

<GEP formulation>

Table 1: The GEP formulation of some SDR methods

Method	A	B
SIR	$\text{cov}[E\{\mathbf{x} - E(\mathbf{x})\} y]$	$\Sigma_{\mathbf{x}}$
PFC	Σ_{fit}	$\Sigma_{\mathbf{x}}$
SAVE	$\Sigma_{\mathbf{x}}^{1/2} E[\{I - \text{cov}(\mathbf{z} y)\}^2] \Sigma_{\mathbf{x}}^{1/2}$, where $\mathbf{z} = \Sigma_{\mathbf{x}}^{-1/2}\{\mathbf{x} - E(\mathbf{x})\}$	$\Sigma_{\mathbf{x}}$
PHD(y-based)	$\Sigma_{\mathbf{x}}^{1/2} \Sigma_{y\mathbf{z}\mathbf{z}} \Sigma_{y\mathbf{z}\mathbf{z}} \Sigma_{\mathbf{x}}^{1/2}$, where $\Sigma_{y\mathbf{z}\mathbf{z}} = E[\{y - E(y)\}\mathbf{z}\mathbf{z}^T]$	$\Sigma_{\mathbf{x}}$
PHD(r-based)	$\Sigma_{\mathbf{x}}^{1/2} \Sigma_{\mathbf{r}\mathbf{z}\mathbf{z}} \Sigma_{\mathbf{r}\mathbf{z}\mathbf{z}} \Sigma_{\mathbf{x}}^{1/2}$, where $\Sigma_{\mathbf{r}\mathbf{z}\mathbf{z}} = E[\{y - E(y) - E(y\mathbf{z}^T)\mathbf{z}\}\mathbf{z}\mathbf{z}^T]$	$\Sigma_{\mathbf{x}}$
DR	$\Sigma_{\mathbf{x}}^{1/2} \{2E[E^2(\mathbf{z}\mathbf{z}^T y)] + 2E^2[E(\mathbf{z} y)E(\mathbf{z}^T y)] + 2E[E(\mathbf{z} y)E(\mathbf{z} y)]E[E(\mathbf{z} y)E(\mathbf{z}^T y)] - 2\mathbf{I}_p\} \Sigma_{\mathbf{x}}^{1/2}$	$\Sigma_{\mathbf{x}}$

<simulation 결과 1>

$$\mathbf{y} = \text{sign}(\mathbf{b}_1' \mathbf{X}) * \log(|\mathbf{b}_2' \mathbf{X} + 5|) + 0.2 * \mathbf{e}$$

p=20

- Case (i) : true beta의 0개수: 32

$$(i) \beta_1 = (1, 1, 1, 1, 0, \dots, 0)^T \text{ and } \beta_2 = (0, \dots, 0, 1, 1, 1, 1)^T;$$

method	p	corr1	corr2	mse1	mse2
dr	0.000	0.7698654	0.7831131	3.212332	2.104733
sparse dr	22.735	0.9446084	0.9686214	3.518986	2.545846

- Case(ii): true beta의 0개수: 32

$$(ii) \beta_1 = (1, 1, 0.1, 0.1, 0, \dots, 0)^T \text{ and } \beta_2 = (0, \dots, 0, 0.1, 0.1, 1, 1)^T$$

method	p	corr1	corr2	mse1	mse2
dr	0.0	0.8280320	0.7915093	3.198981	0.8406891
sparse dr	23.9	0.9728796	0.9764424	3.562182	0.5640522

- Case(iii): true beta의 0개수: 20

$$(iii) \beta_1 = (1, \dots, 1, 0, \dots, 0)^T \text{ and } \beta_2 = (0, \dots, 0, 1, \dots, 1)^T,$$

method	p	corr1	corr2	mse1	mse2
dr	0.000	0.7963774	0.8138976	3.567960	1.715391
sparse dr	16.115	0.9581149	0.9099131	3.986382	2.114001

모든 case에서 Correlation 결과가 그냥 dr일때보다 sparse dr일 때 많이 좋아졌음

<논문 s-sir simulation result>

		$\hat{\beta}_1$			$\hat{\beta}_2$			$(\hat{\beta}_1, \hat{\beta}_2)$
		NUM	COR	MSE	NUM	COR	MSE	VCC
Case (i)	SIR	0.000	0.926	1.604	0.000	0.911	1.245	0.934
	S-SIR	15.16	0.975	1.352	15.38	0.974	1.026	0.946
Case (ii)	SIR	0.000	0.884	0.551	0.000	0.856	0.544	0.932
	S-SIR	17.67	0.984	0.245	17.68	0.986	0.205	0.968
Case (iii)	SIR	0.000	0.916	4.793	0.000	0.942	4.168	0.917
	S-SIR	9.220	0.877	5.006	9.630	0.908	4.329	0.816

그러나, 모든 case에서 sir과 s-sir 결과가 zero의 개수가 true값에 더 가깝다. Case3의 경우, dr과 sparse dr의 결과가 더 좋다.

<simulation 결과 2>

$$y = \cos(2*b_1'X) - \cos(b_2'X) + 0.5*e$$

p=10, true beta의 0 개수: 18

$$\beta_1 = (1, 0, \dots, 0)^T \text{ and } \beta_2 = (0, 1, 0, \dots, 0)^T.$$

- n=100

method	p	corr1	corr2	mse1	mse2
dr	0.00	0.7783505	0.8714745	0.5937655	2.571295
sparse dr	8.65	0.8215246	0.9570768	0.5305380	3.022709

- n=200

method	p	corr1	corr2	mse1	mse2
dr	0.00	0.8224365	0.8978549	0.4947189	2.070118
sparse dr	10.52	0.9102280	0.9532808	0.2818287	1.984507

n=200일 때 결과가 더 좋음, dr보다 sparse dr의 결과가 훨씬 좋아짐

		$\hat{\beta}_1$			$\hat{\beta}_2$			$(\hat{\beta}_1, \hat{\beta}_2)$
		NUM	COR	MSE	NUM	COR	MSE	VCC
$n = 100$	PHD	0.000	0.793	0.437	0.000	0.726	0.406	0.651
	S-PHD	8.280	0.911	0.194	8.140	0.902	0.158	0.837
$n = 200$	PHD	0.000	0.894	0.184	0.000	0.933	0.150	0.848
	S-PHD	8.360	0.970	0.055	8.230	0.991	0.020	0.963

그러나 s-phd 결과가 훨씬 좋음