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KLASIFIKASI JENIS TUMOR OTAK BERDASARKAN CITRA MRI
MENGGUNAKAN *CONVOLUTIONAL NEURAL NETWORK* MODEL
VGG-19 & ResNet-50

Oleh

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BAB I

PENDAHULUAN

I. Latar Belakang Masalah

Identifikasi citra medis merupakan salah satu dari banyak alasan dari pengembangan teknologi *neural network* yang telah banyak berkembang selama beberapa tahun terakhir. Banyak arsitektur *neural network* yang muncul dengan fokus dan lapisan yang berbeda untuk menyelesaikan permasalahan-permasalahan berbeda yang spesifik dan unik untuk masing-masing masalahnya. Salah satunya adalah untuk menganalisa citra medis MRI yang banyak digunakan untuk mendiagnosa penyakit atau kelainan pada otak.

Citra MRI dari otak pasien ini kemudian banyak digunakan oleh para dokter untuk membantu mendiagnosa penyakit apa yang sedang diderita oleh pasien yang dengan demikian, dokter dapat menentukan perawatan apa yang diperlukan oleh pasien untuk dapat mengatasi penyakitnya. Namun, citra MRI yang merupakan gambar dengan resolusi yang cukup rendah dan pencitraan yang kurang baik (karena bukan merupakan pencitraan dari refleksi cahaya, tetapi dengan menggunakan medan magnet), sehingga melakukan diagnosa yang akurat menjadi tantangan tersendiri bagi para dokter.

Oleh sebab itu, munculah gagasan untuk menggunakan teknologi *neural network* untuk melakukan klasifikasi citra medis MRI dengan mengekstrak fitur yang dapat dikenali sebagai ciri-ciri dari suatu penyakit atau kelainan yang kemudian dapat membuat jaringan dapat mengenali citra tersebut secara lebih objektif. Banyak arsitektur yang dapat dimanfaatkan untuk aplikasi ini yang diantara lain adalah arsitektur VGGnet, VGG-19 dan arsitektur residual, ResNet-50.

II. Rumusan Masalah

Pada percobaan ini, akan dicobakan untuk mengolah *dataset* ^[LINK 1] mengenai identifikasi tumor otak yang diklasifikasikan menjadi 4 kategori (Glioma, Meningioma, Pituitary, dan Tanpa Tumor). Adapun arsitektur yang digunakan untuk mengolah *dataset* tersebut adalah arsitektur VGG-19 dan ResNet-50 yang kemudian dari kedua arsitektur tersebut akan dibandingkan performanya (dengan parameter pembanding adalah akurasi

validasi dalam proses *Training* dan akurasi *Testing*). Dimana dari hasil percobaan, diharapkan penguji dapat kemudian menentukan arsitektur *neural network* mana yang lebih cocok untuk digunakan dalam proses klasifikasi tumor otak berdasarkan *dataset*^[LINK 1] yang tersedia.

III. Batasan Masalah

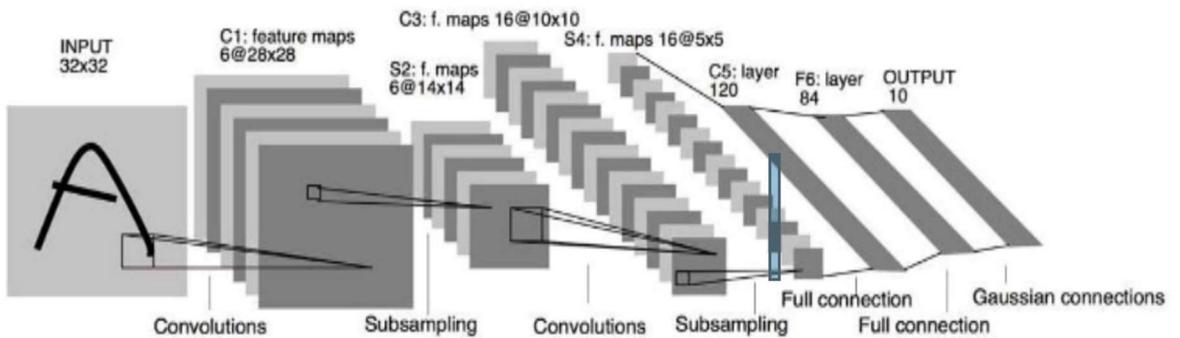
Pada percobaan ini untuk dapat menilai perbedaan performa antara arsitektur VGG-19 dan ResNet-50 secara lebih objektif, maka proses *Training* dan *Testing* akan dibatasi menggunakan konfigurasi layer VGG-19^[LINK 2] dan layer ResNet-50^[LINK 3] *default* tanpa adanya modifikasi lanjutan. Selain itu, kondisi dan parameter-parameter proses *Training* dan *Testing* juga akan dilaksanakan dalam kondisi dan konfigurasi yang sama persis, sehingga harapannya hasil yang keluar adalah murni karena perbedaan performa antara kedua arsitektur terhadap berdasarkan *dataset*^[LINK 1] yang tersedia.

BAB II

KAJIAN PUSTAKA DAN LANDASAN TEORI

I. Convolutional Neural Network (CNN)

Convolutional Neural Network (CNN) adalah salah satu metode pengolahan citra yang bekerja dengan cara membagi citra menjadi lapis RGB (*red*, *green*, dan *blue* sebagai representasi penyusun citra dari warna nya), dan kemudian dilakukan konvolusi dengan besar matriks konvolusi yang disesuaikan dengan besar *kernel* yang bekerja pada masing-masing lapis dengan tujuan untuk mengekstrak fitur, yang pada umumnya merupakan tipe dari suatu citra objek dengan tujuan untuk mengumpulkan fitur tersebut lalu menggabungkannya pada bagian akhir guna menyusun fitur yang telah diekstrak menjadi citra yang dapat dikenali untuk kemudian dapat mengklasifikasikan citra tersebut berdasarkan kelas/*label* yang telah didaftarkan sebelumnya.



Gambar 12 - Arsitektur Convolutional Neural Network (CNN) secara umum [1].

Adapun, berdasarkan **Gambar 1**, CNN terdiri atas beberapa lapis jaringan yang mendukung fungsi konvolusi sebagai kerangka utamanya. Dimana lapisan utamanya terdiri atas *Input Layer*, *Convolution Layer*, *Full Connection Layer*, dan *Classification Layer* sebagai lapis keluarannya. Sementara itu pada lapisan tersembunyi nya terdapat *Non-Linearity Layer* yang pada umumnya terdiri atas lapisan *Rectified Linear Units* (ReLU), yang deskripsinya masing-masing akan dijelaskan pada poin-poin berikut:

1. Convolution Layer

Lapis konvolusi merupakan lapis inti dari seluruh jaringan CNN, dimana pada lapis ini citra yang kompleks di pecah menjadi sampel yang lebih kecil sehingga kompleksitas citra dapat diminimalisir agar *Full Connected Layer* dapat mengumpulkan dan menghubungkan fitur-fitur yang sudah di ekstrak dari masing-masing sampel dengan lebih efektif [2,4].

2. Non-Linearity Layer

Pada jaringan yang memiliki banyak lapisan, agar koneksi antara masing-masing lapis konvolusi dapat bekerja dan terhubung dengan baik, diperlukan lapisan non-linear (dengan ReLU yang paling umum dipakai) untuk dapat membedakan lapisan, kernel, dan besar subsampel satu dengan yang lain (karena pada umumnya ada noda beberapa lapisan dengan konfigurasi yang sama). Sehingga lapis yang selanjutnya dapat melanjutkan proses konvolusi berdasarkan besar subsampel yang terdaftar pada lapis yang sebelumnya [2].

3. Pooling Layer

Setelah seluruh proses konvolusi telah berhasil dilakukan, maka tahap selanjutnya adalah mengumpulkan dan menyatukan seluruh *feature maps* yang telah dibuat menjadi sebuah output untuk mengurangi parameter dan ukuran spasial keluaran untuk memudahkan proses komputasi yang juga akan berguna untuk mengendalikan proses *overfitting* [3].

Selain dari ketiga lapis proses yang sudah disebutkan, ada pula lapis masukan yang berfungsi untuk menyesuaikan dimensi citra yang ingin diproses agar sesuai dengan kompatibilitas jaringan, dan lapis keluaran yang berupa lapis klasifikasi yang berguna untuk melakukan klasifikasi dengan menggunakan koneksi gaussian, dimana hasil bacaan akan disimpulkan menjadi peluang atas kemungkinan bahwa citra yang diklasifikasi lebih memungkinkan untuk masuk ke dalam kelas/label yang mana, berdasarkan daftar label yang telah terdaftar sebelumnya [2].

Dari arsitektur dasar CNN yang sudah dijelaskan, kemudian para pengembang muncul dengan berbagai pengembangan baru dengan karakteristik-karakteristik khusus

yang diharapkan dapat menunjang proses pengolahan citra dan aplikasi *neural network* secara umum. Dari antara banyak model pengembangan baru, beberapa diantaranya adalah arsitektur *Very Deep Learning*, VGGnet, dan *Residual Learning*, ResNet.

II. *Very Deep Learning* (VGG-net)

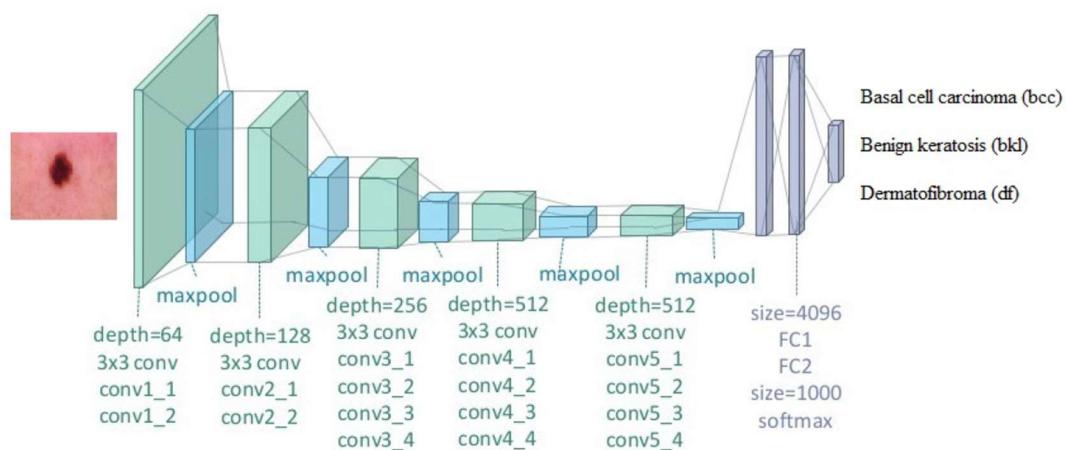
Metode pembelajaran jaringan *Very Deep Learning* adalah metode yang serupa dengan CNN yang sudah dibahas, yang menjadi pembeda adalah besar *kernel* yang dibuat sekecil mungkin (3×3) dan dengan harapan bahwa semakin kecil *kernel*, maka kedalaman hasil konvolusi akan semakin dalam, yang dengan demikian meningkatkan resolusi pembacaan fitur yang akan meningkatkan akurasi bacaan/klasifikasi [5].

Pooling Layer juga dibuat dengan jendela yang kecil (2×2) dengan *stride* sebesar 2 untuk meningkatkan resolusi subsampel dengan harapan bahwa *error* yang akan muncul akan berkurang. Selain itu juga dapat ditambahkan *Linearity Layar* dengan jendela sebesar 1×1 pixels untuk dapat meningkatkan *non-linearity* tanpa mengubah komposisi keluaran (meningkatkan efektivitas lapis ReLU) [5]. Adapun konfigurasi lapisan jaringan nya secara umum dapat dilihat pada **Tabel 1**. Sementara itu pada pengujian pelatihan pada [5], didapatkan dari pengujian 11 sampai 19 lapis VGG-net, VGG-19 merupakan model dengan performa terbaik (dimana dapat disimpulkan adalah semakin banyak lapis, maka kompleksitas akan semakin tinggi dan akurasi akan meningkat, namun perlu diketahui bahwa ada batasan tertentu dalam menentukan jumlah lapisan dalam jaringan, karena apabila jumlah lapisan terlalu banyak dan kompleksitas menjadi tidak terkendali, maka akan muncul masalah kehilangan gradien yang akan menyebabkan pada divergen nya akurasi dan *loss*).

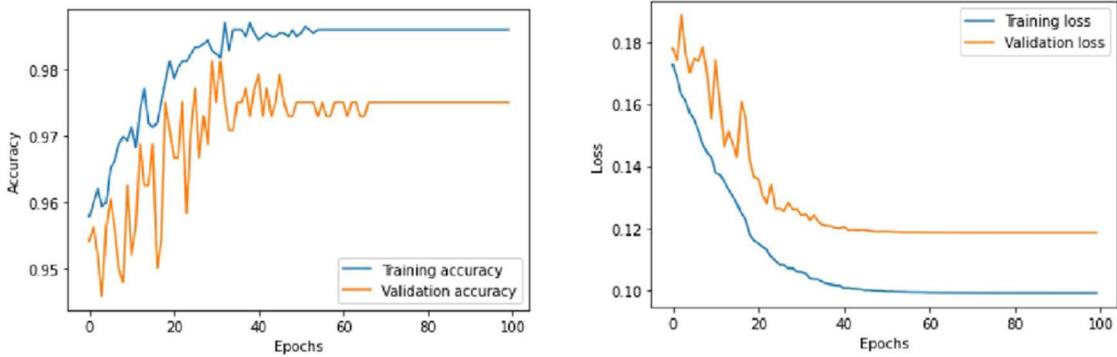
Sementara itu, konfigurasi lapisan jaringan untuk model VGG-19 secara spesifik dapat dilihat pada **Gambar 2**. Sementara itu, hasil pengujian pada proses pelatihan dan validasi dengan menggunakan *dataset* HAM1000 menghasilkan akurasi keluaran yang cukup baik pada percobaan dengan 100 *epoch*. Adapun grafik pada **Gambar 3**, menunjukkan bahwa akurasi sudah konvergen pada *epoch* ke-50, dan tidak menunjukkan perubahan yang cukup berarti hingga *epoch* ke-100. Namun dari hasil akurasi validasi pada **Tabel 2** masih belum menunjukkan tanda-tanda *overfitting*. Ditunjukkan dengan akurasi pelatihan dan validasi yang tidak jauh berbeda.

ConvNet Configuration					
A	A-LRN	B	C	D	E
11 weight layers	11 weight layers	13 weight layers	16 weight layers	16 weight layers	19 weight layers
input (224×224 RGB image)					
conv3-64	conv3-64 LRN	conv3-64 conv3-64	conv3-64 conv3-64	conv3-64 conv3-64	conv3-64 conv3-64
maxpool					
conv3-128	conv3-128	conv3-128 conv3-128	conv3-128 conv3-128	conv3-128 conv3-128	conv3-128 conv3-128
maxpool					
conv3-256 conv3-256	conv3-256 conv3-256	conv3-256 conv3-256	conv3-256 conv3-256 conv1-256	conv3-256 conv3-256 conv3-256	conv3-256 conv3-256 conv3-256 conv3-256
maxpool					
conv3-512 conv3-512	conv3-512 conv3-512	conv3-512 conv3-512	conv3-512 conv3-512 conv1-512	conv3-512 conv3-512 conv3-512	conv3-512 conv3-512 conv3-512 conv3-512
maxpool					
conv3-512 conv3-512	conv3-512 conv3-512	conv3-512 conv3-512	conv3-512 conv3-512 conv1-512	conv3-512 conv3-512 conv3-512	conv3-512 conv3-512 conv3-512 conv3-512
maxpool					
FC-4096					
FC-4096					
FC-1000					
soft-max					

Tabel 6 - Konfigurasi Umum Lapisan Jaringan VGG-net [5].



Gambar 13 - Konfigurasi Arsitektur VGG-19 [6].



Gambar 14 - Perubahan nilai Training dan Validation Accuracy (kiri) dan Loss (kanan) [6].

Epoch	Training		Validation	
	Accuracy	Loss	Accuracy	Loss
25	0.9823	0.1094	0.9708	0.1264
50	0.9849	0.0997	0.9750	0.1188
100	0.9859	0.0991	0.9750	0.1185

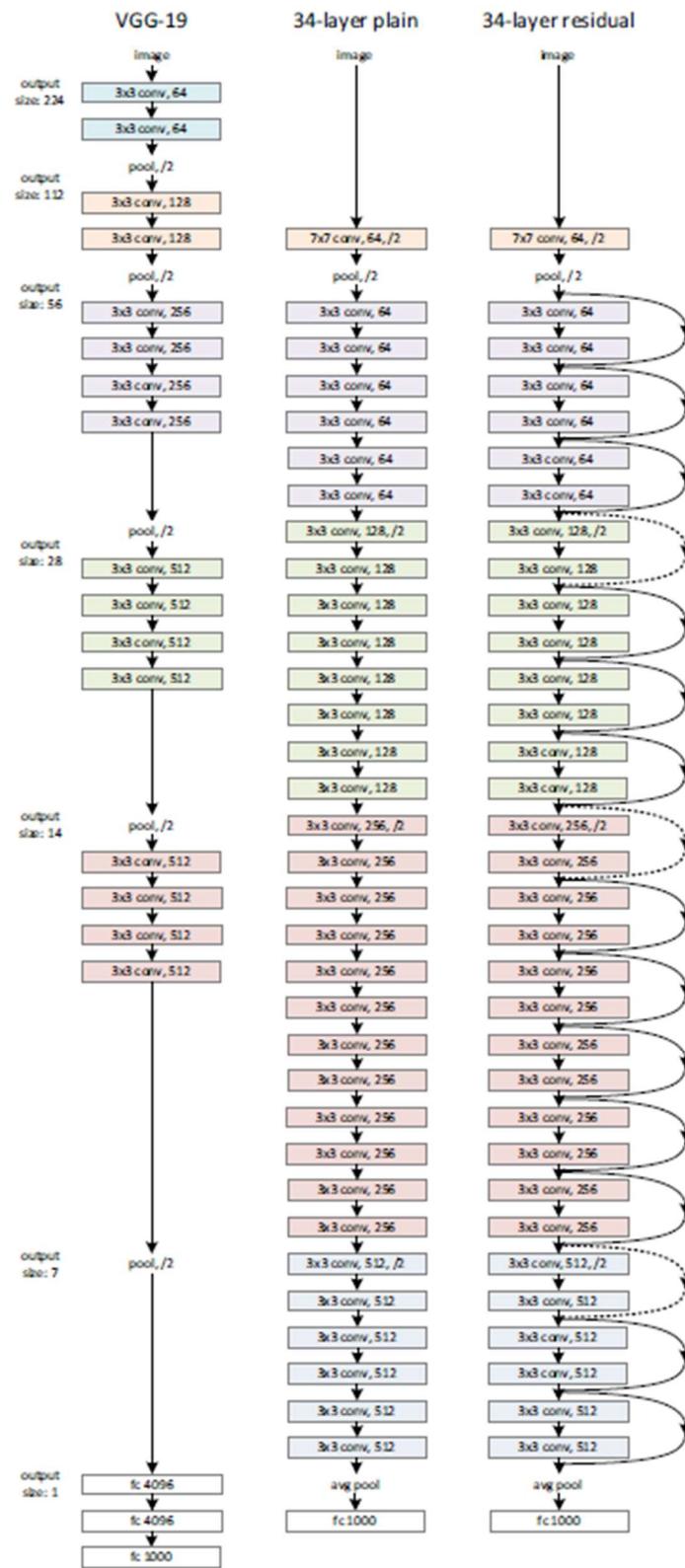
Tabel 7 - Catatan Akurasi VGG-19 [6].

III. Residual Deep Learning (ResNet)

Jika pada VGG-net atau arsitektur CNN biasa, pengguna memiliki keterbatasan untuk menentukan jumlah lapisan dari arsitektur yang mereka rancang karena masalah *vanishing gradient* dan subsampel yang terlalu dalam dengan fitur yang tidak cukup jelas untuk dianalisis yang dapat menyebabkan divergensi nilai bacaan akurasi pada jaringan dengan lapis berlebih, yang akan berakibat pada turunnya akurasi [7].

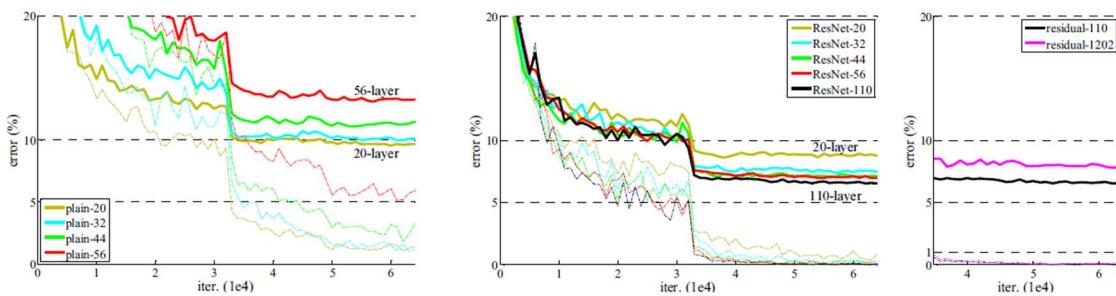
Untuk mengatasi permasalahan tersebut, ResNet di rancang untuk memanfaatkan *Residual Mapping* alih-alih *Original Mapping* yang umumnya dipakai sebagai parameter pembacaan fitur pada proses klasifikasi citra, dengan anggapan bahwa *Residual Mapping* sudah mengandung informasi yang cukup bagi model untuk dapat dimanfaatkan untuk mengkategorikan citra ke dalam kelas yang sudah terdaftar.

Pada arsitektur ResNet, digunakan *shortcut connection* yang dapat digunakan untuk melewati beberapa lapis tanpa perlu diproses lebih lanjut apabila dirasa hasil bacaan sudah cukup untuk dapat dikenali tanpa perlu proses yang lebih detail. Dengan demikian, meskipun memiliki konfigurasi lapisan yang lebih banyak, beban komputasi dapat dikurangi untuk mengurangi waktu pelatihan/validasi.



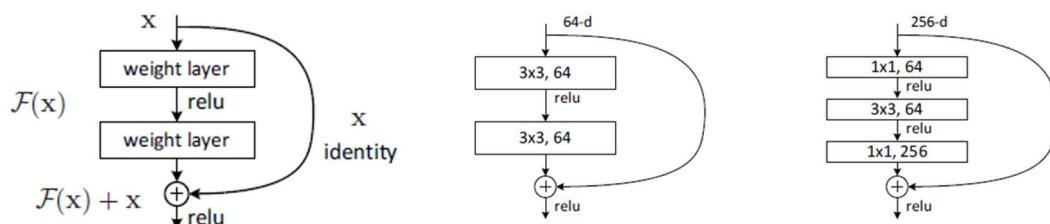
Gambar 15 - Perbandingan Konfigurasi antara arsitektur VGG-19 (kiri), CNN 34 Lapis (tengah), dan ResNet-34 (kanan) [8].

Dengan memanfaatkan *shortcut connection* ini, jaringan dapat menjaga konvergensi akurasinya meskipun membuat lapisan jaringan menjadi sebanyak mungkin (dapat dilihat pada **Gambar 5**). Dan dengan demikian, meskipun jaringan residual yang dirancang memiliki jumlah lapisan yang jauh lebih banyak jika dibandingkan dengan model-model lainnya, namun kestabilan akurasi nya masih dapat terjaga dengan baik dan terhindar dari masalah *vanishing gradient*. Dengan *shortcut connection* ini pula, arsitektur jaringan residual memiliki proses yang lebih cepat karena memang memiliki jumlah filter yang lebih sedikit [8].



