP) problems	Heuristic method (direct parallel search)
Requirement	Standard form of LP problems	System parameter
Solution	Global optimum	Close to Global optimum
Stopping criterion	Optimality of current basic solution is guaranteed	Fixed iteration number

Main Result : Proposed Scheme

Inputs :
Initial N vectors, $G = 0$, G_{max} , D

Optimize s^*
by using
Differential Evolution

(I) Differential Evolution

Pre-processing :
Find a good pair of $\rho(x)$, $\lambda(x)$

$\rho(x)$, s^* (results of differential evolution)

Optimize R
for given $\rho(x)$

$\lambda(x)$

Optimize R
for given $\lambda(x)$

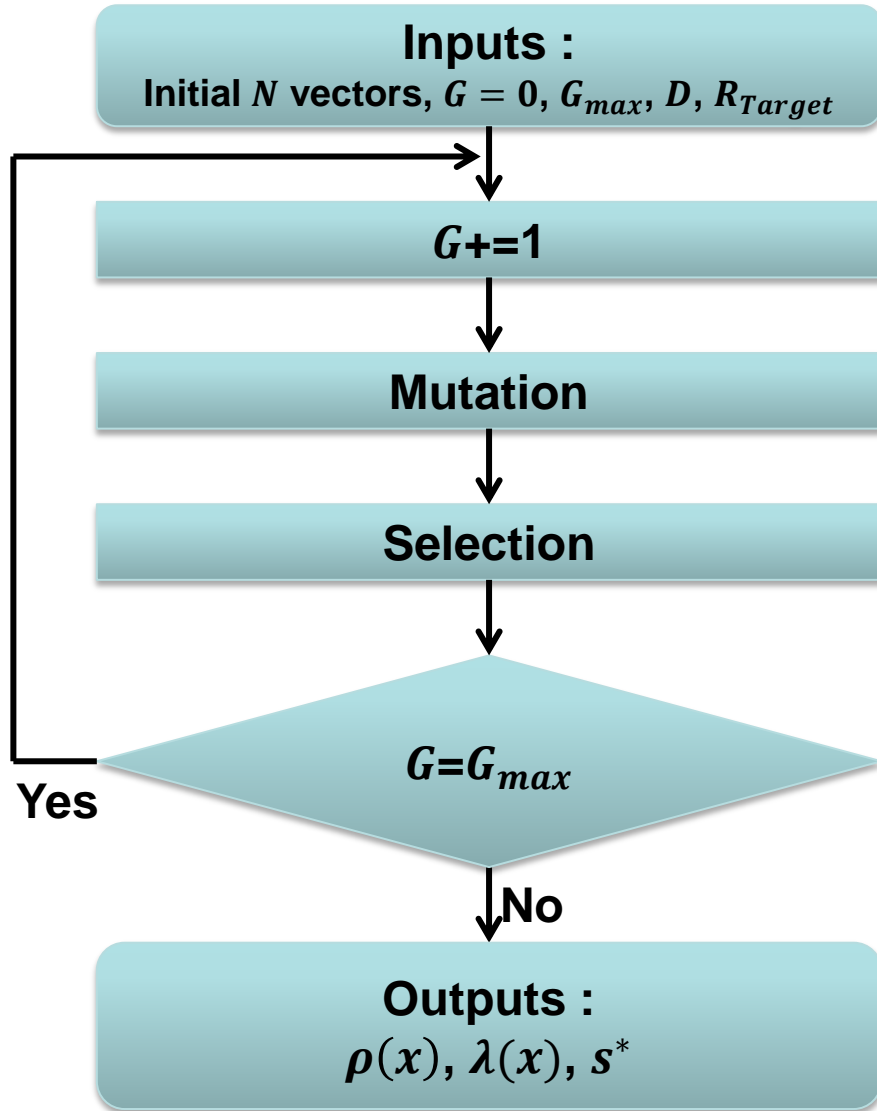
$\rho(x)$

(II) Iterative Simplex Algorithm

Post-processing :
Enhance code rate

Outputs :
 $\rho_{opt}(x)$, $\lambda_{opt}(x)$

(I) Differential Evolution



Initial N vectors :

$$x_i = \begin{bmatrix} \rho_{d_c-1,i}^0 \\ \rho_{d_c,i}^0 \\ \lambda_{1,i}^0 \\ \vdots \\ \lambda_{d_v,i}^0 \end{bmatrix}, 0 \leq i \leq N$$