Gambar 16 - Perbandingan nilai error terhadap jumlah lapisan pada jaringan CNN biasa (kiri), ResNet (tengah), dan percobaan pada jumlah lapisan yang ekstrim (1202) pada jaringan residual (kanan) [8].

Sementara itu, untuk arsitektur ResNet dengan jumlah lapisan 50 dan selebihnya, alih-alih menggunakan blok residual biasa, diaplikasikanlah *Bottleneck Building Block* yang serupa dengan blok residual biasa, namun dengan tambahan blok konvolusional 1×1 yang berguna untuk membuat proyeksi linear dari *Feature Maps* yang ditumpuk (menjadi *depth*) dengan tujuan untuk mengurangi (atau mengembalikan) dimensi, atau untuk menambahkan jumlah *Feature Maps* [8]. Adapun hasil pengujian [8] menunjukkan bahwa arsitektur yang menggunakan blok *Bottleneck* menampilkan performa yang lebih baik seperti yang dapat dilihat pada **Tabel 3**.



Gambar 17 - Blok Residual Biasa (kiri), beserta konfigurasi lapis konvolusinya (tengah), dan Blok Bottleneck untuk konfigurasi jaringan ResNet-50/101/152 (kanan) [8].

method	top-1 err.	top-5 err.
VGG [40] (ILSVRC'14)	-	8.43 [†]
GoogLeNet [43] (ILSVRC'14)	-	7.89
VGG [40] (v5)	24.4	7.1
PReLU-net [12]	21.59	5.71
BN-inception [16]	21.99	5.81
ResNet-34 B	21.84	5.71
ResNet-34 C	21.53	5.60
ResNet-50	20.74	5.25
ResNet-101	19.87	4.60
ResNet-152	19.38	4.49

Tabel 8 - Error dari percobaan single-model menggunakan dataset ImageNet ^[8].

IV. Tumor Otak

Tumor Otak merupakan pertumbuhan sel asing secara tidak terkendali, yang terbentuk pada bagian tertentu pada otak. Adapun pertumbuhan sel tersebut dapat terjadi secara alami, atau merupakan *metastasis*, atau penyebaran sel kanker dari organ lainnya yang juga memiliki sel kanker. Sedangkan untuk jenis-jenis dari tumor otak ini kemudian dibagi berdasarkan lokasi dan tingkat keganasannya ^[LINK 4].

Sel tumor dapat tumbuh di hampir seluruh bagian pada otak, namun mayoritas kasus tumor otak yang serius (primer) terjadi pada sel glial, atau sel yang menghubungkan sel saraf pada otak. Sedangkan untuk tingkat keganasannya, tumor otak dibedakan menjadi jinak (*benign*) atau ganas (*malignant*). Dimana pada kasus tumor jinak, sel kanker cenderung tidak menyebar, namun harus tetap diawasi secara berkala. Sedangkan pada kasus tumor ganas, pasien harus segera mendapat perawatan yang intensif untuk mencegah persebaran sel kanker dan mengatasi komplikasi yang mungkin akan terjadi ^[LINK 4].

Adapun beberapa jenis tumor otak yang sering terjadi adalah: Meningioma, Adenoma pituitari, Neuroma akustik, Craniopharyngioma, Tumor kelenjar pineal, Tumor otak glioma, Limfoma sistem saraf pusat, dan Tumor otak metastasis ^[LINK 4].

V. Tumor Sel Glial (Glioma)

Glioma merupakan sel tumor yang mulai berkembang dari sel glial pada otak, yaitu sel yang menghubungkan saraf pada otak. Adapun dari tingkat keganasannya, WHO membaginya ke dalam 4 kategori, yaitu: WHO grade I (potensi proliferasi rendah, kurabilitas cukup baik), WHO grade II (infiltratif, mitosis rendah, namun kerap timbul rekurensi/kambuhan), WHO grade III (mitosis jelas, infiltrasi tinggi, terdapat anaplasia), WHO grade IV (mitosis aktif, nekrosis, progresivitas penyakit cepat) [11].

Sedangkan, secara histologik, Tumor Sel Glial terbagi menjadi beberapa jenis, yaitu: *Astrocytic tumours*, *Oligodendrial tumours*, *Oligoastrocytic tumours*, *Ependymal tumours*, *Choroid plexus tumour*, *Other neuroepithelial tumours*, *Neuronal and mixed neuronal-glial tumours*, *Tumours of the pineal region*, dan *Embryonal tumours* [11].

VI. Meningioma

Meningioma merupakan tumor jinak intrakranial yang sering dijumpai pada pasien tumor otak dengan estimasi 13-26% dari keseluruhan tumor primer intrakranial dan berasal dari *arachnoid cap cells* durameter. Pada umumnya tidak berbahaya (paling umum, untuk grade I). Gambaran lesi Meningioma umumnya jelas, namun pada kasus khusus yang tumbuh pada *sphenoid ridge* atau disebut sebagai *meningioma en plaque*, gambaran lesi difus/tidak terlalu jelas. Meskipun umumnya jinak, namun pada grade II, bisa terjadi rekurensi dan pada grade III (jarang terjadi), bersifat papiler, rhabdoid, dan anaplastik. Meningioma grade III diklasifikasikan sebagai Meningioma Malignant [11].

VII. Adenoma Pituitari

Ademoma pituitari adalah tumor otak yang tumbuh pada kelenjar pituitari. Kelenjar pituitary sendiri adalah kelenjar yang berfungsi untuk melepaskan hormon ke dalam aliran darah. Tumor jenis ini pada umumnya jinak, namun dapat menjadi serius ketika timbul komplikasi yang berhubungan dengan kelainan hormon. Untuk mengatasi kelainan hormon akibat tumor jenis ini, pasien dapat disuplai dengan obat pengganti hormone, dan dapat disembuhkan dengan prosedur radioterapi [LINK 4].

BAB III

METODOLOGI

I. Spesifikasi Perangkat yang Digunakan

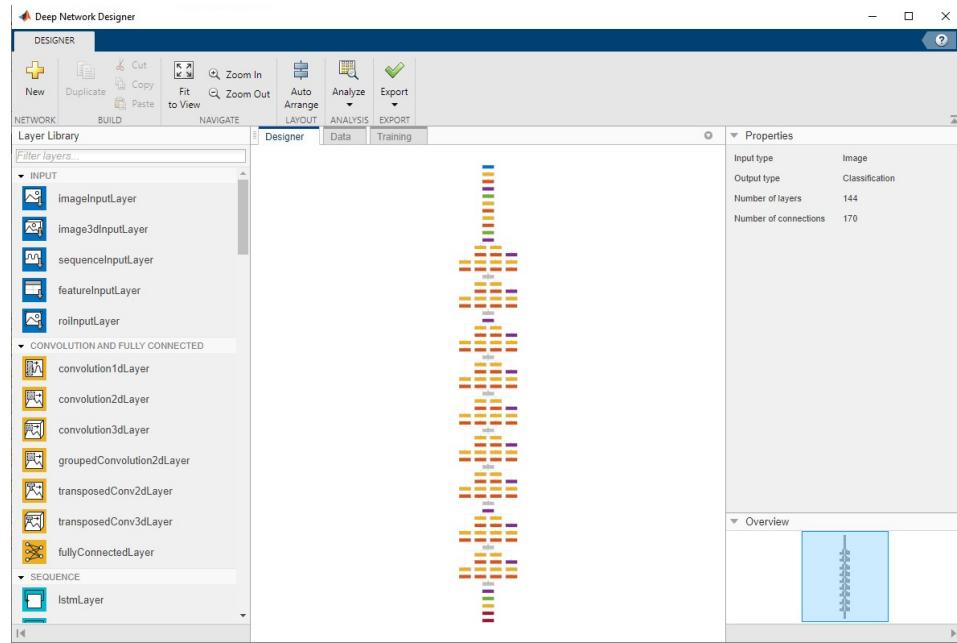
Pengujian untuk melatih dan menggunakan *neural network* dengan arsitektur VGG-19 dan ResNet-50, seluruhnya dilakukan pada perangkat *Laptop* Lenovo YOGA C640 dengan spesifikasi lengkap dapat dilihat pada tautan berikut [LINK 5]. Adapun perangkat bergerak dengan sistem operasi Windows 10 dengan aplikasi MATLAB versi 2019b terpasang di dalamnya sebagai aplikasi pemroses utama yang akan dipakai dalam pengujian ini.

II. Aplikasi Pemrograman yang Digunakan

Pengujian pelatihan dan validasi penggunaan *neural network* ini akan dilaksanakan menggunakan *platform* MATLAB 2019b, khususnya dengan menggunakan *Deep Network Designer Toolbox* yang tersedia pada aplikasi MATLAB [LINK 6] sebagai instrumen yang digunakan untuk menyusun lapisan *neural network* untuk kemudian dapat dilatih untuk mengerjakan tugas tertentu. *Deep Network Designer Toolbox* dapat diakses dengan menggunakan perintah `deepNetworkDesigner` pada *console*.

Adapun untuk MATLAB 2019b, karena *Deep Network Designer Toolbox* hanya dapat digunakan untuk merancang, namun tidak dapat mensimulasikan kerja jaringan secara langsung, maka akan direkomendasikan untuk meng-*generate equivalent code* nya untuk kemudian dijalankan guna menghasilkan mencatatkan jaringan pada *Workspace* di MATLAB (*source code* dapat dilihat pada [LAMPIRAN 3] untuk VGG-19 dan [LAMPIRAN 6] untuk ResNet-50, atau bisa juga dilakukan secara langsung dengan menekan tombol *Export* lalu memilih untuk *export to workspace*, untuk mencatatkan jaringan di *Workspace*).

Setelah *toolbox* berhasil terbuka, maka pengguna dapat menyusun lapisan jaringan menggunakan pilihan jenis jaringan yang tersedia pada *Layer Library* yang dapat diakses pada *toolbox* bagian kiri, seperti yang dapat dilihat pada **Gambar 7**. Dari masing-masing lapisan kemudian dapat dihubungkan satu sama lain dengan menarik garis pada bagian bawah *layer* dan dihubungkan pada bagian atas *layer* yang selanjutnya.



Gambar 18 - Contoh tampilan Deep Network Designer pada MATLAB [LINK 6].

Setelah *neural network* yang diinginkan berhasil disusun dan masing-masing lapisannya telah dihubungkan berdasarkan [LINK 2, LAMPIRAN 1] untuk arsitektur VGG-19, dan berdasarkan [LINK 3, LAMPIRAN 2] untuk arsitektur ResNet-50 (dengan nilai parameter yang telah disesuaikan dengan konfigurasi umum untuk masing-masing arsitekturm, dan penyesuaian lebar keluaran untuk jumlah kelas yang telah terdaftar pada *dataset* yang digunakan [LINK 1]), maka tahap selanjutnya adalah untuk melakukan *Export* pada masing-masing arsitektur ke *Workspace* agar dapat diproses untuk dilatih.

III. Langkah Pelatihan Neural Network

Jaringan *neural network* yang sudah di rancang dan ditempatkan pada *Workspace*, pada konsol MATLAB, kemudian dapat di latih dengan menggunakan data latih yang sudah disiapkan, dimana proses pelatihan nya dapat di picu berdasarkan *source code* yang tertera pada [LAMPIRAN 4] untuk jaringan VGG-19 dan [LAMPIRAN 7] untuk jaringan ResNet-50. Pada dasarnya keduanya memiliki fungsi dan runtutan perintah yang sama, dimana yang menjadi pembeda adalah nama variabel (jaringan) yang kemudian diproses untuk di latih. Adapun pelatihan dilakukan dengan tahapan sebagai berikut:

```

addpath('C:\Users\LENOVO\OneDrive\Desktop\Materi
Kuliah\Semester 7\Teknik Neurofuzi\Tugas\Deep Learning
Project\Program Files\VGG19');

unzip('Training.zip');

imds =
imageDatastore('Training','IncludeSubfolders',true,'LabelSo
urce','foldernames');

[imdsTrain,imdsValidation] =
splitEachLabel(imds,0.7,'randomized');

```

Pada bagian inisiasi awal, pengujii perlu untuk mendaftarkan lokasi dari *dataset* untuk mengakses berkas yang dimaksud. Untuk dapat melakukan ini, pengujii perlu menjalankan MATLAB sebagai *Administrator* untuk mendapatkan akses menggunakan perintah `addpath`. Setelah berkas berhasil dimuat dalam format ZIP, berkas kemudian dapat di ekstrak menggunakan perintah `unzip`, dimana berkas yang berhasil di ekstrak tersebut dapat dikategorikan berdasarkan lokasi folder yang akan di daftarkan sebagai kelas pada pembacaan direktori di MATLAB menggunakan perintah `imageDatastore`.

Pada proses pelatihan, *dataset* yang dibagi menjadi dua bagian untuk pelatihan itu sendiri dan untuk validasi hasil pelatihan. Pada *source code* yang tertera pada [LAMPIRAN 4, LAMPIRAN 7], ditentukan bahwa *dataset* akan dibagi dengan proporsi 70% untuk pelatihan dan 30% untuk validasi, dengan proses pemisahan secara acak menggunakan perintah `splitEachLabel`. *Dataset* yang berhasil dipisahkan, kemudian di daftarkan sebagai variabel baru dengan nama `imdsTrain` untuk *dataset* pelatihan dan `imdsValidation` untuk *dataset* validasi.

```

augimdsTrain = augmentedImageDatastore([224
224],imdsTrain);
augimdsValidation = augmentedImageDatastore([224
224],imdsValidation);

options = trainingOptions('sgdm', ...
'MiniBatchSize',10, ...
'MaxEpochs',6, ...
'Shuffle','every-epoch', ...
'InitialLearnRate',1e-4, ...
'ValidationData',augimdsValidation, ...
'ValidationFrequency',6, ...

```

```
'Verbose', false, ...
'Plots', 'training-progress');
```

Tahap selanjutnya adalah melakukan penyesuaian ukuran citra untuk disesuaikan dengan besar lapis masukan (224×224) dengan menggunakan perintah `augmentedImage datastore`, sementara itu konfigurasi kondisi pelatihan dapat dilakukan dengan menggunakan perintah `trainingOptions`. Perintah ini dilengkapi dengan beberapa sub-perintah yang dapat digunakan untuk menentukan nilai masing-masing dari parameter pelatihan yang dapat disesuaikan untuk menunjang proses pelatihan. Pada pelatihan ini, akan dilakukan sebanyak 6 *epoch* dengan masing-masing *epoch* terdiri atas 200 iterasi dan frekuensi validasi sebesar 6. Pengaturan ini diterapkan untuk melatih jaringan VGG-19 dan ResNet-50. Lingkungan pelatihan yang sama diterapkan untuk kedua model jaringan untuk dapat menghasilkan keluaran pembanding yang lebih obyektif. Adapun parameter pengaturan pelatihan yang sudah diterapkan kemudian didaftarkan dengan nama variabel `options` untuk dapat dipanggil kemudian, pada proses pelatihan.

```
netTransfer = trainNetwork(augimdsTrain, layers_1, options);
```

Pada tahap pelatihan yang terakhir, jendela pelatihan kemudian dibuka dan pelatihan akan dilaksanakan menggunakan perintah `trainNetwork`, berdasarkan *dataset* yang sudah didaftarkan pada variabel `augimdsTrain`, dijalankan dengan jaringan yang sudah terdaftar di *Workspace* dengan nama variabel `layers_1` untuk VGG-19 [LAMPIRAN 4] dan `lgraph` untuk ResNet-50 [LAMPIRAN 7], dengan konfigurasi parameter pelatihan yang terdaftar pada variabel `options`. Hasil pelatihan kemudian akan didaftarkan dengan nama variabel `netTransfer`, sebagai *DAGNetwork*. Dari jaringan yang sudah selesai dilatih, penguji dapat melihat nilai *weights* dan *bias* dari jaringan yang telah dilatih [LAMPIRAN 9, LAMPIRAN 10] dengan menggunakan perintah `netTransfer.Layers(2).Weights` dan `netTransfer.Layers(2).Bias`.

IV. Langkah Pengujian Neural Network

Jaringan yang sudah terlatih dan terdaftar dengan nama `netTransfer`, kemudian dapat digunakan untuk menjalankan tugas yang telah dirancang, yaitu untuk mengklasifikasikan jenis citra tumor otak yang tersimpan dalam *dataset* dengan nama *Testing*. Proses untuk memuat berkas *dataset* sama dengan apa yang dilakukan dalam proses pelatihan, yang menjadi pembeda hanyalah nama berkas dari yang tadinya *Training.zip*, menjadi *Testing.zip*. Dan alih-alih membagi *dataset* menjadi dua bagian, pada proses pengujian tidak ada pembagian *dataset* karena seluruh data akan digunakan untuk proses validasi, yang demikian juga berlaku untuk proses penyesuaian ukuran citra.

```
[YPred,probs] = classify(netTransfer,augimdsValidation);  
accuracy = mean(YPred == imdsValidation.Labels)
```

Sementara itu, untuk proses pengujian nya sendiri dilakukan dengan menggunakan perintah `classify`. Adapun keluaran dari dari fungsi tersebut didaftarkan kedalam 2 variabel, yaitu `YPred` yang berfungsi untuk menyatakan prediksi dari kelas atas citra yang telah di proses dan variabel `probs` yang menyatakan peluang/tingkat keyakinan atas prediksi yang diberikan. Sedangkan akurasi pengujian diukur dengan menghitung rerata dari keberhasilan prediksi kelas pada *dataset* dengan membandingkan kelas yang diprediksi (pada variabel `YPred`) dan kelas yang sebenarnya (pada variabel `imdsValidation.Labels`).

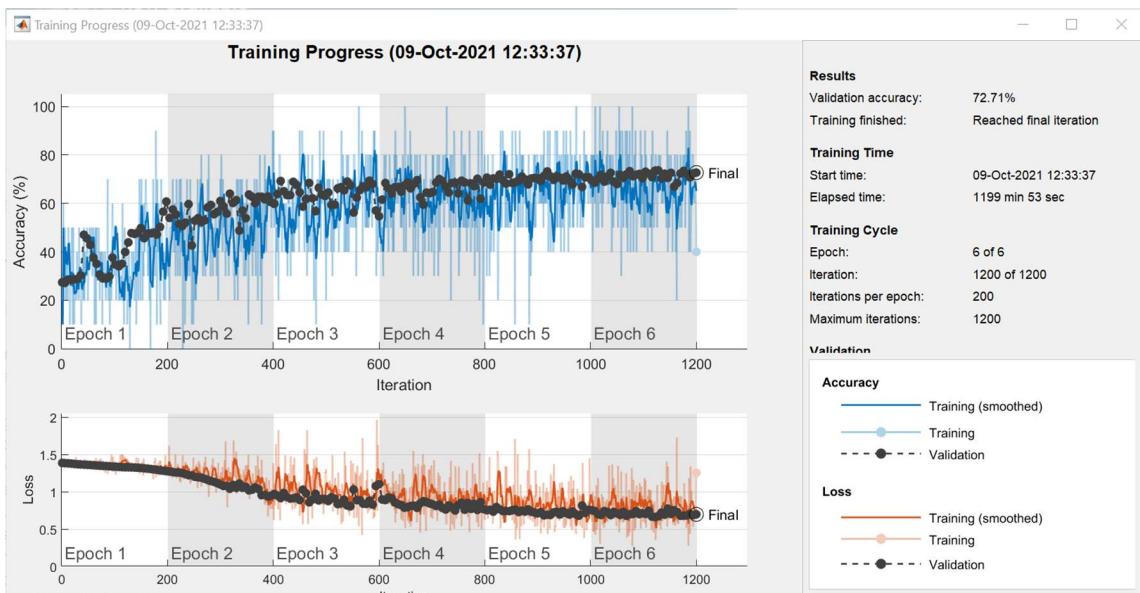
```
idx = randperm(numel(augimdsValidation.Files), 9);  
figure  
for i = 1:9  
    subplot(3,3,i)  
    I = readimage(imdsValidation,idx(i));  
    imshow(I)  
    label = YPred(idx(i));  
    title(string(label) + ", " +  
num2str(100*max(probs(idx(i),:)),3) + "%");  
end
```

Setelah mengukur nilai akurasi, meskipun opsional, penguji dapat mencetak beberapa sampel citra beserta dengan prediksi kelas dan tingkat keyakinan/peluang atas benarnya prediksi tersebut menggunakan perintah diatas [LAMPIRAN 5, LAMPIRAN 8].

BAB IV

HASIL DAN ANALISIS

I. Hasil Training & Testing Jaringan VGG-19

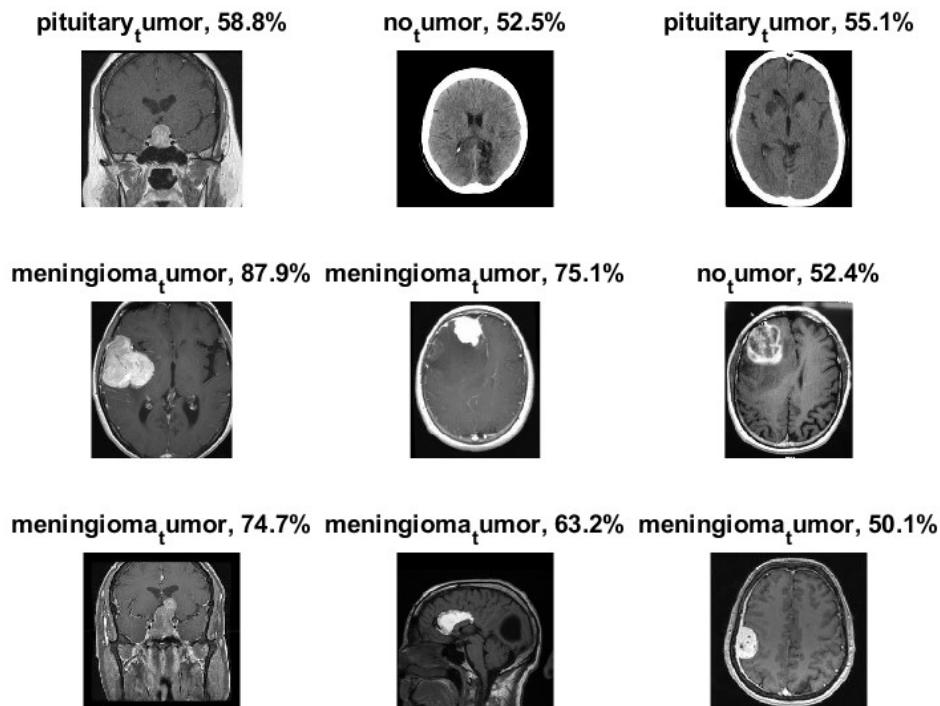


Gambar 19 - Hasil Training dan Validation Jaringan VGG-19.

Setelah berhasil melatih jaringan sebanyak 6 *epoch* dengan total mencapai 1200 iterasi dengan masa pelatihan selama 1199 menit dan 53 detik, dapat dilihat pada **Gambar 8**, bahwa sejauh hasil pengamatan, akurasi pelatihan dan validasi masih bergerak naik meskipun sudah mulai berkurang kenaikannya sejak memasuki pertengahan *epoch* ke-2. Dengan akurasi validasi tercatat pada iterasi terakhir sebesar 72,71%, namun jika melihat ke karakteristik kurva validasi, dapat diketahui bahwa hasil tersebut belum merupakan potensi maksimum dari model jaringan ini, mengingat bahwa kurva masih belum menunjukkan tanda-tanda konvergen meskipun kenaikannya sudah cukup melambat. Hal yang sama berlaku terbalik untuk nilai *loss* dari bacaan gambar diatas.

Adapun pada jaringan yang sudah berhasil dilatih, maka nilai *Weight* dan *Bias* dari jaringan berubah dari nilai asalnya, menjadi rangkaian nilai yang dapat di lihat pada [LAMPIRAN 9]. Nilai-nilai tersebut menyesuaikan dengan proses pelatihan, yang dalam kasus ini, dirancang untuk dapat mendekripsi/mengklasifikasikan citra MRI dari pasien tumor otak. Setelah jaringan dirasa sudah selesai di latih dan sudah siap untuk digunakan, maka

tahap selanjutnya adalah melakukan proses pengujian dengan menggunakan beberapa citra MRI pasien tumor otak dengan *dataset* yang berbeda dari *dataset* pelatihan, untuk proses ini, digunakan *dataset* terpisah yang di labeli dengan *dataset Testing*. Dari hasil pengujian ini, didapati bahwa rerata akurasi dari masing-masing pengujian tercatat sebesar 0.4086, atau 40,86%. Dari keseluruhan citra dalam *dataset* yang diujikan, berikut beberapa sampel pengujian yang dapat di lihat pada **Gambar 9**.

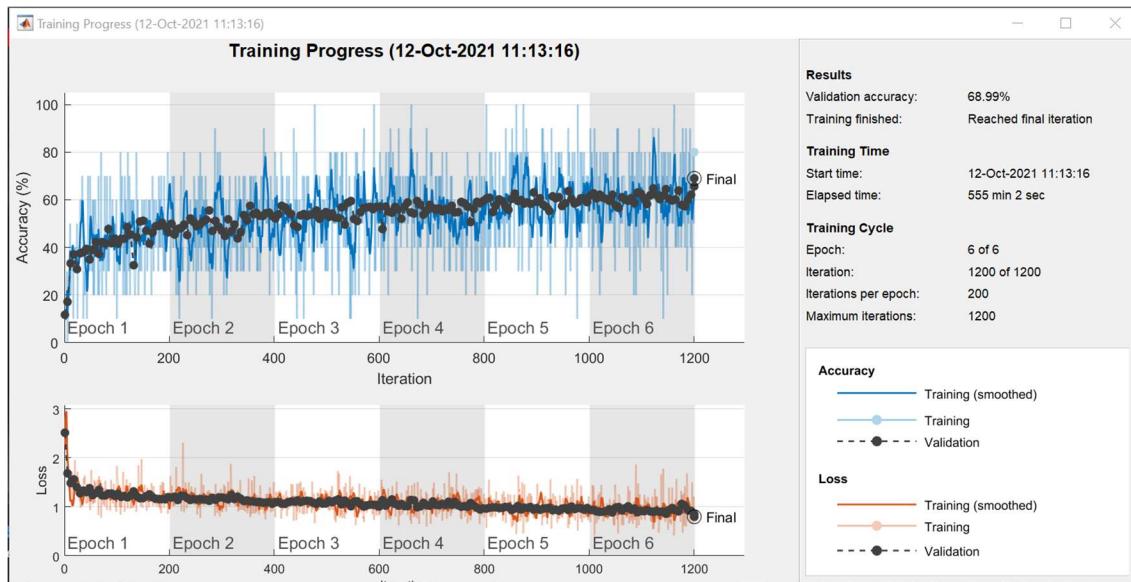


Gambar 20 - Beberapa sampel hasil pengujian/klasifikasi citra berdasarkan kelas yang terdaftar dan derajat kepastiannya (dalam bentuk nilai peluang) menggunakan jaringan VGG-19.

II. Hasil Training & Testing Jaringan ResNet-50

Dengan kondisi pelatihan yang sama dengan konfigurasi lingkungan pelatihan yang digunakan untuk melatih jaringan VGG-19, setelah berhasil dilatih selama 555 menit dan 2 detik, didapati bahwa pada iterasi terakhir, akurasi validasi tercatat sebesar 68,99%. Apabila diamati laju kenaikan akurasi dan penurunan *loss* bergerak secara linier, tercatat sejak pertengahan *epoch* 1, seperti yang dapat dilihat pada **Gambar 10**. Laju pergerakan

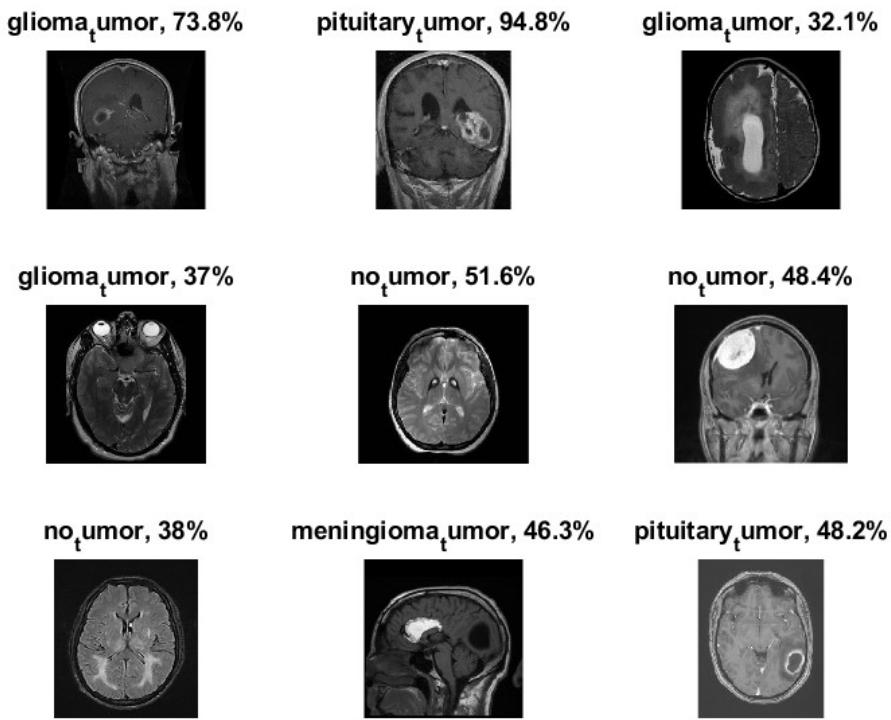
gradien kenaikan nilai akurasi juga belum terlihat tanda-tanda bahwa nilai akurasi akan segera konvergen. Sehingga dapat diestimasikan bahwa akurasi validasi yang tercatat belum dapat dikatakan sebagai nilai akurasi maksimal yang dapat dicapai oleh arsitektur ResNet-50 untuk *dataset* terkait [LINK 1].



Gambar 21 - Hasil Training dan Validation Jaringan ResNet-50.

Jaringan ResNet-50 yang dilatih dengan parameter yang ditunjukkan pada **Gambar 10**, Kemudian akan menghasilkan jaringan dengan nilai *Weight* dan *Bias* yang dapat di lihat pada [LAMPIRAN 10]. Dari jaringan ini, kemudian diujikan dengan *dataset Testing* yang sama dengan yang digunakan untuk pengujian menggunakan jaringan VGG-19, dimana mendapatkan hasil rata-rata akurasi pengujian sebesar 0.3447 atau 34.47%. Adapun beberapa sampel pengujian dapat dilihat pada **Gambar 11**.

Hasil pengujian menggunakan jaringan ResNet-50 ini terpantau masih sangat rendah, sama halnya dengan pengujian menggunakan jaringan VGG-19, meskipun masih tercatat menampilkan performa yang sedikit lebih baik. Adapun detail mengenai perbandingan performa antara kedua jaringan dan hasil analisis nya akan disampaikan pada bagian yang selanjutnya.



Gambar 22 - Beberapa sampel hasil pengujian/klasifikasi citra berdasarkan kelas yang terdaftar dan derajat kepastiannya (dalam bentuk nilai peluang) menggunakan jaringan ResNet-50.

III. Perbandingan Antara VGG-19 & ResNet-50

Setelah berhasil melatih dan menguji performa dari kedua jaringan *neural network*, penguji dapat kemudian membandingkan hasil dari kedua jaringan untuk menentukan jaringan mana yang lebih cocok untuk digunakan dalam memproses *dataset* yang berisi citra MRI pasien tumor otak [LINK 1]. Dari hasil akurasi validasi pada pelatihan jaringan VGG-19 dan ResNet-50 yang ditujukkan pada **Gambar 8** dan **Gambar 10** serta hasil rerata akurasi pada pengujian kedua jaringan yang telah disampaikan pada bagian sebelumnya, maka dapat dapat dikatakan bahwa VGG-19 memiliki performa keseluruhan yang lebih baik daripada ResNet-50 dalam lingkungan pelatihan yang terkontrol untuk kedua jaringan.

Dari akurasi validasi sampai rerata akurasi pengujian, didapatkan bahwa VGG-19 memiliki akurasi yang lebih tinggi dibanding ResNet-50. Untuk dapat mengkonfirmasi temuan ini, beberapa perbandingan performa dari VGG-19 dan ResNet-50 untuk citra

medis dari beberapa karya tulis yang dapat dilihat pada **Tabel 4** dan **Tabel 5** yang dikutip dari [9, 10]. Dari referensi yang dapat dilihat pada **Tabel 4**, diketahui dari akurasi keseluruhan untuk *dataset* COVID-19 yang dapat dilihat pada kolom *Pooled*, untuk *Batchsize* sebesar 16, ResNet-50 mengungguli VGG-19 sebesar 0.001, atau 0.1%. Sementara untuk *Batchsize* sebesar 32, diketahui bahwa baik untuk VGG-19 dan ResNet-50 memiliki akurasi yang sama sebesar 0.997 atau 99.7% [10].

Network	BS	COVID-19	No pneumonia	Non-COVID-19 pneumonia	Pooled
AlexNet	16	0.986	0.977	0.977	0.980
DenseNet-121	16	1.000	0.995	0.995	0.997
DenseNet-161	16	1.000	0.996	0.996	0.997
DenseNet-169	16	1.000	0.995	0.995	0.997
DenseNet-201	16	1.000	0.996	0.995	0.997
Inception v4	16	0.995	0.985	0.985	0.988
ResNet-18	16	1.000	0.992	0.992	0.995
ResNet-34	16	1.000	0.994	0.994	0.996
ResNet-50	16	1.000	0.995	0.994	0.996
ResNet-101	16	1.000	0.995	0.994	0.996
ResNet-152	16	1.000	0.996	0.996	0.997
SqueezeNet-1.0	16	0.995	0.980	0.980	0.985
SqueezeNet-1.1	16	0.995	0.979	0.978	0.984
VGG-13	16	0.999	0.991	0.991	0.994
VGG-16	16	1.000	0.993	0.993	0.995
VGG-19	16	0.999	0.993	0.993	0.995
AlexNet	32	0.994	0.978	0.978	0.983
DenseNet-121	32	1.000	0.996	0.996	0.997
DenseNet-169	32	1.000	0.997	0.997	0.998
DenseNet-201	32	1.000	0.997	0.997	0.998
Inception v4	32	0.995	0.987	0.987	0.990
ResNet-18	32	1.000	0.993	0.993	0.995
ResNet-34	32	1.000	0.995	0.995	0.997
ResNet-50	32	1.000	0.995	0.995	0.997
ResNet-101	32	1.000	0.996	0.996	0.997
ResNet-152	32	1.000	0.995	0.995	0.997
SqueezeNet-1.0	32	0.996	0.982	0.982	0.987
SqueezeNet-1.1	32	0.998	0.981	0.980	0.986
VGG-13	32	1.000	0.992	0.992	0.995
VGG-16	32	1.000	0.995	0.995	0.997
VGG-19	32	1.000	0.995	0.995	0.997

Tabel 9 - Perbandingan Akurasi dari beberapa metode Deep Learning untuk klasifikasi hasil radiograf pada dada pasien [10].

Dari perbandingan yang disajikan pada **Tabel 4**, kemudian dapat disimpulkan bahwa baik jaringan VGG-19 dan ResNet-50 memiliki perbedaan akurasi yang tidak cukup signifikan, mengingat waktu prosesnya yang sangat berbeda, dimana VGG-19 memakan jauh lebih banyak waktu. Sedangkan pada **Tabel 5**, menyatakan secara gamblang bahwa VGG-19 memiliki performa yang lebih baik dari ResNet-50, dengan penggunaan untuk mengklasifikasikan citra X-Ray.

Sehingga, dari hasil percobaan dan referensi yang sudah disebutkan sebelumnya [9, 10], dapat diambil titik tengah bahwa kecenderungan VGG-19 untuk menjadi *performer* yang lebih baik lebih dari ResNet-50. Namun mengingat waktu dan kebutuhan daya komputasi yang cukup besar untuk menjalankan, terlebih melatih jaringan VGG-19, ResNet-50 tetap dapat menjadi alternatif apabila proses pengujian terkendala waktu, atau mendapat *dataset* latih yang ternyata dapat bekerja dengan lebih baik untuk jaringan ResNet-50 (dimana perlu penelitian lanjutan terkait hal ini).

Metrics	VGG-19	ResNet-50	IykeNet
Accuracy (%)	97.3	96.2	93.60
Specificity (NPV) (%)	97.2	96.4	91.66
Precision (%)	96.7	95.3	91.30
Recall (%)	99.2	98.4	92.03

Tabel 10 - Perbandingan Akurasi antara VGG-19, ResNet-50, dan IykeNet untuk klasifikasi Pneumonia dengan menggunakan citra X-Ray [9].

IV. Hasil Analisa Keseluruhan

Pada awal proses analisa, hal pertama yang perlu dianalisis adalah hasil latihan dan akurasi validasi proses pelatihan dari kedua jaringan yang dapat dilihat pada **Gambar 8** dan **Gambar 10**. Dari kedua gambar tersebut, dapat dilihat bahwa untuk konfigurasi pelatihan yang sama, 6 *epoch* dengan masing-masing *epoch* terdiri atas 200 iterasi, didapatkan bahwa akurasi masih bergerak naik dan belum menunjukkan tanda-tanda akan konvergen ke nilai tertentu (meskipun kenaikan nya sudah tergolong lambat). Namun, yang membedakan adalah jika pada VGG-19, gradien akurasi mulai linier pada akhir *epoch* ke-

2, pada ResNet-50 akurasi mulai linier sejak pertengahan *epoch* pertama. Hal ini mengindikasikan bahwa *progress* yang ditempuh dengan arsitektur ResNet-50 lebih stabil dan linier. Hal ini tentunya dipengaruhi oleh penyelesaian masalah kehilangan gradien dengan menggunakan *shortcut connection* yang terdapat pada *Bottleneck Block* yang merupakan blok pembangun arsitektur ResNet-50.

Dengan demikian arsitektur jaringan residual memiliki kemampuan lebih untuk dapat menambah jumlah *layer* tanpa kehilangan gradien dan tetap menjadi kekonvergenan nilai akurasi nya, yang di sisi lain juga membantu untuk mencegah *overfitting*, yang dengan demikian, meskipun pada *dataset* dan jumlah iterasi yang sama VGG-19 mungkin lebih unggul dari ResNet-50, tetapi ResNet-50 memiliki kapabilitas latih yang lebih tinggi dari VGG-19. Berdasarkan informasi ini, dapat diestimasikan bahwa ResNet-50 dapat bekerja lebih baik dari VGG-19 apabila dilatih dengan *dataset* yang lebih besar atau dengan iterasi yang lebih banyak, dimana pada tahapan tersebut, VGG-19 lebih berisiko untuk mengalami *overfitting* dan kehilangan nilai akurasinya.

Di sisi lain, meskipun nilai akurasi validasi pada tahap pelatihan untuk kedua model arsitektur jaringan tergolong cukup baik dan deterministik, tetapi saat diujikan dengan *dataset Testing*, didapati akurasi pengujian cukup buruk di angka 40,86% untuk VGG-19 dan 34,47% untuk ResNet-50. Hal ini diperkirakan dapat terjadi karena citra MRI yang digunakan, meskipun sudah cukup mewakili keseluruhan kondisi yang memungkinkan dari pasien yang menderita tumor otak, tetapi membuat rancu dalam proses pengenalan fitur yang diterapkan pada *neural network*.

Neural network dapat mengenali sebuah citra dengan mengekstrak fitur dari sebuah citra dan membandingkan nya dengan contoh berdasarkan masing-masing kelas. Pada citra MRI Tumor Otak, fitur dari tumor otak itu sendiri berbentuk gumpalan berwarna cerah yang berbentuk acak, sehingga akan sulit untuk menentukan jenis tumor berdasarkan bentuknya, cara paling masuk akal untuk mengidentifikasi jenis tumor nya adalah dengan memanfaatkan lokasi dimana tumor tersebut berada, mengingat jenis tumor yang didaftarkan sebagai kelas klasifikasi memang dibedakan berdasarkan tempat tumbuhnya tumor tersebut (glioma pada sel glial, adenoma pituitary pada kelenjar pituitary, dan meningioma pada *arachnoid cap cells*).

Namun yang menjadi masalah adalah, dalam *dataset* ini ^[LINK 1], citra diambil dari berbagai sisi, sehingga identifikasi lokasi dari tumor bisa menjadi kacau karena adanya citra yang diambil dari sisi lain. Jika kita melihat pada referensi ^[9, 10], dapat diketahui bahwa kedua referensi tersebut menggunakan citra X-Ray bagian dada, yang sudah jelas hanya dapat diambil dari bagian depan, sehingga identifikasi lokasi fitur dapat dimanfaatkan dengan maksimal untuk menunjang proses klasifikasi.

Demikian adalah alasan mengapa akurasi pengujian didapati masih cukup rendah, untuk dapat menghasilkan akurasi yang tinggi dan deterministik, baiknya untuk menggunakan *dataset* dengan orientasi yang seragam, sehingga jaringan dapat berfokus pada fitur nya secara langsung, yang dalam kasus ini adalah tumor otak, dan tidak dibingungkan oleh orientasi citra yang berubah-ubah.

BAB V

KESIMPULAN

Dari hasil pelatihan, pengujian, serta analisis yang telah dilakukan untuk jaringan *Neural Network* VGG-19 dan ResNet-50 untuk *dataset* mengenai citra MRI dari pasien tumor otak ^[LINK]^{1]}, maka dapat disimpulkan bahwa dalam hal performa dasar (dengan perbandingan pengujian di lingkungan yang sama) VGG-19 dapat mengungguli ResNet-50 dalam akurasi klasifikasinya. Namun dapat dipertimbangkan kembali waktu proses dan pelatihan nya, karena VGG-19 memakan waktu lebih dari dua kali lebih lama dari ResNet-50 dengan kenaikan akurasi yang tidak terlalu signifikan.

Di sisi lain, ResNet-50 dapat bekerja dengan lebih baik apabila *dataset* yang digunakan lebih banyak, atau menggunakan sesi pelatihan yang lebih lama. Karena *shortcut connection* nya yang membuat jaringan ResNet-50 tidak rentan terhadap masalah *vanishing gradient* dan membutuhkan proses yang lebih banyak sampai akhirnya *overfitting* jika dibandingkan dengan jaringan VGG-19.

Sementara itu, penyebab dari kurangnya akurasi saat sesi pengujian dapat disebabkan oleh kondisi pelatihan yang belum cukup melatih jaringan sampai potensi maksimal nya. Atau dapat pula karena orientasi citra pada *dataset* yang berbeda-beda, sehingga proses klasifikasi berdasarkan lokasi tumor (fitur pada citra) menjadi terganggu, yang pada akhirnya menyebabkan nilai akurasi bacaan yang rendah. Sehingga untuk dapat menghasilkan akurasi klasifikasi yang lebih tinggi, akan lebih baik apabila citra yang digunakan merupakan hasil rekaman pada orientasi yang sama untuk semua citra yang akan digunakan dalam proses pelatihan dan proses klasifikasi atau pengujian nya itu sendiri.

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LAMPIRAN

LAMPIRAN 1 – Konfigurasi Layer VGG-19

1	imageinput 224x224x3 images with 'zeroc...	Image Input	224x224x3	-
2	conv_1 64 3x3x3 convolutions with stri...	Convolution	224x224x64	Weights 3x3x3x64 Bias 1x1x64
3	relu_1 ReLU	ReLU	224x224x64	-
4	conv_2 64 3x3x64 convolutions with st...	Convolution	224x224x64	Weights 3x3x64x64 Bias 1x1x64
5	relu_2 ReLU	ReLU	224x224x64	-
6	maxpool_1 2x2 max pooling with stride [2 ...	Max Pooling	112x112x64	-
7	conv_3 128 3x3x64 convolutions with ...	Convolution	112x112x128	Weights 3x3x64x128 Bias 1x1x128
8	relu_3 ReLU	ReLU	112x112x128	-
9	conv_4 128 3x3x128 convolutions with...	Convolution	112x112x128	Weights 3x3x128x128 Bias 1x1x128
10	relu_4 ReLU	ReLU	112x112x128	-
11	maxpool_2 2x2 max pooling with stride [2 ...	Max Pooling	56x56x128	-
12	conv_5 256 3x3x128 convolutions with...	Convolution	56x56x256	Weights 3x3x128x256 Bias 1x1x256
13	relu_5 ReLU	ReLU	56x56x256	-
14	conv_6 256 3x3x256 convolutions with...	Convolution	56x56x256	Weights 3x3x256x256 Bias 1x1x256
15	relu_6 ReLU	ReLU	56x56x256	-
16	conv_7 256 3x3x256 convolutions with...	Convolution	56x56x256	Weights 3x3x256x256 Bias 1x1x256
17	relu_7 ReLU	ReLU	56x56x256	-
18	conv_8 256 3x3x256 convolutions with...	Convolution	56x56x256	Weights 3x3x256x256 Bias 1x1x256
19	relu_8 ReLU	ReLU	56x56x256	-
20	maxpool_3 2x2 max pooling with stride [2 ...	Max Pooling	28x28x256	-

21	conv_9 512 3x3x256 convolutions with...	Convolution	28x28x512	Weights 3x3x256x512 Bias 1x1x512
22	relu_9 ReLU	ReLU	28x28x512	-
23	conv_10 512 3x3x512 convolutions with...	Convolution	28x28x512	Weights 3x3x512x512 Bias 1x1x512
24	relu_10 ReLU	ReLU	28x28x512	-
25	conv_11 512 3x3x512 convolutions with...	Convolution	28x28x512	Weights 3x3x512x512 Bias 1x1x512
26	relu_11 ReLU	ReLU	28x28x512	-
27	conv_12 512 3x3x512 convolutions with...	Convolution	28x28x512	Weights 3x3x512x512 Bias 1x1x512
28	relu_12 ReLU	ReLU	28x28x512	-
29	maxpool_4 2x2 max pooling with stride [2 ...	Max Pooling	14x14x512	-
30	conv_13 512 3x3x512 convolutions with...	Convolution	14x14x512	Weights 3x3x512x512 Bias 1x1x512
31	relu_13 ReLU	ReLU	14x14x512	-
32	conv_14 512 3x3x512 convolutions with...	Convolution	14x14x512	Weights 3x3x512x512 Bias 1x1x512
33	relu_14 ReLU	ReLU	14x14x512	-
34	conv_15 512 3x3x512 convolutions with...	Convolution	14x14x512	Weights 3x3x512x512 Bias 1x1x512
35	relu_15 ReLU	ReLU	14x14x512	-
36	conv_16 512 3x3x512 convolutions with...	Convolution	14x14x512	Weights 3x3x512x512 Bias 1x1x512
37	relu_16 ReLU	ReLU	14x14x512	-
38	maxpool_5 2x2 max pooling with stride [2 ...	Max Pooling	7x7x512	-
39	fc_1 4096 fully connected layer	Fully Connected	1x1x4096	Weights 4096x25088 Bias 4096x1
40	relu_17 ReLU	ReLU	1x1x4096	-