Mutation :

$$v_i^G = x_{best}^{G-1} + \sum_{k=1}^D (x_{r1,k}^{G-1} - x_{r2,k}^{G-1})$$

Controlled Parameters :

$$N, G_{max}, D$$

(I) Differential Evolution : Issues

- Implementation issue
 - $\phi(x)$ and $\phi^{-1}(x)$
 - Size of N
 - Diversity
 - Iteration number
 - Cumulative error
 - Threshold calculation

$$\phi(x) \triangleq \begin{cases} 1 - \frac{1}{\sqrt{4\pi x}} \int_R \tanh \frac{u}{2} e^{-\frac{(u-x)^2}{4x}} du, & \text{if } x > 0 \\ 1 & \text{if } x \leq 0 \end{cases},$$

(II) Iterative Simplex Algorithm : Applying LP Solver

Optimize s^*

Minimize s^*

subject to

$$\sum_{i=1}^{d_c} \rho_i = 1$$

$$\sum_{i=1}^{d_v} \lambda_i = 1$$

$$\begin{aligned} r &> h(s, r) \\ \lambda_2 &< \lambda_2^* \end{aligned}$$

by tuning

$\rho(x), \lambda(x)$

for given

R, d_v, d_c

Optimize R

Maximize R

subject to

$$\sum_{i=1}^{d_c} \rho_i = 1$$

$$\sum_{i=1}^{d_v} \lambda_i = 1$$

$$\begin{aligned} r &> h(s, r) \\ \lambda_2 &< \lambda_2^* \end{aligned}$$

by tuning

$\rho(x), \lambda(x)$

for given

s^*, d_v, d_c

The representation of s^* with $\rho(x)$ and $\lambda(x)$ is not known.

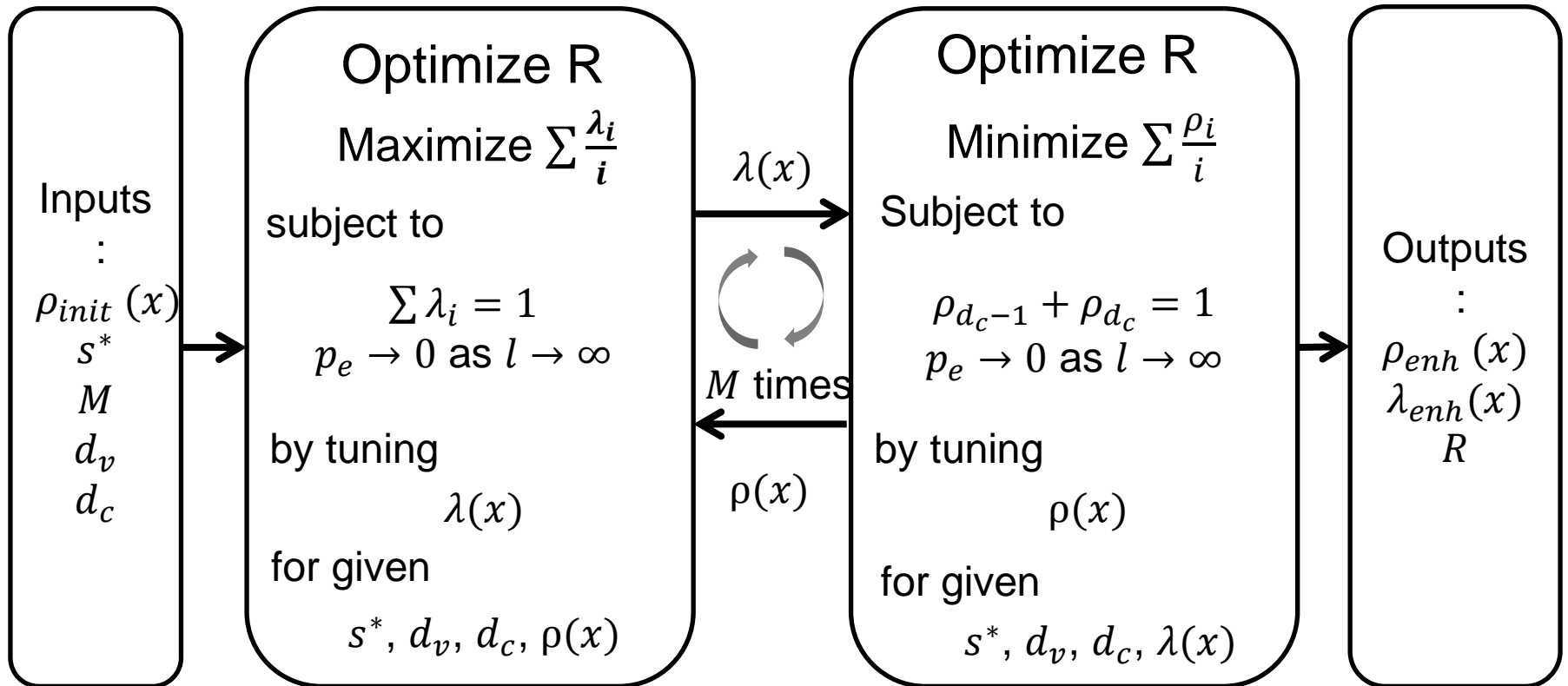
It is not linear program for controlled parameter $\rho(x), \lambda(x)$. But if one of them is given, then **it will be linear program.**