41	dropout_1 50% dropout	Dropout	$1 \times 1 \times 4096$	-
42	fc_2 4096 fully connec...	Fully Connected	$1 \times 1 \times 4096$	Weights 4096×4096 Bias 4096×1
43	relu_18 ReLU	ReLU	$1 \times 1 \times 4096$	-
44	dropout_2 50% dropout	Dropout	$1 \times 1 \times 4096$	-
45	fc_3 4 fully connecte...	Fully Connected	$1 \times 1 \times 4$	Weights 4×4096 Bias 4×1
46	softmax softmax	Softmax	$1 \times 1 \times 4$	-
47	classoutput crossentropyex	Classification Output	-	-

LAMPIRAN 2 – Konfigurasi Layer ResNet-50

1	input_1 224x224x3 ima...	Image Input	224x224x3	-
2	conv1 64 7x7x3 convol...	Convolution	112x112x64	Weights 7x7x3x64 Bias 1x1x64
3	bn_conv1 Batch normaliza...	Batch Normalization	112x112x64	Offset 1x1x64 Scale 1x1x64
4	activation_1_... ReLU	ReLU	112x112x64	-
5	max_pooling... 3x3 max poolin...	Max Pooling	55x55x64	-
6	res2a_branch... 64 1x1x64 conv...	Convolution	55x55x64	Weights 1x1x64x64 Bias 1x1x64
7	bn2a_branch2a Batch normaliza...	Batch Normalization	55x55x64	Offset 1x1x64 Scale 1x1x64
8	activation_2_... ReLU	ReLU	55x55x64	-
9	res2a_branch... 64 3x3x64 conv...	Convolution	55x55x64	Weights 3x3x64x64 Bias 1x1x64
10	bn2a_branch2b Batch normaliza...	Batch Normalization	55x55x64	Offset 1x1x64 Scale 1x1x64
11	activation_3_... ReLU	ReLU	55x55x64	-
12	res2a_branch... 256 1x1x64 con...	Convolution	55x55x256	Weights 1x1x64x256 Bias 1x1x256
13	bn2a_branch2c Batch normaliza...	Batch Normalization	55x55x256	Offset 1x1x256 Scale 1x1x256
14	res2a_branch1 256 1x1x64 con...	Convolution	55x55x256	Weights 1x1x64x256 Bias 1x1x256
15	bn2a_branch1 Batch normaliza...	Batch Normalization	55x55x256	Offset 1x1x256 Scale 1x1x256
16	add_1 Element-wise a...	Addition	55x55x256	-
17	activation_4_... ReLU	ReLU	55x55x256	-
18	res2b_branch... 64 1x1x256 con...	Convolution	55x55x64	Weights 1x1x256x64 Bias 1x1x64
19	bn2b_branch2a Batch normaliza...	Batch Normalization	55x55x64	Offset 1x1x64 Scale 1x1x64
20	activation_5_... ReLU	ReLU	55x55x64	-

21	res2b_branch2b 64 3x3x64 conv...	Convolution	55x55x64	Weights 3x3x64x64 Bias 1x1x64
22	bn2b_branch2b Batch normaliza...	Batch Normalization	55x55x64	Offset 1x1x64 Scale 1x1x64
23	activation_6_... ReLU	ReLU	55x55x64	-
24	res2b_branch2b 256 1x1x64 con...	Convolution	55x55x256	Weights 1x1x64x256 Bias 1x1x256
25	bn2b_branch2c Batch normaliza...	Batch Normalization	55x55x256	Offset 1x1x256 Scale 1x1x256
26	add_2 Element-wise a...	Addition	55x55x256	-
27	activation_7_... ReLU	ReLU	55x55x256	-
28	res2c_branch2b 64 1x1x256 con...	Convolution	55x55x64	Weights 1x1x256x64 Bias 1x1x64
29	bn2c_branch2a Batch normaliza...	Batch Normalization	55x55x64	Offset 1x1x64 Scale 1x1x64
30	activation_8_... ReLU	ReLU	55x55x64	-
31	res2c_branch2b 64 3x3x64 conv...	Convolution	55x55x64	Weights 3x3x64x64 Bias 1x1x64
32	bn2c_branch2b Batch normaliza...	Batch Normalization	55x55x64	Offset 1x1x64 Scale 1x1x64
33	activation_9_... ReLU	ReLU	55x55x64	-
34	res2c_branch2b 256 1x1x64 con...	Convolution	55x55x256	Weights 1x1x64x256 Bias 1x1x256
35	bn2c_branch2c Batch normaliza...	Batch Normalization	55x55x256	Offset 1x1x256 Scale 1x1x256
36	add_3 Element-wise a...	Addition	55x55x256	-
37	activation_10_... ReLU	ReLU	55x55x256	-
38	res3a_branch1 128 1x1x256 co...	Convolution	28x28x128	Weights 1x1x256x128 Bias 1x1x128
39	res3a_branch1 512 1x1x256 co...	Convolution	28x28x512	Weights 1x1x256x512 Bias 1x1x512
40	bn3a_branch1 Batch normaliza...	Batch Normalization	28x28x512	Offset 1x1x512 Scale 1x1x512

41	bn3a_branch2a Batch normaliza...	Batch Normalization	28x28x128	Offset 1x1x128 Scale 1x1x128
42	activation_11... ReLU	ReLU	28x28x128	-
43	res3a_branc... 128 3x3x128 co...	Convolution	28x28x128	Weights 3x3x128x128 Bias 1x1x128
44	bn3a_branch2b Batch normaliza...	Batch Normalization	28x28x128	Offset 1x1x128 Scale 1x1x128
45	activation_12... ReLU	ReLU	28x28x128	-
46	res3a_branc... 512 1x1x128 co...	Convolution	28x28x512	Weights 1x1x128x512 Bias 1x1x512
47	bn3a_branch2c Batch normaliza...	Batch Normalization	28x28x512	Offset 1x1x512 Scale 1x1x512
48	add_4 Element-wise a...	Addition	28x28x512	-
49	activation_13... ReLU	ReLU	28x28x512	-
50	res3b_branc... 128 1x1x512 co...	Convolution	28x28x128	Weights 1x1x512x128 Bias 1x1x128
51	bn3b_branch2a Batch normaliza...	Batch Normalization	28x28x128	Offset 1x1x128 Scale 1x1x128
52	activation_14... ReLU	ReLU	28x28x128	-
53	res3b_branc... 128 3x3x128 co...	Convolution	28x28x128	Weights 3x3x128x128 Bias 1x1x128
54	bn3b_branch2b Batch normaliza...	Batch Normalization	28x28x128	Offset 1x1x128 Scale 1x1x128
55	activation_15... ReLU	ReLU	28x28x128	-
56	res3b_branc... 512 1x1x128 co...	Convolution	28x28x512	Weights 1x1x128x512 Bias 1x1x512
57	bn3b_branch2c Batch normaliza...	Batch Normalization	28x28x512	Offset 1x1x512 Scale 1x1x512
58	add_5 Element-wise a...	Addition	28x28x512	-
59	activation_16... ReLU	ReLU	28x28x512	-
60	res3c_branc... 128 1x1x512 co...	Convolution	28x28x128	Weights 1x1x512x128 Bias 1x1x128

61	bn3c_branch2a Batch normaliza...	Batch Normalization	28x28x128	Offset 1x1x128 Scale 1x1x128
62	activation_17... ReLU	ReLU	28x28x128	-
63	res3c_branc... 128 3x3x128 co...	Convolution	28x28x128	Weights 3x3x128x128 Bias 1x1x128
64	bn3c_branch2b Batch normaliza...	Batch Normalization	28x28x128	Offset 1x1x128 Scale 1x1x128
65	activation_18... ReLU	ReLU	28x28x128	-
66	res3c_branc... 512 1x1x128 co...	Convolution	28x28x512	Weights 1x1x128x512 Bias 1x1x512
67	bn3c_branch2c Batch normaliza...	Batch Normalization	28x28x512	Offset 1x1x512 Scale 1x1x512
68	add_6 Element-wise a...	Addition	28x28x512	-
69	activation_19... ReLU	ReLU	28x28x512	-
70	res3d_branc... 128 1x1x512 co...	Convolution	28x28x128	Weights 1x1x512x128 Bias 1x1x128
71	bn3d_branch2a Batch normaliza...	Batch Normalization	28x28x128	Offset 1x1x128 Scale 1x1x128
72	activation_20... ReLU	ReLU	28x28x128	-
73	res3d_branc... 128 3x3x128 co...	Convolution	28x28x128	Weights 3x3x128x128 Bias 1x1x128
74	bn3d_branch2b Batch normaliza...	Batch Normalization	28x28x128	Offset 1x1x128 Scale 1x1x128
75	activation_21... ReLU	ReLU	28x28x128	-
76	res3d_branc... 512 1x1x128 co...	Convolution	28x28x512	Weights 1x1x128x512 Bias 1x1x512
77	bn3d_branch2c Batch normaliza...	Batch Normalization	28x28x512	Offset 1x1x512 Scale 1x1x512
78	add_7 Element-wise a...	Addition	28x28x512	-
79	activation_22... ReLU	ReLU	28x28x512	-
80	res4a_branch1 1024 1x1x512 c...	Convolution	14x14x1024	Weigh... 1x1x512x10... Bias 1x1x1024

81	bn4a_branch1 Batch normaliza...	Batch Normalization	14x14x1024	Offset 1x1x1024 Scale 1x1x1024
82	res4a_branc... 256 1x1x512 co...	Convolution	14x14x256	Weights 1x1x512x256 Bias 1x1x256
83	bn4a_branch2a Batch normaliza...	Batch Normalization	14x14x256	Offset 1x1x256 Scale 1x1x256
84	activation_23... ReLU	ReLU	14x14x256	-
85	res4a_branc... 256 3x3x256 co...	Convolution	14x14x256	Weights 3x3x256x256 Bias 1x1x256
86	bn4a_branch2b Batch normaliza...	Batch Normalization	14x14x256	Offset 1x1x256 Scale 1x1x256
87	activation_24... ReLU	ReLU	14x14x256	-
88	res4a_branc... 1024 1x1x256 c...	Convolution	14x14x1024	Weigh... 1x1x256x10... Bias 1x1x1024
89	bn4a_branch2c Batch normaliza...	Batch Normalization	14x14x1024	Offset 1x1x1024 Scale 1x1x1024
90	add_8 Element-wise a...	Addition	14x14x1024	-
91	activation_25... ReLU	ReLU	14x14x1024	-
92	res4b_branc... 256 1x1x1024 c...	Convolution	14x14x256	Weigh... 1x1x1024x2... Bias 1x1x256
93	bn4b_branch2a Batch normaliza...	Batch Normalization	14x14x256	Offset 1x1x256 Scale 1x1x256
94	activation_26... ReLU	ReLU	14x14x256	-
95	res4b_branc... 256 3x3x256 co...	Convolution	14x14x256	Weights 3x3x256x256 Bias 1x1x256
96	bn4b_branch2b Batch normaliza...	Batch Normalization	14x14x256	Offset 1x1x256 Scale 1x1x256
97	activation_27... ReLU	ReLU	14x14x256	-
98	res4b_branc... 1024 1x1x256 c...	Convolution	14x14x1024	Weigh... 1x1x256x10... Bias 1x1x1024
99	bn4b_branch2c Batch normaliza...	Batch Normalization	14x14x1024	Offset 1x1x1024 Scale 1x1x1024
100	add_9 Element-wise a...	Addition	14x14x1024	-

101	activation_28... ReLU	ReLU	14x14x1024	-
102	res4c_branch... 256 1x1x1024 c...	Convolution	14x14x256	Weigh... 1x1x1024x2... Bias 1x1x256
103	bn4c_branch2a Batch normaliza...	Batch Normalization	14x14x256	Offset 1x1x256 Scale 1x1x256
104	activation_29... ReLU	ReLU	14x14x256	-
105	res4c_branch... 256 3x3x256 co...	Convolution	14x14x256	Weights 3x3x256x256 Bias 1x1x256
106	bn4c_branch2b Batch normaliza...	Batch Normalization	14x14x256	Offset 1x1x256 Scale 1x1x256
107	activation_30... ReLU	ReLU	14x14x256	-
108	res4c_branch... 1024 1x1x256 c...	Convolution	14x14x1024	Weigh... 1x1x256x10... Bias 1x1x1024
109	bn4c_branch2c Batch normaliza...	Batch Normalization	14x14x1024	Offset 1x1x1024 Scale 1x1x1024
110	add_10 Element-wise a...	Addition	14x14x1024	-
111	activation_31... ReLU	ReLU	14x14x1024	-
112	res4d_branch... 256 1x1x1024 c...	Convolution	14x14x256	Weigh... 1x1x1024x2... Bias 1x1x256
113	bn4d_branch2a Batch normaliza...	Batch Normalization	14x14x256	Offset 1x1x256 Scale 1x1x256
114	activation_32... ReLU	ReLU	14x14x256	-
115	res4d_branch... 256 3x3x256 co...	Convolution	14x14x256	Weights 3x3x256x256 Bias 1x1x256
116	bn4d_branch2b Batch normaliza...	Batch Normalization	14x14x256	Offset 1x1x256 Scale 1x1x256
117	activation_33... ReLU	ReLU	14x14x256	-
118	res4d_branch... 1024 1x1x256 c...	Convolution	14x14x1024	Weigh... 1x1x256x10... Bias 1x1x1024
119	bn4d_branch2c Batch normaliza...	Batch Normalization	14x14x1024	Offset 1x1x1024 Scale 1x1x1024
120	add_11 Element-wise a...	Addition	14x14x1024	-

121	activation_34... ReLU	ReLU	14x14x1024	-
122	res4e_branch... 256 1x1x1024 c...	Convolution	14x14x256	Weigh... 1x1x1024x2... Bias 1x1x256
123	bn4e_branch2a Batch normaliza...	Batch Normalization	14x14x256	Offset 1x1x256 Scale 1x1x256
124	activation_35... ReLU	ReLU	14x14x256	-
125	res4e_branch... 256 3x3x256 co...	Convolution	14x14x256	Weights 3x3x256x256 Bias 1x1x256
126	bn4e_branch2b Batch normaliza...	Batch Normalization	14x14x256	Offset 1x1x256 Scale 1x1x256
127	activation_36... ReLU	ReLU	14x14x256	-
128	res4e_branch... 1024 1x1x256 c...	Convolution	14x14x1024	Weigh... 1x1x256x10... Bias 1x1x1024
129	bn4e_branch2c Batch normaliza...	Batch Normalization	14x14x1024	Offset 1x1x1024 Scale 1x1x1024
130	add_12 Element-wise a...	Addition	14x14x1024	-
131	activation_37... ReLU	ReLU	14x14x1024	-
132	res4f_branch2a 256 1x1x1024 c...	Convolution	14x14x256	Weigh... 1x1x1024x2... Bias 1x1x256
133	bn4f_branch2a Batch normaliza...	Batch Normalization	14x14x256	Offset 1x1x256 Scale 1x1x256
134	activation_38... ReLU	ReLU	14x14x256	-
135	res4f_branch2b 256 3x3x256 co...	Convolution	14x14x256	Weights 3x3x256x256 Bias 1x1x256
136	bn4f_branch2b Batch normaliza...	Batch Normalization	14x14x256	Offset 1x1x256 Scale 1x1x256
137	activation_39... ReLU	ReLU	14x14x256	-
138	res4f_branch2c 1024 1x1x256 c...	Convolution	14x14x1024	Weigh... 1x1x256x10... Bias 1x1x1024
139	bn4f_branch2c Batch normaliza...	Batch Normalization	14x14x1024	Offset 1x1x1024 Scale 1x1x1024
140	add_13 Element-wise a...	Addition	14x14x1024	-

141	activation_40...	ReLU	$14 \times 14 \times 1024$	-
142	res5a_branch...	Convolution 512 $1 \times 1 \times 1024$ c...	$7 \times 7 \times 512$	Weigh... $1 \times 1 \times 1024 \times 5...$ Bias $1 \times 1 \times 512$
143	bn5a_branch2a	Batch Normalization Batch normaliza...	$7 \times 7 \times 512$	Offset $1 \times 1 \times 512$ Scale $1 \times 1 \times 512$
144	activation_41...	ReLU	$7 \times 7 \times 512$	-
145	res5a_branch...	Convolution 512 $3 \times 3 \times 512$ co...	$7 \times 7 \times 512$	Weights $3 \times 3 \times 512 \times 512$ Bias $1 \times 1 \times 512$
146	bn5a_branch2b	Batch Normalization Batch normaliza...	$7 \times 7 \times 512$	Offset $1 \times 1 \times 512$ Scale $1 \times 1 \times 512$
147	activation_42...	ReLU	$7 \times 7 \times 512$	-
148	res5a_branch...	Convolution 2048 $1 \times 1 \times 512$ c...	$7 \times 7 \times 2048$	Weigh... $1 \times 1 \times 512 \times 20...$ Bias $1 \times 1 \times 2048$
149	res5a_branch1	Convolution 2048 $1 \times 1 \times 1024$...	$7 \times 7 \times 2048$	Weigh... $1 \times 1 \times 1024 \times 20...$ Bias $1 \times 1 \times 2048$
150	bn5a_branch2c	Batch Normalization Batch normaliza...	$7 \times 7 \times 2048$	Offset $1 \times 1 \times 2048$ Scale $1 \times 1 \times 2048$
151	bn5a_branch1	Batch Normalization Batch normaliza...	$7 \times 7 \times 2048$	Offset $1 \times 1 \times 2048$ Scale $1 \times 1 \times 2048$
152	add_14	Addition Element-wise a...	$7 \times 7 \times 2048$	-
153	activation_43...	ReLU	$7 \times 7 \times 2048$	-
154	res5b_branch...	Convolution 512 $1 \times 1 \times 2048$ c...	$7 \times 7 \times 512$	Weigh... $1 \times 1 \times 2048 \times 5...$ Bias $1 \times 1 \times 512$
155	bn5b_branch2a	Batch Normalization Batch normaliza...	$7 \times 7 \times 512$	Offset $1 \times 1 \times 512$ Scale $1 \times 1 \times 512$
156	activation_44...	ReLU	$7 \times 7 \times 512$	-
157	res5b_branch...	Convolution 512 $3 \times 3 \times 512$ co...	$7 \times 7 \times 512$	Weights $3 \times 3 \times 512 \times 512$ Bias $1 \times 1 \times 512$
158	bn5b_branch2b	Batch Normalization Batch normaliza...	$7 \times 7 \times 512$	Offset $1 \times 1 \times 512$ Scale $1 \times 1 \times 512$
159	activation_45...	ReLU	$7 \times 7 \times 512$	-
160	res5b_branch...	Convolution 2048 $1 \times 1 \times 512$ c...	$7 \times 7 \times 2048$	Weigh... $1 \times 1 \times 512 \times 20...$ Bias $1 \times 1 \times 2048$

161	bn5b_branch2c Batch normaliza...	Batch Normalization	7x7x2048	Offset 1x1x2048 Scale 1x1x2048
162	add_15 Element-wise a...	Addition	7x7x2048	-
163	activation_46... ReLU	ReLU	7x7x2048	-
164	res5c_branc... 512 1x1x2048 c...	Convolution	7x7x512	Weigh... 1x1x2048x5... Bias 1x1x512
165	bn5c_branch2a Batch normaliza...	Batch Normalization	7x7x512	Offset 1x1x512 Scale 1x1x512
166	activation_47... ReLU	ReLU	7x7x512	-
167	res5c_branc... 512 3x3x512 co...	Convolution	7x7x512	Weights 3x3x512x512 Bias 1x1x512
168	bn5c_branch2b Batch normaliza...	Batch Normalization	7x7x512	Offset 1x1x512 Scale 1x1x512
169	activation_48... ReLU	ReLU	7x7x512	-
170	res5c_branc... 2048 1x1x512 c...	Convolution	7x7x2048	Weigh... 1x1x512x20... Bias 1x1x2048
171	bn5c_branch2c Batch normaliza...	Batch Normalization	7x7x2048	Offset 1x1x2048 Scale 1x1x2048
172	add_16 Element-wise a...	Addition	7x7x2048	-
173	activation_49... ReLU	ReLU	7x7x2048	-
174	avg_pool 7x7 average po...	Average Pooling	1x1x2048	-
175	fc 4 fully connecte...	Fully Connected	1x1x4	Weights 4x2048 Bias 4x1
176	fc1000_soft... softmax	Softmax	1x1x4	-
177	classoutput crossentropyex	Classification Output	-	-

LAMPIRAN 3 – *Source Code Network VGG-19*

```
layers = [
    imageInputLayer([224 224 3], "Name", "imageinput")
    convolution2dLayer([3 3], 64, "Name", "conv_1", "Padding", [1 1 1 1])
    reluLayer("Name", "relu_1")
    convolution2dLayer([3 3], 64, "Name", "conv_2", "Padding", [1 1 1 1])
    reluLayer("Name", "relu_2")
    maxPooling2dLayer([2 2], "Name", "maxpool_1", "Stride", [2 2])
    convolution2dLayer([3 3], 128, "Name", "conv_3", "Padding", [1 1 1 1])
    reluLayer("Name", "relu_3")
    convolution2dLayer([3 3], 128, "Name", "conv_4", "Padding", [1 1 1 1])
    reluLayer("Name", "relu_4")
    maxPooling2dLayer([2 2], "Name", "maxpool_2", "Stride", [2 2])
    convolution2dLayer([3 3], 256, "Name", "conv_5", "Padding", [1 1 1 1])
    reluLayer("Name", "relu_5")
    convolution2dLayer([3 3], 256, "Name", "conv_6", "Padding", [1 1 1 1])
    reluLayer("Name", "relu_6")
    convolution2dLayer([3 3], 256, "Name", "conv_7", "Padding", [1 1 1 1])
    reluLayer("Name", "relu_7")
    convolution2dLayer([3 3], 256, "Name", "conv_8", "Padding", [1 1 1 1])
    reluLayer("Name", "relu_8")
    maxPooling2dLayer([2 2], "Name", "maxpool_3", "Stride", [2 2])
    convolution2dLayer([3 3], 512, "Name", "conv_9", "Padding", [1 1 1 1])
    reluLayer("Name", "relu_9")
    convolution2dLayer([3 3], 512, "Name", "conv_10", "Padding", [1 1 1 1])
    reluLayer("Name", "relu_10")
    convolution2dLayer([3 3], 512, "Name", "conv_11", "Padding", [1 1 1 1])
    reluLayer("Name", "relu_11")
    convolution2dLayer([3 3], 512, "Name", "conv_12", "Padding", [1 1 1 1])
    reluLayer("Name", "relu_12")
    maxPooling2dLayer([2 2], "Name", "maxpool_4", "Stride", [2 2])
    convolution2dLayer([3 3], 512, "Name", "conv_13", "Padding", [1 1 1 1])
    reluLayer("Name", "relu_13")
    convolution2dLayer([3 3], 512, "Name", "conv_14", "Padding", [1 1 1 1])
    reluLayer("Name", "relu_14")
    convolution2dLayer([3 3], 512, "Name", "conv_15", "Padding", [1 1 1 1])
    reluLayer("Name", "relu_15")
    convolution2dLayer([3 3], 512, "Name", "conv_16", "Padding", [1 1 1 1])
    reluLayer("Name", "relu_16")
    maxPooling2dLayer([2 2], "Name", "maxpool_5", "Stride", [2 2])
    fullyConnectedLayer(4096, "Name", "fc_1")
    reluLayer("Name", "relu_17")
    dropoutLayer(0.5, "Name", "dropout_1")
    fullyConnectedLayer(4096, "Name", "fc_2")
    reluLayer("Name", "relu_18")
    dropoutLayer(0.5, "Name", "dropout_2")]
```

```
fullyConnectedLayer(1000, "Name", "fc_3")
softmaxLayer("Name", "softmax")
classificationLayer("Name", "classoutput")];
plot(layerGraph(layers));
```

LAMPIRAN 4 – *Source Code Training VGG-19*

```
addpath('C:\Users\LENOVO\OneDrive\Desktop\Materi Kuliah\Semester 7\Teknik Neurofuzi\Tugas\Deep Learning Project\Program Files\VGG19');
unzip('Training.zip');
imds =
 imageDatastore('Training','IncludeSubfolders',true,'LabelSource','foldernames')
;
[imdsTrain,imdsValidation] = splitEachLabel(imds,0.7,'randomized');

augimdsTrain = augmentedImageDatastore([224 224],imdsTrain);
augimdsValidation = augmentedImageDatastore([224 224],imdsValidation);

options = trainingOptions('sgdm', ...
    'MiniBatchSize',10, ...
    'MaxEpochs',6, ...
    'Shuffle','every-epoch', ...
    'InitialLearnRate',1e-4, ...
    'ValidationData',augimdsValidation, ...
    'ValidationFrequency',6, ...
    'Verbose',false, ...
    'Plots','training-progress');

netTransfer = trainNetwork(augimdsTrain,layers_1,options);
```

LAMPIRAN 5 – *Source Code Testing VGG-19*

```
addpath('C:\Users\LENOVO\OneDrive\Desktop\Materi Kuliah\Semester 7\Teknik Neurofuzi\Tugas\Deep Learning Project\Program Files\VGG19');
unzip('Testing.zip');
imds =
 imageDatastore('Testing','IncludeSubfolders',true,'LabelSource','foldernames');
imdsValidation = imds;

augimdsValidation = augmentedImageDatastore([224 224], imdsValidation);

[YPred,probs] = classify(netTransfer,augimdsValidation);
accuracy = mean(YPred == imdsValidation.Labels)

idx = randperm(numel(augimdsValidation.Files),9);
figure
for i = 1:9
    subplot(3,3,i)
    I = readimage(imdsValidation,idx(i));
    imshow(I)
    label = YPred(idx(i));
    title(string(label) + ", " + num2str(100*max(probs(idx(i),:))),3) + "%");
end
```

LAMPIRAN 6 – Source Code Network ResNet-50

```
lgraph = layerGraph();

tempLayers = [
    imageInputLayer([224 224 3], "Name", "input_1")
    convolution2dLayer([7 7], 64, "Name", "conv1", "Padding", [3 3 3 3], "Stride", [2
2])
    batchNormalizationLayer("Name", "bn_conv1", "Epsilon", 0.001)
    reluLayer("Name", "activation_1_relu")
    maxPooling2dLayer([3 3], "Name", "max_pooling2d_1", "Stride", [2 2])];
lgraph = addLayers(lgraph,tempLayers);

tempLayers = [
    convolution2dLayer([1 1], 64, "Name", "res2a_branch2a")
    batchNormalizationLayer("Name", "bn2a_branch2a", "Epsilon", 0.001)
    reluLayer("Name", "activation_2_relu")
    convolution2dLayer([3 3], 64, "Name", "res2a_branch2b", "Padding", "same")
    batchNormalizationLayer("Name", "bn2a_branch2b", "Epsilon", 0.001)
    reluLayer("Name", "activation_3_relu")
    convolution2dLayer([1 1], 256, "Name", "res2a_branch2c")
    batchNormalizationLayer("Name", "bn2a_branch2c", "Epsilon", 0.001)];
lgraph = addLayers(lgraph,tempLayers);

tempLayers = [
    convolution2dLayer([1 1], 256, "Name", "res2a_branch1")
    batchNormalizationLayer("Name", "bn2a_branch1", "Epsilon", 0.001)];
lgraph = addLayers(lgraph,tempLayers);

tempLayers = [
    additionLayer(2, "Name", "add_1")
    reluLayer("Name", "activation_4_relu")];
lgraph = addLayers(lgraph,tempLayers);

tempLayers = [
    convolution2dLayer([1 1], 64, "Name", "res2b_branch2a")
    batchNormalizationLayer("Name", "bn2b_branch2a", "Epsilon", 0.001)
    reluLayer("Name", "activation_5_relu")
    convolution2dLayer([3 3], 64, "Name", "res2b_branch2b", "Padding", "same")
    batchNormalizationLayer("Name", "bn2b_branch2b", "Epsilon", 0.001)
    reluLayer("Name", "activation_6_relu")
    convolution2dLayer([1 1], 256, "Name", "res2b_branch2c")
    batchNormalizationLayer("Name", "bn2b_branch2c", "Epsilon", 0.001)];
```

```

lgraph = addLayers(lgraph,tempLayers);

tempLayers = [
    additionLayer(2, "Name", "add_2")
    reluLayer("Name", "activation_7_relu")];
lgraph = addLayers(lgraph,tempLayers);

tempLayers = [
    convolution2dLayer([1 1],64, "Name", "res2c_branch2a")
    batchNormalizationLayer("Name", "bn2c_branch2a", "Epsilon",0.001)
    reluLayer("Name", "activation_8_relu")
    convolution2dLayer([3 3],64, "Name", "res2c_branch2b", "Padding", "same")
    batchNormalizationLayer("Name", "bn2c_branch2b", "Epsilon",0.001)
    reluLayer("Name", "activation_9_relu")
    convolution2dLayer([1 1],256, "Name", "res2c_branch2c")
    batchNormalizationLayer("Name", "bn2c_branch2c", "Epsilon",0.001)];
lgraph = addLayers(lgraph,tempLayers);

tempLayers = [
    additionLayer(2, "Name", "add_3")
    reluLayer("Name", "activation_10_relu")];
lgraph = addLayers(lgraph,tempLayers);

tempLayers = [
    convolution2dLayer([1 1],128, "Name", "res3a_branch2a", "Stride", [2 2])
    batchNormalizationLayer("Name", "bn3a_branch2a", "Epsilon",0.001)
    reluLayer("Name", "activation_11_relu")
    convolution2dLayer([3 3],128, "Name", "res3a_branch2b", "Padding", "same")
    batchNormalizationLayer("Name", "bn3a_branch2b", "Epsilon",0.001)
    reluLayer("Name", "activation_12_relu")
    convolution2dLayer([1 1],512, "Name", "res3a_branch2c")
    batchNormalizationLayer("Name", "bn3a_branch2c", "Epsilon",0.001)];
lgraph = addLayers(lgraph,tempLayers);

tempLayers = [
    convolution2dLayer([1 1],512, "Name", "res3a_branch1", "Stride", [2 2])
    batchNormalizationLayer("Name", "bn3a_branch1", "Epsilon",0.001)];
lgraph = addLayers(lgraph,tempLayers);

tempLayers = [
    additionLayer(2, "Name", "add_4")
    reluLayer("Name", "activation_13_relu")];

```

```

lgraph = addLayers(lgraph,tempLayers);

tempLayers = [
    convolution2dLayer([1 1],128,"Name","res3b_branch2a")
    batchNormalizationLayer("Name","bn3b_branch2a","Epsilon",0.001)
    reluLayer("Name","activation_14_relu")
    convolution2dLayer([3 3],128,"Name","res3b_branch2b","Padding","same")
    batchNormalizationLayer("Name","bn3b_branch2b","Epsilon",0.001)
    reluLayer("Name","activation_15_relu")
    convolution2dLayer([1 1],512,"Name","res3b_branch2c")
    batchNormalizationLayer("Name","bn3b_branch2c","Epsilon",0.001)];
lgraph = addLayers(lgraph,tempLayers);

tempLayers = [
    additionLayer(2,"Name","add_5")
    reluLayer("Name","activation_16_relu")];
lgraph = addLayers(lgraph,tempLayers);

tempLayers = [
    convolution2dLayer([1 1],128,"Name","res3c_branch2a")
    batchNormalizationLayer("Name","bn3c_branch2a","Epsilon",0.001)
    reluLayer("Name","activation_17_relu")
    convolution2dLayer([3 3],128,"Name","res3c_branch2b","Padding","same")
    batchNormalizationLayer("Name","bn3c_branch2b","Epsilon",0.001)
    reluLayer("Name","activation_18_relu")
    convolution2dLayer([1 1],512,"Name","res3c_branch2c")
    batchNormalizationLayer("Name","bn3c_branch2c","Epsilon",0.001)];
lgraph = addLayers(lgraph,tempLayers);

tempLayers = [
    additionLayer(2,"Name","add_6")
    reluLayer("Name","activation_19_relu")];
lgraph = addLayers(lgraph,tempLayers);

tempLayers = [
    convolution2dLayer([1 1],128,"Name","res3d_branch2a")
    batchNormalizationLayer("Name","bn3d_branch2a","Epsilon",0.001)
    reluLayer("Name","activation_20_relu")
    convolution2dLayer([3 3],128,"Name","res3d_branch2b","Padding","same")
    batchNormalizationLayer("Name","bn3d_branch2b","Epsilon",0.001)
    reluLayer("Name","activation_21_relu")
    convolution2dLayer([1 1],512,"Name","res3d_branch2c")
    batchNormalizationLayer("Name","bn3d_branch2c","Epsilon",0.001)];

```

```

lgraph = addLayers(lgraph,tempLayers);

tempLayers = [
    additionLayer(2, "Name", "add_7")
    reluLayer("Name", "activation_22_relu")];
lgraph = addLayers(lgraph,tempLayers);

tempLayers = [
    convolution2dLayer([1 1],1024, "Name", "res4a_branch1", "Stride", [2 2])
    batchNormalizationLayer("Name", "bn4a_branch1", "Epsilon", 0.001)];
lgraph = addLayers(lgraph,tempLayers);

tempLayers = [
    convolution2dLayer([1 1],256, "Name", "res4a_branch2a", "Stride", [2 2])
    batchNormalizationLayer("Name", "bn4a_branch2a", "Epsilon", 0.001)
    reluLayer("Name", "activation_23_relu")
    convolution2dLayer([3 3],256, "Name", "res4a_branch2b", "Padding", "same")
    batchNormalizationLayer("Name", "bn4a_branch2b", "Epsilon", 0.001)
    reluLayer("Name", "activation_24_relu")
    convolution2dLayer([1 1],1024, "Name", "res4a_branch2c")
    batchNormalizationLayer("Name", "bn4a_branch2c", "Epsilon", 0.001)];
lgraph = addLayers(lgraph,tempLayers);

tempLayers = [
    additionLayer(2, "Name", "add_8")
    reluLayer("Name", "activation_25_relu")];
lgraph = addLayers(lgraph,tempLayers);

tempLayers = [
    convolution2dLayer([1 1],256, "Name", "res4b_branch2a")
    batchNormalizationLayer("Name", "bn4b_branch2a", "Epsilon", 0.001)
    reluLayer("Name", "activation_26_relu")
    convolution2dLayer([3 3],256, "Name", "res4b_branch2b", "Padding", "same")
    batchNormalizationLayer("Name", "bn4b_branch2b", "Epsilon", 0.001)
    reluLayer("Name", "activation_27_relu")
    convolution2dLayer([1 1],1024, "Name", "res4b_branch2c")
    batchNormalizationLayer("Name", "bn4b_branch2c", "Epsilon", 0.001)];
lgraph = addLayers(lgraph,tempLayers);

tempLayers = [
    additionLayer(2, "Name", "add_9")
    reluLayer("Name", "activation_28_relu")];

```

```

lgraph = addLayers(lgraph,tempLayers);

tempLayers = [
    convolution2dLayer([1 1],256,"Name","res4c_branch2a")
    batchNormalizationLayer("Name","bn4c_branch2a","Epsilon",0.001)
    reluLayer("Name","activation_29_relu")
    convolution2dLayer([3 3],256,"Name","res4c_branch2b","Padding","same")
    batchNormalizationLayer("Name","bn4c_branch2b","Epsilon",0.001)
    reluLayer("Name","activation_30_relu")
    convolution2dLayer([1 1],1024,"Name","res4c_branch2c")
    batchNormalizationLayer("Name","bn4c_branch2c","Epsilon",0.001)];
lgraph = addLayers(lgraph,tempLayers);

tempLayers = [
    additionLayer(2,"Name","add_10")
    reluLayer("Name","activation_31_relu")];
lgraph = addLayers(lgraph,tempLayers);

tempLayers = [
    convolution2dLayer([1 1],256,"Name","res4d_branch2a")
    batchNormalizationLayer("Name","bn4d_branch2a","Epsilon",0.001)
    reluLayer("Name","activation_32_relu")
    convolution2dLayer([3 3],256,"Name","res4d_branch2b","Padding","same")
    batchNormalizationLayer("Name","bn4d_branch2b","Epsilon",0.001)
    reluLayer("Name","activation_33_relu")
    convolution2dLayer([1 1],1024,"Name","res4d_branch2c")
    batchNormalizationLayer("Name","bn4d_branch2c","Epsilon",0.001)];
lgraph = addLayers(lgraph,tempLayers);

tempLayers = [
    additionLayer(2,"Name","add_11")
    reluLayer("Name","activation_34_relu")];
lgraph = addLayers(lgraph,tempLayers);

tempLayers = [
    convolution2dLayer([1 1],256,"Name","res4e_branch2a")
    batchNormalizationLayer("Name","bn4e_branch2a","Epsilon",0.001)
    reluLayer("Name","activation_35_relu")
    convolution2dLayer([3 3],256,"Name","res4e_branch2b","Padding","same")
    batchNormalizationLayer("Name","bn4e_branch2b","Epsilon",0.001)
    reluLayer("Name","activation_36_relu")
    convolution2dLayer([1 1],1024,"Name","res4e_branch2c")
    batchNormalizationLayer("Name","bn4e_branch2c","Epsilon",0.001)];

```

```

lgraph = addLayers(lgraph,tempLayers);

tempLayers = [
    additionLayer(2, "Name", "add_12")
    reluLayer("Name", "activation_37_relu")];
lgraph = addLayers(lgraph,tempLayers);

tempLayers = [
    convolution2dLayer([1 1],256, "Name", "res4f_branch2a")
    batchNormalizationLayer("Name", "bn4f_branch2a", "Epsilon",0.001)
    reluLayer("Name", "activation_38_relu")
    convolution2dLayer([3 3],256, "Name", "res4f_branch2b", "Padding", "same")
    batchNormalizationLayer("Name", "bn4f_branch2b", "Epsilon",0.001)
    reluLayer("Name", "activation_39_relu")
    convolution2dLayer([1 1],1024, "Name", "res4f_branch2c")
    batchNormalizationLayer("Name", "bn4f_branch2c", "Epsilon",0.001)];
lgraph = addLayers(lgraph,tempLayers);

tempLayers = [
    additionLayer(2, "Name", "add_13")
    reluLayer("Name", "activation_40_relu")];
lgraph = addLayers(lgraph,tempLayers);

tempLayers = [
    convolution2dLayer([1 1],512, "Name", "res5a_branch2a", "Stride", [2 2])
    batchNormalizationLayer("Name", "bn5a_branch2a", "Epsilon",0.001)
    reluLayer("Name", "activation_41_relu")
    convolution2dLayer([3 3],512, "Name", "res5a_branch2b", "Padding", "same")
    batchNormalizationLayer("Name", "bn5a_branch2b", "Epsilon",0.001)
    reluLayer("Name", "activation_42_relu")
    convolution2dLayer([1 1],2048, "Name", "res5a_branch2c")
    batchNormalizationLayer("Name", "bn5a_branch2c", "Epsilon",0.001)];
lgraph = addLayers(lgraph,tempLayers);

tempLayers = [
    convolution2dLayer([1 1],2048, "Name", "res5a_branch1", "Stride", [2 2])
    batchNormalizationLayer("Name", "bn5a_branch1", "Epsilon",0.001)];
lgraph = addLayers(lgraph,tempLayers);

tempLayers = [
    additionLayer(2, "Name", "add_14")
    reluLayer("Name", "activation_43_relu")];

```

```

lgraph = addLayers(lgraph,tempLayers);

tempLayers = [
    convolution2dLayer([1 1],512,"Name","res5b_branch2a")
    batchNormalizationLayer("Name","bn5b_branch2a","Epsilon",0.001)
    reluLayer("Name","activation_44_relu")
    convolution2dLayer([3 3],512,"Name","res5b_branch2b","Padding","same")
    batchNormalizationLayer("Name","bn5b_branch2b","Epsilon",0.001)
    reluLayer("Name","activation_45_relu")
    convolution2dLayer([1 1],2048,"Name","res5b_branch2c")
    batchNormalizationLayer("Name","bn5b_branch2c","Epsilon",0.001)];
lgraph = addLayers(lgraph,tempLayers);

tempLayers = [
    additionLayer(2,"Name","add_15")
    reluLayer("Name","activation_46_relu")];
lgraph = addLayers(lgraph,tempLayers);

tempLayers = [
    convolution2dLayer([1 1],512,"Name","res5c_branch2a")
    batchNormalizationLayer("Name","bn5c_branch2a","Epsilon",0.001)
    reluLayer("Name","activation_47_relu")
    convolution2dLayer([3 3],512,"Name","res5c_branch2b","Padding","same")
    batchNormalizationLayer("Name","bn5c_branch2b","Epsilon",0.001)
    reluLayer("Name","activation_48_relu")
    convolution2dLayer([1 1],2048,"Name","res5c_branch2c")
    batchNormalizationLayer("Name","bn5c_branch2c","Epsilon",0.001)];
lgraph = addLayers(lgraph,tempLayers);

tempLayers = [
    additionLayer(2,"Name","add_16")
    reluLayer("Name","activation_49_relu")
    averagePooling2dLayer([7 7],"Name","avg_pool","Stride",[7 7])
    fullyConnectedLayer(4,"Name","fc")
    softmaxLayer("Name","fc1000_softmax")
    classificationLayer("Name","classoutput")];
lgraph = addLayers(lgraph,tempLayers);

lgraph = connectLayers(lgraph,"max_pooling2d_1","res2a_branch2a");
lgraph = connectLayers(lgraph,"max_pooling2d_1","res2a_branch1");
lgraph = connectLayers(lgraph,"bn2a_branch2c","add_1/in1");
lgraph = connectLayers(lgraph,"bn2a_branch1","add_1/in2");
lgraph = connectLayers(lgraph,"activation_4_relu","res2b_branch2a");

```

```

lgraph = connectLayers(lgraph, "activation_4_relu", "add_2/in2");
lgraph = connectLayers(lgraph, "bn2b_branch2c", "add_2/in1");
lgraph = connectLayers(lgraph, "activation_7_relu", "res2c_branch2a");
lgraph = connectLayers(lgraph, "activation_7_relu", "add_3/in2");
lgraph = connectLayers(lgraph, "bn2c_branch2c", "add_3/in1");
lgraph = connectLayers(lgraph, "activation_10_relu", "res3a_branch2a");
lgraph = connectLayers(lgraph, "activation_10_relu", "res3a_branch1");
lgraph = connectLayers(lgraph, "bn3a_branch1", "add_4/in2");
lgraph = connectLayers(lgraph, "bn3a_branch2c", "add_4/in1");
lgraph = connectLayers(lgraph, "activation_13_relu", "res3b_branch2a");
lgraph = connectLayers(lgraph, "activation_13_relu", "add_5/in2");
lgraph = connectLayers(lgraph, "bn3b_branch2c", "add_5/in1");
lgraph = connectLayers(lgraph, "activation_16_relu", "res3c_branch2a");
lgraph = connectLayers(lgraph, "activation_16_relu", "add_6/in2");
lgraph = connectLayers(lgraph, "bn3c_branch2c", "add_6/in1");
lgraph = connectLayers(lgraph, "activation_19_relu", "res3d_branch2a");
lgraph = connectLayers(lgraph, "activation_19_relu", "add_7/in2");
lgraph = connectLayers(lgraph, "bn3d_branch2c", "add_7/in1");
lgraph = connectLayers(lgraph, "activation_22_relu", "res4a_branch1");
lgraph = connectLayers(lgraph, "activation_22_relu", "res4a_branch2a");
lgraph = connectLayers(lgraph, "bn4a_branch1", "add_8/in2");
lgraph = connectLayers(lgraph, "bn4a_branch2c", "add_8/in1");
lgraph = connectLayers(lgraph, "activation_25_relu", "res4b_branch2a");
lgraph = connectLayers(lgraph, "activation_25_relu", "add_9/in2");
lgraph = connectLayers(lgraph, "bn4b_branch2c", "add_9/in1");
lgraph = connectLayers(lgraph, "activation_28_relu", "res4c_branch2a");
lgraph = connectLayers(lgraph, "activation_28_relu", "add_10/in2");
lgraph = connectLayers(lgraph, "bn4c_branch2c", "add_10/in1");
lgraph = connectLayers(lgraph, "activation_31_relu", "res4d_branch2a");
lgraph = connectLayers(lgraph, "activation_31_relu", "add_11/in2");
lgraph = connectLayers(lgraph, "bn4d_branch2c", "add_11/in1");
lgraph = connectLayers(lgraph, "activation_34_relu", "res4e_branch2a");
lgraph = connectLayers(lgraph, "activation_34_relu", "add_12/in2");
lgraph = connectLayers(lgraph, "bn4e_branch2c", "add_12/in1");
lgraph = connectLayers(lgraph, "activation_37_relu", "res4f_branch2a");
lgraph = connectLayers(lgraph, "activation_37_relu", "add_13/in2");
lgraph = connectLayers(lgraph, "bn4f_branch2c", "add_13/in1");
lgraph = connectLayers(lgraph, "activation_40_relu", "res5a_branch2a");
lgraph = connectLayers(lgraph, "activation_40_relu", "res5a_branch1");
lgraph = connectLayers(lgraph, "bn5a_branch2c", "add_14/in1");
lgraph = connectLayers(lgraph, "bn5a_branch1", "add_14/in2");
lgraph = connectLayers(lgraph, "activation_43_relu", "res5b_branch2a");
lgraph = connectLayers(lgraph, "activation_43_relu", "add_15/in2");
lgraph = connectLayers(lgraph, "bn5b_branch2c", "add_15/in1");
lgraph = connectLayers(lgraph, "activation_46_relu", "res5c_branch2a");
lgraph = connectLayers(lgraph, "activation_46_relu", "add_16/in2");
lgraph = connectLayers(lgraph, "bn5c_branch2c", "add_16/in1");

```

```
clear tempLayers;  
plot(lgraph);
```

LAMPIRAN 7 – *Source Code Training ResNet-50*

```
addpath('C:\Users\LENOVO\OneDrive\Desktop\Materi Kuliah\Semester 7\Teknik Neurofuzi\Tugas\Deep Learning Project\Program Files\VGG19');
unzip('Training.zip');
imds =
 imageDatastore('Training','IncludeSubfolders',true,'LabelSource','foldernames')
;
[imdsTrain,imdsValidation] = splitEachLabel(imds,0.7,'randomized');

augimdsTrain = augmentedImageDatastore([224 224],imdsTrain);
augimdsValidation = augmentedImageDatastore([224 224],imdsValidation);

options = trainingOptions('sgdm', ...
    'MiniBatchSize',10, ...
    'MaxEpochs',6, ...
    'Shuffle','every-epoch', ...
    'InitialLearnRate',1e-4, ...
    'ValidationData',augimdsValidation, ...
    'ValidationFrequency',6, ...
    'Verbose',false, ...
    'Plots','training-progress');

netTransfer = trainNetwork(augimdsTrain,lgraph,options);
```