It is known that the $\rho(x) = \rho x^{k-1} + (1 - \rho)x^k$ is good enough to obtain good performance.

$p_e \rightarrow 0$ condition, by Gaussian approximation.

- Sae-Young Chung 01

(II) Iterative Simplex Algorithm



(II) Iterative Simplex Algorithm : Convergence

We can see that R converge to R_{opt} with some iterations by simulation.

- We can proof the convergence of R since

$$\begin{array}{ccc}
 \rho^{(i)}(x), \lambda^{(i)}(x) & \longrightarrow & \rho^{(i)}(x), \lambda^{(i+1)}(x) \\
 & \swarrow & \\
 \rho^{(i+1)}(x), \lambda^{(i+1)}(x) & \longrightarrow & \rho^{(i+1)}(x), \lambda^{(i+2)}(x) \\
 & \swarrow & \\
 \vdots & \longrightarrow & \vdots
 \end{array}
 \quad
 \begin{array}{l}
 \lambda^{(k)}(x) = \sum_{i=2}^{d_v} \lambda_i^{(k)} x^{i-1} \\
 \rho^{(l)}(x) = \sum_{i=2}^{d_c} \rho_i^{(l)} x^{i-1}
 \end{array}$$

If the LP solver guarantee that the solution is global optimum, then

$$\underline{R_{i,i+1} \geq R_{i,i} \geq R_{i-1,i}} \quad \text{where} \quad R_{i,j} = 1 - \frac{\sum_l \frac{\rho_l^{(i)}}{l}}{\sum_k \frac{\lambda_k^{(j)}}{k}}$$

It means that the sequence of $R_{i,j}$ is non-decreasing.

Since $R_{i,j}$ is non-decreasing and $R \leq 1$, $R_{i,j}$ will converge to R_{opt} .

Conclusion

- **We describe differential evolution scheme to find good degree distribution.**
 - It gives good degree distribution candidates.
- **And we propose an iterative code rate optimizer.**
 - The iterative code rate optimizer (i.e. iterative simplex algorithm) increases code rate for given $\left(\frac{E_s}{N_0}\right)^*$.
- **Combined LDPC code design scheme.**
 - By simulation, we can see that proposed LDPC code design scheme has better $\left(\frac{E_b}{N_0}\right)^*$ and code rate than differential evolution only.

Conclusion : Simulation Results

	d_v	λ_2	λ_3	λ_4	λ_5	λ_6	λ_7	λ_8	ρ_6	ρ_7	$\left(\frac{E_b}{N_0}\right)_{dB}$	R
Richardson	6	0.33241	0.24632	0.11014		0.31112			0.76611	0.23389	0.7997	0.50840
	8	0.30013	0.28395					0.41592	0.22919	0.77081	0.5778	0.50013
Diff. Evol. Only	6	0.31265	0.17269	0.10492	0.24427	0.16547			0.65951	0.34049	0.7836	0.49953
	8	0.30819	0.24250	0.00601	0.01471	0.00148	0.00856	0.41855	0.16636	0.83364	0.5628	0.49918
Proposed	6	0.34061	0.24400			0.41539			0.65951	0.34049	0.7288	0.50584
	8	0.30259	0.26274					0.43468	0.16636	0.83364	0.5619	0.49927