LAMPIRAN 8 – *Source Code Testing* ResNet-50

```
addpath('C:\Users\LENOVO\OneDrive\Desktop\Materi Kuliah\Semester 7\Teknik Neurofuzi\Tugas\Deep Learning Project\Program Files\VGG19');
unzip('Testing.zip');
imds =
 imageDatastore('Testing','IncludeSubfolders',true,'LabelSource','foldernames');
imdsValidation = imds;

augimdsValidation = augmentedImageDatastore([224 224], imdsValidation);

[YPred,probs] = classify(netTransfer,augimdsValidation);
accuracy = mean(YPred == imdsValidation.Labels)

idx = randperm(numel(augimdsValidation.Files),9);
figure
for i = 1:9
    subplot(3,3,i)
    I = readimage(imdsValidation,idx(i));
    imshow(I)
    label = YPred(idx(i));
    title(string(label) + ", " + num2str(100*max(probs(idx(i),:))),3) + "%");
end
```

LAMPIRAN 9 – Weight & Bias Hasil Latihan VGG-19

<pre> >> netTransfer.Layers(2).Weights 3×3×3×64 single array ans(:,:,1,1) = </pre> <table border="0"> <tr><td>-0.0005</td><td>-0.0871</td><td>-0.0466</td></tr> <tr><td>0.0179</td><td>-0.0582</td><td>-0.0122</td></tr> <tr><td>-0.0067</td><td>0.0710</td><td>0.0601</td></tr> </table> <pre> ans(:,:,2,1) = </pre> <table border="0"> <tr><td>-0.0411</td><td>-0.0898</td><td>0.0580</td></tr> <tr><td>-0.0664</td><td>-0.0914</td><td>-0.0878</td></tr> <tr><td>-0.0476</td><td>0.0674</td><td>-0.0127</td></tr> </table> <pre> ans(:,:,3,1) = </pre> <table border="0"> <tr><td>-0.0297</td><td>-0.0205</td><td>0.0615</td></tr> <tr><td>-0.0671</td><td>0.0582</td><td>-0.0629</td></tr> <tr><td>-0.0246</td><td>0.0712</td><td>-0.0340</td></tr> </table> <pre> ans(:,:,1,2) = </pre> <table border="0"> <tr><td>-0.0173</td><td>0.0089</td><td>-0.0471</td></tr> <tr><td>0.0911</td><td>0.0700</td><td>0.0999</td></tr> <tr><td>-0.0121</td><td>0.0905</td><td>-0.0329</td></tr> </table> <pre> ans(:,:,2,2) = </pre> <table border="0"> <tr><td>-0.0493</td><td>-0.0849</td><td>-0.0084</td></tr> <tr><td>0.0738</td><td>-0.0908</td><td>-0.0636</td></tr> <tr><td>0.0359</td><td>0.0037</td><td>0.0356</td></tr> </table> <pre> ans(:,:,3,2) = </pre> <table border="0"> <tr><td>0.0214</td><td>-0.0679</td><td>0.0628</td></tr> <tr><td>0.0762</td><td>0.0619</td><td>-0.0000</td></tr> <tr><td>-0.0084</td><td>0.0928</td><td>0.0416</td></tr> </table> <pre> ans(:,:,1,3) = </pre> <table border="0"> <tr><td>0.0664</td><td>0.0791</td><td>-0.0194</td></tr> <tr><td>0.0490</td><td>0.0592</td><td>-0.0064</td></tr> <tr><td>-0.0771</td><td>-0.0557</td><td>0.0650</td></tr> </table>	-0.0005	-0.0871	-0.0466	0.0179	-0.0582	-0.0122	-0.0067	0.0710	0.0601	-0.0411	-0.0898	0.0580	-0.0664	-0.0914	-0.0878	-0.0476	0.0674	-0.0127	-0.0297	-0.0205	0.0615	-0.0671	0.0582	-0.0629	-0.0246	0.0712	-0.0340	-0.0173	0.0089	-0.0471	0.0911	0.0700	0.0999	-0.0121	0.0905	-0.0329	-0.0493	-0.0849	-0.0084	0.0738	-0.0908	-0.0636	0.0359	0.0037	0.0356	0.0214	-0.0679	0.0628	0.0762	0.0619	-0.0000	-0.0084	0.0928	0.0416	0.0664	0.0791	-0.0194	0.0490	0.0592	-0.0064	-0.0771	-0.0557	0.0650	<pre> >> netTransfer.Layers(2).Bias 1×1×64 single array ans(:,:,1) = </pre> <table border="0"> <tr><td>9.1119e-05</td></tr> </table> <pre> ans(:,:,2) = </pre> <table border="0"> <tr><td>-3.4369e-05</td></tr> </table> <pre> ans(:,:,3) = </pre> <table border="0"> <tr><td>1.3568e-06</td></tr> </table> <pre> ans(:,:,4) = </pre> <table border="0"> <tr><td>-5.0792e-05</td></tr> </table> <pre> ans(:,:,5) = </pre> <table border="0"> <tr><td>6.9235e-05</td></tr> </table> <pre> ans(:,:,6) = </pre> <table border="0"> <tr><td>-4.5045e-05</td></tr> </table> <pre> ans(:,:,7) = </pre> <table border="0"> <tr><td>-3.8534e-05</td></tr> </table> <pre> ans(:,:,8) = </pre> <table border="0"> <tr><td>7.1908e-06</td></tr> </table> <pre> ans(:,:,9) = </pre> <table border="0"> <tr><td>-1.7569e-05</td></tr> </table> <pre> ans(:,:,10) = </pre> <table border="0"> <tr><td>5.6374e-05</td></tr> </table>	9.1119e-05	-3.4369e-05	1.3568e-06	-5.0792e-05	6.9235e-05	-4.5045e-05	-3.8534e-05	7.1908e-06	-1.7569e-05	5.6374e-05
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5.6374e-05																																																																										

ans (:,:,2,3) =		ans (:,:,11) =
0.0821 -0.0212 -0.0755 0.0551 0.0543 -0.0109 -0.0373 0.0214 0.0795		-6.2490e-05
ans (:,:,3,3) =		ans (:,:,12) =
-0.0537 0.0287 0.0199 0.0200 -0.0893 0.0914 -0.0606 0.0268 0.0798		-3.4067e-05
ans (:,:,1,4) =		ans (:,:,13) =
0.0260 -0.0120 0.0468 -0.0030 -0.0955 -0.0022 0.0937 0.0702 0.0498		-4.2153e-05
ans (:,:,2,4) =		ans (:,:,14) =
0.0547 0.0605 0.0059 -0.0797 -0.0316 -0.0286 -0.0166 0.0054 -0.0334		1.2327e-05
ans (:,:,3,4) =		ans (:,:,15) =
0.0535 0.0103 0.0763 0.0941 -0.0308 -0.0526 0.0237 0.0257 0.0546		1.3591e-05
ans (:,:,1,5) =		ans (:,:,16) =
-0.0111 0.0864 0.0576 -0.0453 0.0863 0.0368 -0.0531 0.0771 0.0257		-1.6440e-06
ans (:,:,2,5) =		ans (:,:,17) =
-0.0620 0.0154 -0.0773 0.0789 -0.0362 0.0362 -0.0994 -0.0701 -0.0982		-4.2746e-05
ans (:,:,3,5) =		ans (:,:,18) =
0.0558 -0.0667 0.0876 -0.0660 -0.0893 -0.0147 -0.0539 -0.0918 -0.1004		1.4308e-04
ans (:,:,1,6) =		ans (:,:,19) =
		6.3387e-05
		ans (:,:,20) =
		-3.2276e-05
		ans (:,:,21) =
		7.9016e-05
		ans (:,:,22) =

0.0423 -0.0203 0.0428	-6.3733e-05
-0.0108 -0.0443 0.0109	
-0.0513 -0.0177 -0.0860	
ans (:,:,2,6) =	
0.0690 0.0734 0.0430	-5.9651e-05
-0.0003 0.0426 0.0108	
0.0916 0.0390 0.0740	
ans (:,:,3,6) =	
-0.0021 -0.0962 -0.0266	ans (:,:,25) =
-0.0873 -0.0408 -0.0415	-9.0131e-05
-0.0561 -0.0947 0.0116	
ans (:,:,1,7) =	
-0.0093 -0.0415 -0.0250	ans (:,:,26) =
0.0070 -0.0640 -0.0100	-6.9201e-05
-0.0835 -0.0533 0.0386	
ans (:,:,2,7) =	
-0.0051 -0.0553 0.0248	ans (:,:,28) =
-0.0159 0.0867 -0.0830	9.5067e-05
-0.0869 0.0837 -0.0566	
ans (:,:,3,7) =	
-0.0751 -0.0543 0.0066	ans (:,:,29) =
0.0476 0.0940 0.0785	-3.9071e-05
-0.0804 0.0364 0.0274	
ans (:,:,1,8) =	
0.0786 -0.0575 -0.0121	ans (:,:,30) =
0.0792 -0.0754 -0.0043	-7.2953e-05
0.0851 -0.0705 -0.0350	
ans (:,:,2,8) =	
-0.0540 0.0520 -0.0757	ans (:,:,31) =
-0.0297 -0.0455 0.0271	-4.3349e-05
-0.0693 0.0306 0.0648	
ans (:,:,3,8) =	
	ans (:,:,32) =
	-3.2519e-05
	ans (:,:,33) =
	1.6067e-05

0.0980	0.0214	-0.0121	
0.0472	0.0505	-0.0303	
-0.0246	0.0525	0.0457	
			ans(:,:,34) =
			-3.2450e-05
ans(:,:,1,9) =			
-0.0334	-0.0028	0.0815	ans(:,:,35) =
0.0534	0.0134	-0.0247	-1.3925e-05
-0.0271	0.0244	0.0706	
			ans(:,:,36) =
			-1.8199e-05
ans(:,:,2,9) =			
-0.0541	-0.0831	-0.0800	ans(:,:,37) =
0.0646	0.0574	-0.0116	3.8500e-05
-0.0560	0.0195	-0.0941	
			ans(:,:,38) =
ans(:,:,3,9) =			1.1234e-04
-0.0054	-0.0903	-0.0295	
-0.0524	0.0023	0.0919	ans(:,:,39) =
0.0763	0.0496	-0.0851	-4.3854e-05
			ans(:,:,40) =
ans(:,:,1,10) =			8.1205e-05
-0.0228	-0.0899	-0.0496	
-0.0442	-0.0489	0.0174	ans(:,:,41) =
-0.0603	0.0042	-0.0898	-1.5705e-05
			ans(:,:,42) =
ans(:,:,2,10) =			-1.8429e-05
0.0099	-0.0353	-0.0925	
0.0864	-0.0485	-0.0751	ans(:,:,43) =
-0.0510	0.0473	-0.0451	-3.3959e-05
			ans(:,:,44) =
ans(:,:,3,10) =			1.2594e-04
-0.0173	-0.0328	-0.0206	
-0.0675	0.0091	0.0450	
-0.0107	-0.0100	0.0425	
ans(:,:,1,11) =			
-0.0432	-0.0726	-0.0466	
-0.0100	-0.0311	0.0802	
-0.0888	-0.0733	0.0668	
ans(:,:,2,11) =			
0.0786	-0.0517	-0.0219	

-0.0414 0.0927 -0.0009 0.0629 -0.0529 0.0238	ans (:,:,45) = -3.7039e-05
ans (:,:,3,11) = 0.0475 -0.0172 0.0638 -0.0223 -0.0937 -0.0489 0.0108 0.0622 0.0093	ans (:,:,46) = -3.8941e-05
ans (:,:,1,12) = 0.0986 -0.0920 -0.0754 -0.0146 0.0147 0.0459 -0.0588 0.0708 0.0379	ans (:,:,47) = 1.0084e-04
ans (:,:,2,12) = -0.0037 0.0783 0.0786 0.0753 0.0996 0.0261 0.0160 0.0626 0.0241	ans (:,:,48) = -2.1202e-05
ans (:,:,3,12) = -0.0616 -0.0374 -0.0518 -0.0480 0.0845 -0.0719 -0.0109 -0.0095 -0.0087	ans (:,:,49) = -2.6101e-05
ans (:,:,1,13) = 0.0728 0.0262 -0.0873 0.1010 -0.0282 -0.0298 -0.0417 0.0605 0.0448	ans (:,:,50) = 2.5839e-05
ans (:,:,2,13) = -0.0009 -0.0241 -0.0714 0.0273 0.0032 -0.0260 0.0807 0.0375 0.0487	ans (:,:,51) = -2.4330e-06
ans (:,:,3,13) = -0.0103 0.0241 0.0307 -0.0227 -0.0368 0.0375 -0.0923 0.0149 0.0204	ans (:,:,52) = -3.1458e-05
ans (:,:,1,14) = 0.0263 -0.0927 0.0971 0.0431 0.0201 0.0405	ans (:,:,53) = 5.5861e-05
	ans (:,:,54) = -3.3608e-05
	ans (:,:,55) = -2.7474e-06
	ans (:,:,56) =

-0.0521	0.0764	-0.0081	1.8676e-05
ans(:,:,2,14) =			ans(:,:,57) =
-0.0299	-0.0983	0.0890	1.3997e-05
-0.0629	0.0482	0.0871	
-0.0605	0.0802	-0.0284	ans(:,:,58) =
ans(:,:,3,14) =			2.2623e-04
0.0810	-0.0572	-0.0970	
0.0453	-0.0554	0.0052	ans(:,:,59) =
0.0288	-0.0172	-0.0266	-2.1785e-05
ans(:,:,1,15) =			ans(:,:,60) =
0.0525	0.0009	0.0755	1.2869e-04
0.0632	-0.0365	-0.0610	
0.0624	0.0163	-0.0498	ans(:,:,61) =
ans(:,:,2,15) =			-2.7603e-05
0.0477	-0.0017	0.0483	
0.0435	-0.0001	0.0755	ans(:,:,62) =
-0.0494	0.0295	-0.0178	-2.9783e-06
ans(:,:,3,15) =			
0.0846	0.0443	-0.0804	ans(:,:,63) =
0.0050	-0.0555	-0.0087	5.1221e-06
0.0238	0.0254	-0.0303	
ans(:,:,1,16) =			ans(:,:,64) =
0.0492	0.0033	-0.0040	9.6376e-06
0.0484	0.0752	0.0879	
0.0287	0.0659	-0.0931	
ans(:,:,2,16) =			
-0.0717	-0.0772	0.0245	
-0.0998	-0.0516	0.0600	
0.0572	-0.0053	-0.0919	
ans(:,:,3,16) =			
0.0068	0.0903	-0.0433	
-0.0252	0.0642	0.0746	
0.0065	-0.0111	-0.0487	

```

ans (:,:,1,17) =
0.0915   -0.0516    0.0989
0.0238   -0.0917    0.0647
0.0987   0.0178   -0.0284

ans (:,:,2,17) =
0.0894   -0.0337   -0.0382
0.1002   -0.0614   -0.0137
0.0812   -0.0575    0.0523

ans (:,:,3,17) =
0.0473    0.1011   -0.0317
0.0842    0.0540    0.0956
-0.0450   0.0797    0.0164

ans (:,:,1,18) =
0.0112    0.0880   -0.0721
-0.0653    0.0262    0.0526
0.0713    0.0664   -0.0472

ans (:,:,2,18) =
0.0512   -0.0889    0.0673
0.0781    0.0069    0.0285
-0.0410   -0.0932   -0.0865

ans (:,:,3,18) =
-0.0363   -0.0612   -0.0674
0.0073    0.0497   -0.0677
-0.0242   0.0070   -0.0259

ans (:,:,1,19) =
-0.0443    0.0071   -0.0215
0.0326    0.0344   -0.0726
0.0910   -0.0841    0.0455

ans (:,:,2,19) =
-0.0432   -0.0985   -0.0673
0.0378   -0.0577    0.0393
0.0909   -0.0752   -0.0154

```

```

ans (:,:,3,19) =
0.0631    0.0266   -0.0616
0.0249    -0.0062   -0.0635
-0.0382   -0.0959   0.0154

ans (:,:,1,20) =
0.0514    0.0623   0.0933
0.1012    -0.0127   -0.0390
0.0296    0.0335   -0.0421

ans (:,:,2,20) =
0.0380    -0.0401   0.0743
-0.0932   -0.0183   -0.0658
0.0656    0.0691   0.0097

ans (:,:,3,20) =
-0.0443    0.0151   -0.0185
-0.0539    0.0160   -0.0938
0.0678    -0.0377   -0.0187

ans (:,:,1,21) =
-0.0995    0.0505   0.0260
-0.0382   -0.0935   -0.0828
0.0440    -0.0229   -0.0869

ans (:,:,2,21) =
-0.0231    0.0484   -0.0893
0.0417    0.0561   -0.0161
-0.0468   -0.0065   -0.0628

ans (:,:,3,21) =
0.0397    -0.0474   -0.0297
0.0249    0.0003   -0.0219
0.0364    0.0549   0.0893

ans (:,:,1,22) =
0.0947    0.0060   0.0068
-0.0162   -0.0524   0.0255
0.0980    0.0663   0.0201

```

<pre>ans (:,:,2,22) = -0.0422 0.0098 0.0790 0.0657 0.0604 -0.0048 -0.0109 -0.0280 -0.0596</pre>	
<pre>ans (:,:,3,22) = 0.0179 0.0266 -0.0620 0.0673 -0.0728 0.0893 -0.0701 0.0189 0.0874</pre>	
<pre>ans (:,:,1,23) = 0.0784 -0.0961 0.0105 -0.0931 -0.0803 0.0837 0.0147 0.0982 0.0441</pre>	
<pre>ans (:,:,2,23) = -0.0017 -0.0164 -0.0508 0.0050 -0.0496 -0.0244 -0.0083 -0.0515 0.0921</pre>	
<pre>ans (:,:,3,23) = -0.0625 0.0750 -0.0913 0.0873 -0.0252 0.0979 -0.0961 0.0663 0.0100</pre>	
<pre>ans (:,:,1,24) = 0.0892 -0.0342 0.0963 0.0718 -0.0542 0.0710 -0.0777 -0.0220 -0.0713</pre>	
<pre>ans (:,:,2,24) = -0.0732 -0.0182 0.0984 -0.0357 0.0348 -0.0908 -0.0252 0.0829 0.0930</pre>	
<pre>ans (:,:,3,24) = 0.0892 -0.0978 -0.0209 -0.0671 -0.0385 -0.0685 0.0404 0.0152 0.0126</pre>	
<pre>ans (:,:,1,25) =</pre>	

```
-0.0726    0.0726    0.0080  
 0.0226   -0.0226    0.0500  
-0.0958    0.1009    0.0561
```

```
ans (:,:,2,25) =  
  
 0.0915    0.0598    0.0541  
 0.0553   -0.0541    0.0478  
 0.0760    0.0603   -0.0031
```

```
ans (:,:,3,25) =  
  
 0.1012    0.0815    0.0429  
-0.0935    0.0526    0.0673  
 0.0380   -0.0300    0.0325
```

```
ans (:,:,1,26) =  
  
-0.0730   -0.0134   -0.0887  
-0.0745   -0.0167    0.0309  
-0.0876    0.0186   -0.0327
```

```
ans (:,:,2,26) =  
  
 0.0793   -0.0906    0.0612  
-0.0190    0.0403    0.0577  
-0.0216    0.0047   -0.0876
```

```
ans (:,:,3,26) =  
  
 0.0332   -0.0194   -0.0729  
 0.0825    0.0506   -0.0731  
-0.0157   -0.0172    0.0939
```

```
ans (:,:,1,27) =  
  
 0.0852    0.0150    0.0609  
 0.0814    0.0919   -0.0478  
-0.0810   -0.0489   -0.0821
```

```
ans (:,:,2,27) =  
  
-0.0043    0.0281   -0.0549  
 0.0258    0.0161    0.0292  
 0.0493   -0.0339    0.0162
```

```
ans (:,:,3,27) =
```

$\begin{array}{ccc} -0.0973 & 0.0664 & 0.0926 \\ -0.0846 & 0.0888 & -0.0973 \\ 0.0505 & 0.0543 & 0.0429 \end{array}$	
$\text{ans}(:,:,1,28) =$ $\begin{array}{ccc} 0.0113 & -0.0271 & 0.0812 \\ -0.0062 & 0.0773 & 0.0393 \\ -0.0522 & -0.0866 & 0.0337 \end{array}$	
$\text{ans}(:,:,2,28) =$ $\begin{array}{ccc} -0.0684 & -0.0088 & 0.0544 \\ -0.0963 & -0.0383 & -0.0141 \\ -0.0890 & -0.0579 & 0.0042 \end{array}$	
$\text{ans}(:,:,3,28) =$ $\begin{array}{ccc} -0.0435 & 0.0967 & -0.0890 \\ 0.0284 & -0.0381 & 0.0627 \\ -0.0021 & 0.0645 & -0.0713 \end{array}$	
$\text{ans}(:,:,1,29) =$ $\begin{array}{ccc} -0.0236 & 0.0034 & 0.0928 \\ 0.0483 & -0.0165 & 0.0146 \\ -0.0122 & -0.0476 & 0.0332 \end{array}$	
$\text{ans}(:,:,2,29) =$ $\begin{array}{ccc} 0.0795 & -0.0010 & -0.0495 \\ -0.0128 & 0.0355 & 0.0866 \\ 0.0645 & -0.0009 & -0.0922 \end{array}$	
$\text{ans}(:,:,3,29) =$ $\begin{array}{ccc} 0.0477 & -0.0621 & 0.1011 \\ 0.0869 & 0.0173 & 0.0167 \\ 0.0495 & -0.0054 & 0.0647 \end{array}$	
$\text{ans}(:,:,1,30) =$ $\begin{array}{ccc} -0.0584 & -0.0905 & -0.0844 \\ -0.0856 & -0.0478 & 0.0485 \\ 0.0892 & 0.0000 & -0.0697 \end{array}$	
$\text{ans}(:,:,2,30) =$ $0.0016 \quad 0.0667 \quad -0.0438$	

$\begin{array}{ccc} -0.0891 & -0.0635 & -0.0221 \\ -0.0510 & -0.0862 & -0.0933 \end{array}$	
$\text{ans}(:,:,3,30) =$ $\begin{array}{ccc} -0.0051 & -0.0147 & 0.0496 \\ 0.0511 & 0.0089 & 0.0094 \\ 0.0864 & 0.0967 & -0.0239 \end{array}$	
$\text{ans}(:,:,1,31) =$ $\begin{array}{ccc} 0.0850 & -0.0096 & 0.0053 \\ -0.0670 & 0.0889 & 0.0801 \\ -0.0608 & -0.0230 & -0.0046 \end{array}$	
$\text{ans}(:,:,2,31) =$ $\begin{array}{ccc} -0.0070 & 0.0771 & -0.0896 \\ 0.0309 & 0.0024 & -0.0039 \\ -0.0608 & 0.0100 & -0.0757 \end{array}$	
$\text{ans}(:,:,3,31) =$ $\begin{array}{ccc} 0.0366 & 0.0636 & 0.0582 \\ 0.0935 & 0.0701 & 0.0437 \\ 0.0772 & 0.0664 & 0.0286 \end{array}$	
$\text{ans}(:,:,1,32) =$ $\begin{array}{ccc} 0.0045 & 0.0403 & 0.0702 \\ -0.0651 & 0.0424 & -0.0657 \\ 0.0931 & 0.0640 & 0.0034 \end{array}$	
$\text{ans}(:,:,2,32) =$ $\begin{array}{ccc} 0.0915 & -0.0551 & 0.0556 \\ -0.0447 & 0.0981 & 0.0343 \\ 0.0825 & -0.0278 & 0.0529 \end{array}$	
$\text{ans}(:,:,3,32) =$ $\begin{array}{ccc} 0.0126 & 0.0858 & 0.0233 \\ 0.0384 & -0.0630 & -0.0219 \\ 0.0208 & -0.0158 & -0.0205 \end{array}$	
$\text{ans}(:,:,1,33) =$ $\begin{array}{ccc} 0.0785 & -0.0901 & 0.0177 \\ -0.0618 & 0.0044 & 0.0828 \end{array}$	

<pre> -0.0104 0.0521 0.0156 ans (:,:,2,33) = </pre>	
<pre> 0.0274 0.0931 -0.0416 0.0939 0.0515 -0.0793 -0.0125 -0.0608 -0.0671 </pre>	
<pre> ans (:,:,3,33) = </pre>	
<pre> 0.0543 0.0927 0.0001 0.0529 0.0925 0.0775 -0.0171 -0.0612 -0.0138 </pre>	
<pre> ans (:,:,1,34) = </pre>	
<pre> -0.0084 0.0076 -0.0102 -0.0737 0.0689 0.0197 -0.0589 -0.0761 0.0469 </pre>	
<pre> ans (:,:,2,34) = </pre>	
<pre> 0.0695 -0.0390 0.0629 -0.0854 0.0604 0.0563 -0.0814 0.1020 0.0719 </pre>	
<pre> ans (:,:,3,34) = </pre>	
<pre> 0.0127 -0.0864 0.0038 0.0422 -0.0865 0.0908 0.0893 0.0905 0.0845 </pre>	
<pre> ans (:,:,1,35) = </pre>	
<pre> -0.0571 0.0108 0.0502 0.0235 0.0970 -0.0474 0.0490 0.0856 0.0894 </pre>	
<pre> ans (:,:,2,35) = </pre>	
<pre> -0.0863 0.0586 0.0633 -0.0396 -0.0543 0.0317 -0.0588 0.0030 -0.0721 </pre>	
<pre> ans (:,:,3,35) = </pre>	
<pre> 0.0364 0.0612 -0.0951 0.0328 -0.0680 0.0094 -0.0528 -0.0650 0.0082 </pre>	

```

ans (:,:,1,36) =
-0.0303    0.0659    0.0189
 0.0623    0.0507   -0.0198
 0.0461   -0.0626    0.0358

ans (:,:,2,36) =
-0.0160    0.0351    0.0725
 0.0314   -0.0945    0.0501
 0.0663    0.0366    0.0811

ans (:,:,3,36) =
-0.0550   -0.0287    0.0408
 0.0189    0.1001    0.0181
 -0.0259    0.0469   -0.0969

ans (:,:,1,37) =
-0.0912    0.0261   -0.0081
 -0.0045   -0.0027    0.0259
 -0.0290   -0.0449    0.0153

ans (:,:,2,37) =
 0.0231   -0.0649   -0.0620
  0.0734   -0.0443   -0.0950
 -0.0041    0.0832    0.0553

ans (:,:,3,37) =
-0.0706   -0.0643    0.0100
 -0.0331    0.0727   -0.0880
 -0.0482   -0.0285   -0.0337

ans (:,:,1,38) =
-0.0524   -0.0192    0.0435
 -0.0463   -0.0360   -0.0952
  0.0383    0.0061   -0.0011

ans (:,:,2,38) =
 0.0414    0.0273   -0.0817
 -0.0035   -0.0176   -0.0326
  0.0799   -0.0833   -0.0922

```

```

ans (:,:,3,38) =
-0.0918    0.0100   -0.0513
-0.0963   -0.0161    0.0860
-0.0170   -0.0011   -0.0775

ans (:,:,1,39) =
0.0808   -0.0038    0.0124
0.0946   -0.0791    0.0759
0.0633   -0.0131    0.0444

ans (:,:,2,39) =
0.0244   -0.0418   -0.0843
0.0370    0.0744    0.0864
0.0361   -0.0650   -0.0980

ans (:,:,3,39) =
0.0226    0.0481   -0.0870
-0.0611   -0.0589    0.0360
0.0111    0.0639   -0.0406

ans (:,:,1,40) =
-0.0061    0.0132   -0.0060
0.0740   -0.0226    0.0395
0.0364    0.0388   -0.0868

ans (:,:,2,40) =
-0.0978   -0.0641   -0.0521
-0.0700    0.0505   -0.0489
-0.0971   0.0438   -0.0071

ans (:,:,3,40) =
-0.0198    0.0974    0.0107
0.0932   -0.0772   -0.0201
-0.0776    0.0266   -0.0359

ans (:,:,1,41) =
0.0358   -0.0765    0.0892
-0.0644   -0.0570   -0.0627
-0.0753    0.0862    0.0957

```

<pre>ans (:,:,2,41) = 0.0860 0.0011 0.0118 0.0569 0.0234 -0.0288 -0.0327 0.0212 -0.0912</pre>	
<pre>ans (:,:,3,41) = 0.0824 0.0545 -0.0378 0.0203 -0.0236 0.0634 -0.0090 0.0385 0.0198</pre>	
<pre>ans (:,:,1,42) = 0.0638 0.0634 -0.0211 -0.0695 0.0074 0.0553 0.0306 0.0376 0.0857</pre>	
<pre>ans (:,:,2,42) = -0.0492 -0.0771 0.0226 -0.0672 0.0338 -0.0350 0.0929 0.0488 0.0918</pre>	
<pre>ans (:,:,3,42) = 0.0859 -0.0071 -0.0118 0.0057 0.0014 0.1034 0.0055 -0.0052 -0.0336</pre>	
<pre>ans (:,:,1,43) = 0.0197 0.0731 -0.0160 -0.0841 0.0281 0.0927 -0.0082 0.0733 0.0045</pre>	
<pre>ans (:,:,2,43) = 0.0897 0.0772 0.0495 0.0593 0.0314 -0.0855 0.0150 0.0702 0.0561</pre>	
<pre>ans (:,:,3,43) = -0.0586 0.0922 -0.0546 0.0493 0.0056 0.0758 -0.0519 0.0911 -0.0526</pre>	
<pre>ans (:,:,1,44) =</pre>	

```
-0.1007 -0.0067 -0.0597  
0.0169 0.0401 -0.0823  
-0.0456 0.0164 -0.0309
```

```
ans (:,:,2,44) =  
  
-0.0330 -0.0321 -0.0014  
-0.0955 0.0433 0.0299  
-0.0998 0.0320 0.0577
```

```
ans (:,:,3,44) =  
  
0.0279 0.0920 -0.0967  
0.0409 0.0727 -0.0356  
0.0603 -0.0403 -0.0426
```

```
ans (:,:,1,45) =  
  
0.0893 0.0500 0.0231  
0.0088 -0.0541 0.0900  
0.0534 0.0718 0.0917
```

```
ans (:,:,2,45) =  
  
-0.0576 -0.0799 -0.0839  
0.0633 -0.0041 -0.0138  
0.0840 -0.0255 -0.0113
```

```
ans (:,:,3,45) =  
  
0.0256 0.0024 -0.0818  
0.0067 0.0606 0.0880  
-0.0351 -0.0149 -0.0909
```

```
ans (:,:,1,46) =  
  
0.0614 0.0862 0.0217  
0.0005 -0.0593 -0.0723  
0.0400 0.0356 -0.0339
```

```
ans (:,:,2,46) =  
  
-0.0008 -0.0186 -0.0578  
-0.0150 0.0270 -0.0595  
0.0885 -0.0039 -0.0710
```

```
ans (:,:,3,46) =
```

0.0242	-0.0194	0.0070
-0.0805	0.0586	-0.0429
0.0758	0.0276	-0.0250

ans (:,:,1,47) =

0.0796	0.0484	0.0654
-0.0700	-0.0532	0.0045
-0.0883	-0.0376	0.0514

ans (:,:,2,47) =

-0.0651	-0.0532	0.0385
0.0544	-0.0759	0.0248
0.0441	-0.0625	0.0728

ans (:,:,3,47) =

-0.0487	-0.0446	-0.0435
0.0399	-0.0950	0.0218
-0.0311	-0.0692	-0.0530

ans (:,:,1,48) =

0.0132	-0.0759	-0.0035
0.0011	0.0720	0.0326
-0.0279	-0.0492	-0.0290

ans (:,:,2,48) =

-0.0873	0.0236	-0.0593
0.0640	0.0451	0.0781
0.0368	-0.0176	0.0278

ans (:,:,3,48) =

0.0402	-0.0702	-0.0758
-0.0236	0.0140	0.0511
-0.0425	-0.0126	0.0142

ans (:,:,1,49) =

0.0760	-0.0256	0.0273
0.0991	0.0128	0.0982
-0.0427	0.0192	-0.0819

ans (:,:,2,49) =

0.0691	-0.0322	-0.0007
--------	---------	---------

<pre> 0.0867 -0.0469 0.0848 -0.0335 0.0209 -0.0060 </pre>	
<pre> ans (:,:,3,49) = </pre>	
<pre> 0.0744 0.0360 0.0690 0.0338 -0.0401 -0.0090 -0.0967 0.0443 0.0883 </pre>	
<pre> ans (:,:,1,50) = </pre>	
<pre> -0.0560 0.0466 0.0525 0.0967 0.0734 0.0323 -0.0575 0.0966 -0.0625 </pre>	
<pre> ans (:,:,2,50) = </pre>	
<pre> 0.0945 -0.0672 0.0000 0.0835 -0.0466 -0.0334 -0.0873 0.0858 0.0461 </pre>	
<pre> ans (:,:,3,50) = </pre>	
<pre> -0.0255 -0.0441 0.0379 -0.0348 -0.0940 -0.0482 -0.0058 -0.0426 -0.0394 </pre>	
<pre> ans (:,:,1,51) = </pre>	
<pre> -0.0332 0.0507 0.0593 0.0806 -0.0909 -0.0431 -0.0287 0.0355 0.0395 </pre>	
<pre> ans (:,:,2,51) = </pre>	
<pre> 0.0344 0.0836 0.0825 -0.0935 -0.0125 0.0987 0.0074 -0.0867 -0.0905 </pre>	
<pre> ans (:,:,3,51) = </pre>	
<pre> -0.0548 0.0585 -0.0718 0.0455 -0.0385 0.0666 0.0305 0.0468 0.0834 </pre>	
<pre> ans (:,:,1,52) = </pre>	
<pre> -0.0508 0.0098 -0.0326 -0.0464 0.0224 0.0856 </pre>	

<pre> -0.0524 0.0125 0.0331 </pre>	
<pre> ans(:,:,2,52) = </pre>	
<pre> 0.0739 -0.0329 0.0245 0.0377 0.0597 0.0216 0.0512 0.0286 -0.0485 </pre>	
<pre> ans(:,:,3,52) = </pre>	
<pre> -0.0095 -0.0119 0.0952 0.0137 -0.0242 -0.0404 -0.0143 0.0331 -0.0790 </pre>	
<pre> ans(:,:,1,53) = </pre>	
<pre> 0.0503 -0.0648 -0.0339 0.0491 0.0887 0.0967 0.0280 0.0087 -0.0620 </pre>	
<pre> ans(:,:,2,53) = </pre>	
<pre> -0.0708 -0.0854 0.0695 -0.0632 0.0382 0.0721 -0.0951 0.0776 0.0418 </pre>	
<pre> ans(:,:,3,53) = </pre>	
<pre> -0.0696 -0.0081 0.0615 -0.0174 0.0465 -0.0249 -0.0534 -0.0048 -0.0922 </pre>	
<pre> ans(:,:,1,54) = </pre>	
<pre> 0.0884 0.0295 0.0882 -0.0305 -0.0730 -0.0097 0.0851 0.0995 0.0719 </pre>	
<pre> ans(:,:,2,54) = </pre>	
<pre> 0.0938 -0.0353 -0.0340 -0.0503 0.0972 0.0654 0.0315 -0.0604 -0.0651 </pre>	
<pre> ans(:,:,3,54) = </pre>	
<pre> -0.0620 0.0331 0.0497 -0.0093 0.0722 0.0320 0.0168 0.0806 -0.0226 </pre>	

```

ans (:,:,1,55) =
-0.0367    0.0226   -0.0423
 0.0545   -0.0107    0.0929
 0.0007   -0.0127   -0.0482

ans (:,:,2,55) =
-0.0363   -0.0749    0.0132
 0.0906    0.0203    0.0851
 0.0843    0.0927    0.0763

ans (:,:,3,55) =
-0.0905    0.0601    0.0433
 0.0451    0.0391    0.0311
 -0.0823   -0.0727   -0.0374

ans (:,:,1,56) =
-0.0546    0.0750    0.0210
 0.0835    0.0087    0.0892
 -0.0949   -0.0778   -0.0238

ans (:,:,2,56) =
 0.0752    0.0534   -0.0116
  0.0110   -0.0917    0.0065
  0.0251   -0.0377   -0.0786

ans (:,:,3,56) =
 0.0687    0.0089    0.0227
 -0.0742   -0.0030   -0.0794
  0.0060   -0.0966   -0.0113

ans (:,:,1,57) =
-0.0943    0.0398    0.0410
 0.0540   -0.0219   -0.0415
  0.0058    0.0144    0.0146

ans (:,:,2,57) =
-0.0319    0.0322    0.0793
 -0.0419    0.0383   -0.0981
  0.0813    0.0730    0.0901

```

```

ans (:,:,3,57) =
-0.0295   -0.0227   -0.0032
-0.0461   -0.0992    0.0210
-0.0672    0.0961    0.0755

ans (:,:,1,58) =
 0.0898    0.0243   -0.0387
-0.0352    0.0817   -0.0948
-0.0332   -0.0388    0.0181

ans (:,:,2,58) =
-0.0644   -0.0436   -0.0777
-0.0715    0.0688   -0.0181
-0.1010   -0.0598    0.0476

ans (:,:,3,58) =
-0.0352   -0.0773   -0.0566
-0.0298    0.0888   -0.0733
 0.0407   -0.0062   -0.0771

ans (:,:,1,59) =
 0.0431    0.0113    0.0263
 0.0451    0.0957   -0.0946
 0.0246    0.0591    0.0566

ans (:,:,2,59) =
-0.0820   -0.0187    0.0522
-0.0518   -0.0782    0.0391
-0.0061   -0.0010    0.0044

ans (:,:,3,59) =
-0.0362    0.0699   -0.0472
-0.0778   -0.0569    0.0847
 0.0249    0.0186   -0.0635

ans (:,:,1,60) =
-0.0070    0.0051   -0.0418
-0.1006   -0.0975    0.0858
 0.0171    0.0350    0.0546

```

<pre>ans (:,:,2,60) = 0.0058 -0.0558 -0.0451 -0.0555 -0.0694 0.0716 0.0939 -0.0509 -0.0458</pre>	
<pre>ans (:,:,3,60) = -0.0848 -0.0910 0.0285 0.0760 -0.0640 0.0544 -0.0971 -0.0106 -0.0563</pre>	
<pre>ans (:,:,1,61) = -0.0941 -0.0012 0.0988 0.0907 0.0516 0.0358 -0.0022 0.0652 -0.0257</pre>	
<pre>ans (:,:,2,61) = 0.0768 -0.0160 0.0674 -0.0103 0.0072 -0.0029 -0.0297 0.0509 0.0776</pre>	
<pre>ans (:,:,3,61) = -0.0455 0.0129 0.0927 0.0648 -0.0762 0.0775 0.0829 -0.0090 0.0528</pre>	
<pre>ans (:,:,1,62) = 0.0764 -0.0568 0.0072 0.0785 0.0736 0.0774 -0.0467 -0.0950 0.0034</pre>	
<pre>ans (:,:,2,62) = 0.0482 -0.0854 0.0866 -0.0040 0.0549 -0.0604 -0.0888 0.0594 0.0779</pre>	
<pre>ans (:,:,3,62) = 0.0719 0.0659 0.0349 -0.0542 -0.0538 0.0882 -0.0487 -0.0205 -0.0730</pre>	
<pre>ans (:,:,1,63) =</pre>	

$\begin{array}{ccc} -0.0692 & -0.0417 & -0.0891 \\ -0.0954 & -0.0234 & 0.0572 \\ 0.0660 & 0.0236 & -0.0803 \end{array}$	
$\text{ans}(:,:,2,63) =$ $\begin{array}{ccc} -0.0217 & -0.0970 & -0.0314 \\ -0.0727 & 0.0855 & -0.0573 \\ 0.0349 & -0.0713 & 0.0465 \end{array}$	
$\text{ans}(:,:,3,63) =$ $\begin{array}{ccc} 0.0130 & -0.0474 & 0.0083 \\ -0.0511 & -0.0320 & -0.0550 \\ 0.0226 & 0.0591 & 0.0731 \end{array}$	
$\text{ans}(:,:,1,64) =$ $\begin{array}{ccc} -0.0815 & 0.0122 & -0.0910 \\ 0.0437 & 0.0048 & -0.0144 \\ -0.0964 & 0.0864 & 0.0759 \end{array}$	
$\text{ans}(:,:,2,64) =$ $\begin{array}{ccc} 0.0820 & 0.0202 & 0.0686 \\ -0.0445 & 0.0812 & 0.0554 \\ 0.0290 & -0.0082 & -0.0133 \end{array}$	
$\text{ans}(:,:,3,64) =$ $\begin{array}{ccc} -0.0128 & -0.0563 & -0.0072 \\ 0.0840 & 0.0971 & 0.0044 \\ -0.0401 & 0.0776 & -0.0441 \end{array}$	

LAMPIRAN 10 – *Weight & Bias* Hasil Latihan ResNet-50

<pre>>> netTransfer.Layers(2).Weights 7×7×3×64 single array ans(:,:,1,1) = </pre> <table border="0"> <tr><td>-0.0273</td><td>-0.0316</td><td>0.0021</td><td>-0.0237</td></tr> <tr><td>0.0015</td><td>-0.0283</td><td>0.0506</td><td></td></tr> <tr><td>0.0473</td><td>-0.0251</td><td>0.0351</td><td>0.0162</td></tr> <tr><td>-0.0424</td><td>0.0278</td><td>-0.0021</td><td></td></tr> <tr><td>-0.0124</td><td>0.0320</td><td>-0.0055</td><td>-0.0218</td></tr> <tr><td>-0.0052</td><td>-0.0194</td><td>0.0206</td><td></td></tr> <tr><td>-0.0072</td><td>0.0210</td><td>0.0079</td><td>0.0433</td></tr> <tr><td>0.0316</td><td>0.0532</td><td>0.0157</td><td></td></tr> <tr><td>0.0556</td><td>0.0343</td><td>0.0272</td><td>-0.0208</td></tr> <tr><td>0.0012</td><td>-0.0282</td><td>-0.0047</td><td></td></tr> <tr><td>0.0067</td><td>-0.0099</td><td>0.0458</td><td>0.0093</td></tr> <tr><td>-0.0215</td><td>-0.0307</td><td>-0.0091</td><td></td></tr> <tr><td>-0.0003</td><td>-0.0047</td><td>-0.0341</td><td>-0.0168</td></tr> <tr><td>0.0375</td><td>-0.0001</td><td>-0.0235</td><td></td></tr> </table> <pre>ans(:,:,2,1) = </pre> <table border="0"> <tr><td>0.0131</td><td>0.0368</td><td>0.0255</td><td>0.0305</td></tr> <tr><td>-0.0148</td><td>-0.0247</td><td>0.0010</td><td></td></tr> <tr><td>-0.0194</td><td>0.0315</td><td>0.0478</td><td>0.0037</td></tr> <tr><td>0.0012</td><td>-0.0189</td><td>0.0330</td><td></td></tr> <tr><td>0.0451</td><td>0.0574</td><td>0.0170</td><td>-0.0253</td></tr> <tr><td>-0.0093</td><td>-0.0219</td><td>0.0450</td><td></td></tr> <tr><td>-0.0246</td><td>0.0379</td><td>0.0385</td><td>-0.0263</td></tr> <tr><td>0.0249</td><td>0.0273</td><td>0.0392</td><td></td></tr> <tr><td>0.0463</td><td>0.0524</td><td>-0.0024</td><td>0.0260</td></tr> <tr><td>0.0438</td><td>-0.0169</td><td>0.0149</td><td></td></tr> <tr><td>0.0312</td><td>0.0247</td><td>0.0438</td><td>-0.0235</td></tr> <tr><td>-0.0169</td><td>0.0488</td><td>0.0096</td><td></td></tr> <tr><td>-0.0071</td><td>0.0278</td><td>-0.0314</td><td>0.0386</td></tr> <tr><td>0.0526</td><td>0.0458</td><td>0.0195</td><td></td></tr> </table> <pre>ans(:,:,3,1) = </pre> <table border="0"> <tr><td>0.0347</td><td>0.0489</td><td>0.0487</td><td>-0.0214</td></tr> <tr><td>0.0082</td><td>0.0072</td><td>-0.0252</td><td></td></tr> <tr><td>-0.0099</td><td>-0.0076</td><td>0.0491</td><td>-0.0334</td></tr> <tr><td>0.0188</td><td>0.0134</td><td>0.0153</td><td></td></tr> <tr><td>0.0619</td><td>0.0036</td><td>0.0073</td><td>-0.0064</td></tr> <tr><td>0.0107</td><td>-0.0240</td><td>0.0152</td><td></td></tr> <tr><td>0.0511</td><td>-0.0074</td><td>0.0293</td><td>0.0179</td></tr> <tr><td>0.0050</td><td>0.0118</td><td>0.0307</td><td></td></tr> <tr><td>0.0045</td><td>-0.0090</td><td>0.0342</td><td>-0.0084</td></tr> <tr><td>-0.0192</td><td>0.0290</td><td>-0.0184</td><td></td></tr> </table>	-0.0273	-0.0316	0.0021	-0.0237	0.0015	-0.0283	0.0506		0.0473	-0.0251	0.0351	0.0162	-0.0424	0.0278	-0.0021		-0.0124	0.0320	-0.0055	-0.0218	-0.0052	-0.0194	0.0206		-0.0072	0.0210	0.0079	0.0433	0.0316	0.0532	0.0157		0.0556	0.0343	0.0272	-0.0208	0.0012	-0.0282	-0.0047		0.0067	-0.0099	0.0458	0.0093	-0.0215	-0.0307	-0.0091		-0.0003	-0.0047	-0.0341	-0.0168	0.0375	-0.0001	-0.0235		0.0131	0.0368	0.0255	0.0305	-0.0148	-0.0247	0.0010		-0.0194	0.0315	0.0478	0.0037	0.0012	-0.0189	0.0330		0.0451	0.0574	0.0170	-0.0253	-0.0093	-0.0219	0.0450		-0.0246	0.0379	0.0385	-0.0263	0.0249	0.0273	0.0392		0.0463	0.0524	-0.0024	0.0260	0.0438	-0.0169	0.0149		0.0312	0.0247	0.0438	-0.0235	-0.0169	0.0488	0.0096		-0.0071	0.0278	-0.0314	0.0386	0.0526	0.0458	0.0195		0.0347	0.0489	0.0487	-0.0214	0.0082	0.0072	-0.0252		-0.0099	-0.0076	0.0491	-0.0334	0.0188	0.0134	0.0153		0.0619	0.0036	0.0073	-0.0064	0.0107	-0.0240	0.0152		0.0511	-0.0074	0.0293	0.0179	0.0050	0.0118	0.0307		0.0045	-0.0090	0.0342	-0.0084	-0.0192	0.0290	-0.0184		<pre>>> netTransfer.Layers(2).Bias 1×1×64 single array ans(:,:,1) = </pre> <table border="0"> <tr><td>2.6659e-10</td></tr> </table> <pre>ans(:,:,2) = </pre> <table border="0"> <tr><td>4.9017e-10</td></tr> </table> <pre>ans(:,:,3) = </pre> <table border="0"> <tr><td>-8.1748e-10</td></tr> </table> <pre>ans(:,:,4) = </pre> <table border="0"> <tr><td>5.0024e-10</td></tr> </table> <pre>ans(:,:,5) = </pre> <table border="0"> <tr><td>9.2982e-10</td></tr> </table> <pre>ans(:,:,6) = </pre> <table border="0"> <tr><td>-1.1518e-10</td></tr> </table> <pre>ans(:,:,7) = </pre> <table border="0"> <tr><td>5.3356e-11</td></tr> </table> <pre>ans(:,:,8) = </pre> <table border="0"> <tr><td>8.8464e-10</td></tr> </table> <pre>ans(:,:,9) = </pre> <table border="0"> <tr><td>-1.7279e-10</td></tr> </table> <pre>ans(:,:,10) = </pre> <table border="0"> <tr><td>1.9054e-10</td></tr> </table>	2.6659e-10	4.9017e-10	-8.1748e-10	5.0024e-10	9.2982e-10	-1.1518e-10	5.3356e-11	8.8464e-10	-1.7279e-10	1.9054e-10
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0.0333 0.0038 0.0161 0.0155 0.0282 -0.0269 0.0375 0.0512 0.0278 0.0345 -0.0035 -0.0182 -0.0184 0.0422	ans (:,:,11) = -5.0674e-10
ans (:,:,1,2) = 0.0247 -0.0387 0.0234 0.0297 -0.0139 0.0305 0.0025 -0.0254 0.0451 0.0315 -0.0087 0.0303 0.0407 0.0508 -0.0193 0.0123 -0.0180 0.0150 -0.0105 0.0265 0.0291 0.0487 0.0538 0.0473 -0.0151 0.0211 0.0119 -0.0206 0.0505 0.0546 -0.0214 0.0438 0.0321 0.0065 0.0007 -0.0046 -0.0125 0.0213 0.0418 0.0346 0.0382 -0.0062 -0.0446 -0.0319 -0.0420 -0.0188 0.0457 0.0342 0.0087	ans (:,:,12) = -2.9782e-10 ans (:,:,13) = -1.1976e-09 ans (:,:,14) = -4.9338e-10 ans (:,:,15) = -1.2775e-11
ans (:,:,2,2) = 0.0154 0.0244 -0.0063 -0.0285 0.0383 0.0297 -0.0223 0.0221 -0.0026 0.0168 -0.0130 0.0360 0.0486 -0.0126 0.0499 0.0292 0.0152 0.0496 -0.0070 0.0375 -0.0238 0.0021 0.0116 0.0189 0.0126 0.0431 -0.0200 0.0434 0.0135 0.0440 0.0091 0.0210 0.0522 -0.0047 0.0506 0.0375 0.0124 0.0423 -0.0048 0.0326 0.0056 0.0613 0.0208 -0.0214 0.0389 0.0226 0.0135 -0.0080 -0.0094	ans (:,:,16) = 6.4724e-11 ans (:,:,17) = 4.6389e-10 ans (:,:,18) = 6.1284e-11 ans (:,:,19) = 4.2648e-10
ans (:,:,3,2) = -0.0435 -0.0191 -0.0252 -0.0339 0.0188 0.0086 0.0219 -0.0305 0.0136 0.0200 0.0314 -0.0012 0.0168 -0.0204 -0.0042 -0.0041 -0.0232 -0.0083 -0.0058 0.0100 -0.0126 0.0569 0.0205 0.0426 0.0133 0.0471 -0.0195 0.0038 0.0026 0.0481 -0.0125 -0.0186 -0.0034 -0.0072 0.0483 0.0177 -0.0138 0.0366 -0.0329 0.0201 -0.0076 0.0416	ans (:,:,20) = 4.3641e-10 ans (:,:,21) = -9.3051e-10 ans (:,:,22) =

$\begin{bmatrix} -0.0386 & -0.0032 & -0.0082 & 0.0073 \\ -0.0316 & -0.0289 & -0.0051 \end{bmatrix}$	$8.8418e-10$
$\text{ans}(:,:,1,3) =$ $\begin{bmatrix} 0.0075 & -0.0024 & -0.0066 & -0.0306 \\ 0.0066 & -0.0045 & 0.0539 \\ -0.0278 & 0.0334 & 0.0410 & -0.0222 \\ 0.0114 & -0.0011 & 0.0504 \\ -0.0179 & 0.0310 & -0.0315 & 0.0267 \\ 0.0159 & 0.0071 & 0.0199 \\ -0.0279 & -0.0105 & 0.0434 & 0.0288 \\ 0.0310 & 0.0337 & 0.0313 \\ 0.0063 & 0.0359 & 0.0154 & 0.0460 \\ 0.0187 & -0.0261 & 0.0335 \\ 0.0233 & 0.0369 & 0.0349 & 0.0490 \\ 0.0277 & 0.0529 & -0.0231 \\ -0.0007 & 0.0027 & 0.0143 & -0.0356 \\ 0.0279 & -0.0280 & 0.0372 \end{bmatrix}$	$\text{ans}(:,:,23) =$ $7.6651e-10$
$\text{ans}(:,:,2,3) =$ $\begin{bmatrix} 0.0179 & -0.0144 & 0.0101 & -0.0183 \\ 0.0156 & 0.0169 & 0.0177 \\ 0.0105 & 0.0276 & 0.0326 & -0.0018 \\ 0.0054 & -0.0228 & -0.0038 \\ 0.0134 & -0.0105 & -0.0351 & 0.0376 \\ 0.0359 & -0.0100 & 0.0136 \\ 0.0383 & 0.0286 & 0.0020 & 0.0168 \\ 0.0111 & -0.0181 & -0.0199 \\ -0.0158 & -0.0045 & 0.0277 & 0.0019 \\ 0.0153 & 0.0142 & 0.0333 \\ -0.0104 & 0.0018 & 0.0399 & 0.0441 \\ 0.0214 & 0.0309 & -0.0148 \\ 0.0054 & 0.0064 & 0.0264 & 0.0045 \\ -0.0342 & 0.0404 & 0.0362 \end{bmatrix}$	$\text{ans}(:,:,24) =$ $2.8492e-10$
$\text{ans}(:,:,3,3) =$ $\begin{bmatrix} 0.0014 & -0.0194 & -0.0252 & 0.0433 \\ 0.0011 & -0.0125 & 0.0087 \\ -0.0013 & -0.0152 & 0.0175 & -0.0375 \\ 0.0001 & 0.0360 & 0.0347 \\ 0.0422 & 0.0414 & 0.0348 & 0.0237 \\ 0.0226 & -0.0188 & 0.0423 \\ 0.0414 & 0.0022 & 0.0445 & 0.0368 \\ 0.0418 & 0.0492 & -0.0078 \\ 0.0001 & -0.0189 & 0.0436 & -0.0338 \\ 0.0318 & -0.0168 & 0.0560 \\ 0.0142 & 0.0162 & 0.0267 & -0.0070 \\ 0.0098 & -0.0209 & 0.0397 \\ 0.0007 & 0.0352 & 0.0085 & -0.0198 \\ -0.0368 & -0.0323 & 0.0016 \end{bmatrix}$	$\text{ans}(:,:,25) =$ $6.1528e-11$
	$\text{ans}(:,:,26) =$ $4.3168e-10$
	$\text{ans}(:,:,27) =$ $-2.8789e-10$
	$\text{ans}(:,:,28) =$ $6.1716e-10$
	$\text{ans}(:,:,29) =$ $2.6780e-10$
	$\text{ans}(:,:,30) =$ $5.6596e-10$
	$\text{ans}(:,:,31) =$ $-4.2421e-10$
	$\text{ans}(:,:,32) =$ $2.1446e-10$
	$\text{ans}(:,:,33) =$ $-4.3285e-10$

```

ans (:,:,1,4) =

```

-0.0564	-0.0047	0.0135	0.0400
0.0509	0.0522	-0.0098	
0.0111	0.0417	0.0339	0.0406
0.0248	0.0425	0.0407	
0.0103	0.0153	0.0543	-0.0261
-0.0108	0.0309	0.0326	
0.0300	0.0336	0.0367	-0.0089
-0.0003	-0.0286	0.0332	
-0.0295	0.0500	-0.0124	0.0088
0.0066	0.0283	0.0183	
-0.0098	0.0210	0.0452	-0.0144
-0.0375	0.0164	-0.0137	
0.0333	0.0470	0.0205	0.0208
-0.0553	-0.0599	0.0285	

```

ans (:,:,2,4) =

```

-0.0561	-0.0222	0.0252	0.0395
0.0303	0.0033	0.0391	
0.0121	-0.0041	0.0538	0.0518
0.0140	0.0107	-0.0059	
-0.0105	0.0069	-0.0023	-0.0199
-0.0023	0.0438	-0.0028	
-0.0052	0.0581	0.0165	-0.0024
0.0281	0.0204	-0.0090	
0.0016	0.0470	0.0453	-0.0284
-0.0237	-0.0091	0.0389	
0.0286	-0.0250	-0.0250	0.0218
-0.0225	-0.0334	0.0271	
0.0363	0.0283	-0.0100	-0.0543
-0.0420	-0.0322	-0.0331	

```

ans (:,:,3,4) =

```

-0.0344	0.0113	0.0213	0.0007
0.0530	0.0419	0.0686	
0.0113	-0.0045	0.0339	-0.0195
0.0308	-0.0029	-0.0010	
0.0425	0.0124	0.0200	-0.0042
0.0293	0.0072	0.0337	
-0.0123	0.0339	0.0119	0.0317
-0.0333	0.0263	-0.0103	
-0.0135	0.0449	-0.0082	0.0157
-0.0366	-0.0075	0.0140	
0.0417	0.0393	0.0012	-0.0036
0.0135	0.0083	-0.0230	
-0.0102	0.0221	-0.0104	-0.0601
-0.0011	-0.0370	-0.0254	

```

ans (:,:,1,5) =

```

```

ans (:,:,34) =

```

4.3792e-10

```

ans (:,:,35) =

```

1.0296e-09

```

ans (:,:,36) =

```

-3.9954e-10

```

ans (:,:,37) =

```

2.3627e-10

```

ans (:,:,38) =

```

1.7206e-10

```

ans (:,:,39) =

```

4.0136e-10

```

ans (:,:,40) =

```

6.8313e-10

```

ans (:,:,41) =

```

6.5211e-10

```

ans (:,:,42) =

```

1.4209e-10

```

ans (:,:,43) =

```

-2.3432e-10

```

ans (:,:,44) =

```

4.2423e-10

0.0423 0.0038 0.0080 0.0548	ans (:,:,45) =
-0.0046 0.0559 -0.0222	5.9869e-10
0.0379 -0.0015 0.0259 0.0297	
0.0489 0.0224 0.0163	
-0.0144 0.0600 -0.0005 -0.0238	
-0.0286 0.0198 0.0292	ans (:,:,46) =
-0.0261 0.0327 0.0492 0.0272	1.2344e-10
-0.0269 0.0043 -0.0345	
0.0302 -0.0179 0.0300 -0.0024	
0.0116 0.0440 0.0244	ans (:,:,47) =
0.0201 0.0166 0.0059 -0.0031	4.0838e-10
0.0059 0.0115 -0.0340	
-0.0066 -0.0121 -0.0221 -0.0068	
0.0437 -0.0291 -0.0316	
ans (:,:,2,5) =	ans (:,:,48) =
0.0565 0.0011 0.0487 0.0259	-8.3573e-10
0.0499 0.0163 0.0132	
0.0024 0.0218 0.0044 0.0746	ans (:,:,49) =
0.0025 0.0492 -0.0157	-1.0337e-09
0.0120 -0.0072 0.0525 0.0309	
-0.0043 0.0357 0.0377	ans (:,:,50) =
0.0009 0.0408 -0.0001 -0.0250	-8.3291e-11
0.0270 -0.0002 0.0155	
0.0003 0.0118 0.0234 -0.0061	ans (:,:,51) =
0.0239 0.0156 -0.0211	-1.4597e-10
-0.0170 0.0242 0.0217 0.0302	
0.0031 0.0134 0.0011	
0.0287 -0.0127 0.0296 0.0162	
0.0200 -0.0261 -0.0242	
ans (:,:,3,5) =	ans (:,:,52) =
0.0140 0.0115 0.0488 0.0765	6.1066e-10
0.0450 0.0126 0.0427	
0.0494 0.0161 0.0190 0.0150	ans (:,:,53) =
0.0133 0.0194 0.0305	1.7308e-10
0.0170 0.0617 0.0680 0.0046	
0.0125 -0.0236 0.0044	ans (:,:,54) =
-0.0059 0.0242 -0.0201 0.0402	5.4837e-10
0.0288 0.0276 -0.0081	
-0.0200 0.0293 -0.0206 0.0302	ans (:,:,55) =
-0.0032 0.0132 -0.0195	3.9214e-10
0.0142 -0.0241 0.0439 -0.0344	
-0.0235 -0.0350 0.0226	ans (:,:,56) =
-0.0322 -0.0271 0.0191 0.0365	
0.0436 -0.0287 -0.0331	
ans (:,:,1,6) =	
-0.0130 -0.0153 0.0271 -0.0417	
-0.0177 0.0159 0.0135	

0.0192 -0.0322 0.0005 -0.0200	8.7837e-10		
0.0096 0.0212 -0.0170		ans(:,:,57) =	
-0.0386 -0.0202 -0.0371 0.0003		2.2101e-10	
-0.0249 -0.0147 0.0257			
-0.0217 0.0220 -0.0238 0.0161			
-0.0467 0.0367 0.0339			
0.0274 -0.0077 0.0014 -0.0351			
0.0306 0.0180 -0.0033		ans(:,:,58) =	
-0.0419 -0.0080 -0.0378 -0.0451		6.1246e-10	
-0.0322 -0.0436 0.0196			
-0.0387 0.0144 -0.0336 -0.0334			
-0.0003 -0.0393 0.0070			
		ans(:,:,59) =	
ans(:,:,2,6) =		-3.3195e-10	
-0.0452 0.0283 -0.0246 -0.0411			
-0.0123 0.0387 0.0182		ans(:,:,60) =	
-0.0421 0.0077 -0.0419 0.0258		-6.3985e-10	
0.0023 0.0026 0.0346			
-0.0348 -0.0274 0.0079 -0.0003			
-0.0365 -0.0136 0.0426			
0.0065 -0.0219 -0.0398 -0.0387			
-0.0393 -0.0401 0.0192		ans(:,:,61) =	
0.0164 -0.0475 -0.0351 -0.0490		5.3441e-10	
-0.0012 -0.0312 0.0194			
-0.0108 0.0036 -0.0311 -0.0189			
0.0199 0.0084 -0.0375		ans(:,:,62) =	
0.0109 -0.0423 0.0064 -0.0362		8.7308e-10	
0.0235 -0.0212 -0.0231			
		ans(:,:,63) =	
ans(:,:,3,6) =		4.5698e-10	
0.0114 -0.0138 0.0036 -0.0327			
-0.0280 0.0353 0.0335			
-0.0298 0.0216 -0.0027 0.0034			
-0.0136 0.0056 0.0238		ans(:,:,64) =	
-0.0420 -0.0090 -0.0079 -0.0377		3.7422e-10	
-0.0441 0.0248 -0.0045			
-0.0369 -0.0141 -0.0074 0.0129			
-0.0178 0.0322 0.0409			
0.0017 -0.0180 -0.0089 0.0300			
0.0026 -0.0181 -0.0269			
-0.0167 -0.0032 -0.0276 0.0110			
-0.0433 -0.0382 0.0072			
-0.0246 -0.0107 -0.0100 -0.0351			
-0.0148 0.0080 0.0309			
		ans(:,:,1,7) =	
-0.0398 -0.0175 0.0086 0.0008			
0.0397 0.0367 0.0551			
-0.0288 0.0202 -0.0264 -0.0126			
-0.0081 0.0245 0.0476			

-0.0085	-0.0008	0.0209	0.0123
0.0476	0.0281	-0.0142	
0.0604	0.0489	0.0003	0.0112
-0.0212	0.0314	0.0098	
0.0298	0.0072	-0.0153	0.0164
0.0215	0.0574	0.0467	
-0.0239	-0.0040	-0.0124	0.0157
-0.0112	-0.0051	0.0600	
0.0268	-0.0290	0.0092	-0.0103
0.0397	0.0378	-0.0071	
ans (:,:,2,7) =			
0.0032	-0.0155	-0.0379	0.0417
0.0462	0.0101	0.0143	
-0.0360	-0.0398	0.0345	-0.0199
-0.0189	-0.0198	0.0650	
0.0318	-0.0085	-0.0233	0.0082
-0.0135	0.0609	0.0296	
0.0228	0.0216	0.0190	0.0436
0.0044	0.0277	-0.0186	
0.0389	0.0286	-0.0229	-0.0063
0.0088	0.0454	0.0609	
0.0492	0.0214	-0.0295	0.0021
0.0398	0.0193	0.0162	
-0.0155	-0.0116	0.0421	0.0238
0.0277	-0.0109	0.0313	
ans (:,:,3,7) =			
0.0048	0.0252	-0.0157	-0.0287
0.0417	-0.0136	-0.0115	
-0.0415	-0.0425	0.0355	-0.0317
-0.0272	0.0452	0.0113	
-0.0226	0.0363	-0.0365	-0.0414
-0.0072	0.0096	0.0068	
-0.0201	0.0267	0.0397	0.0336
-0.0181	-0.0058	-0.0163	
0.0602	0.0522	-0.0059	0.0366
0.0495	-0.0144	0.0320	
0.0349	-0.0195	0.0001	0.0122
0.0365	0.0050	0.0530	
-0.0167	-0.0216	0.0256	0.0100
0.0486	0.0441	-0.0114	
ans (:,:,1,8) =			
0.0006	0.0074	0.0481	0.0151
-0.0236	0.0357	-0.0294	
-0.0166	0.0487	-0.0318	0.0185
-0.0181	0.0223	-0.0112	
0.0466	0.0294	0.0297	0.0273
-0.0062	0.0205	0.0165	

0.0523 0.0596 -0.0316 0.0055
0.0334 0.0062 -0.0157
0.0060 0.0283 -0.0034 0.0294
-0.0429 -0.0109 -0.0357
0.0637 0.0456 0.0148 0.0222
0.0288 0.0322 0.0119
0.0503 0.0491 0.0176 -0.0320
-0.0353 -0.0465 0.0020
ans (:,:,2,8) =
0.0203 0.0206 0.0196 -0.0264
0.0194 0.0313 0.0228
0.0444 0.0477 0.0072 0.0075
-0.0022 0.0154 -0.0374
0.0577 0.0604 -0.0232 0.0393
-0.0308 -0.0380 -0.0048
0.0710 0.0576 -0.0131 -0.0099
0.0132 -0.0469 -0.0275
0.0045 0.0105 -0.0231 0.0075
0.0161 -0.0436 -0.0099
0.0475 0.0101 0.0321 -0.0271
0.0181 0.0114 -0.0143
0.0331 -0.0176 -0.0111 -0.0202
-0.0181 0.0136 -0.0280
ans (:,:,3,8) =
0.0236 0.0213 0.0228 -0.0200
-0.0337 0.0414 0.0071
0.0686 -0.0157 0.0475 0.0315
0.0242 -0.0102 0.0201
0.0377 0.0075 -0.0030 -0.0073
0.0009 -0.0583 -0.0028
0.0714 0.0471 0.0066 0.0423
0.0222 0.0302 -0.0206
0.0206 0.0153 -0.0292 0.0100
-0.0195 -0.0350 -0.0263
0.0346 0.0110 -0.0201 -0.0425
-0.0062 -0.0469 -0.0513
0.0715 0.0490 -0.0212 -0.0283
0.0334 -0.0035 -0.0455
ans (:,:,1,9) =
-0.0246 0.0219 0.0108 0.0310
0.0457 0.0044 -0.0245
0.0397 0.0294 -0.0012 0.0478
-0.0030 0.0213 0.0020
0.0401 0.0196 -0.0131 0.0013
-0.0202 -0.0194 0.0462
0.0332 0.0123 -0.0233 0.0300
0.0175 0.0536 0.0536

0.0024	0.0485	0.0194	-0.0367
0.0299	0.0244	0.0575	
0.0217	-0.0210	0.0018	-0.0117
0.0175	-0.0212	0.0285	
0.0234	-0.0291	0.0222	0.0248
-0.0118	0.0525	0.0153	
 ans (:,:,2,9) =			
0.0287	0.0578	0.0168	-0.0111
0.0187	0.0584	0.0177	
-0.0153	0.0294	0.0009	0.0123
0.0000	0.0087	0.0408	
0.0188	0.0463	0.0269	-0.0128
0.0231	0.0487	0.0440	
0.0069	0.0210	0.0354	-0.0109
-0.0024	0.0309	-0.0033	
-0.0082	-0.0345	-0.0424	0.0317
0.0342	-0.0195	-0.0132	
-0.0005	0.0102	0.0236	0.0096
-0.0112	0.0475	0.0065	
-0.0017	-0.0347	-0.0388	-0.0089
-0.0077	0.0560	0.0159	
 ans (:,:,3,9) =			
-0.0115	0.0113	0.0357	-0.0004
0.0192	0.0262	0.0237	
0.0109	0.0446	0.0112	0.0070
0.0266	-0.0218	-0.0184	
0.0089	0.0398	-0.0350	0.0344
-0.0174	-0.0141	-0.0027	
0.0285	0.0279	-0.0057	0.0273
-0.0008	-0.0109	0.0307	
0.0068	-0.0066	0.0222	-0.0489
-0.0068	0.0140	0.0199	
0.0191	-0.0123	-0.0168	-0.0455
-0.0101	-0.0236	-0.0236	
0.0005	0.0111	-0.0028	0.0299
-0.0008	0.0463	0.0161	
 ans (:,:,1,10) =			
-0.0495	-0.0200	0.0183	-0.0221
-0.0001	-0.0276	-0.0043	
-0.0012	0.0213	0.0178	-0.0171
-0.0300	-0.0195	0.0251	
-0.0145	-0.0227	0.0020	-0.0011
-0.0261	0.0149	-0.0530	
-0.0091	-0.0026	0.0231	-0.0052
0.0143	-0.0109	-0.0422	
-0.0205	-0.0384	-0.0042	0.0098
-0.0297	0.0181	-0.0042	

<pre> 0.0155 -0.0351 -0.0532 -0.0283 -0.0365 0.0251 0.0010 0.0204 -0.0130 -0.0124 -0.0181 -0.0463 -0.0664 -0.0027 </pre>	
<pre> ans (:,:,2,10) = </pre>	
<pre> 0.0207 -0.0162 0.0133 -0.0267 0.0149 0.0425 -0.0232 0.0297 0.0074 -0.0143 -0.0187 0.0233 -0.0329 -0.0128 -0.0395 0.0067 -0.0273 0.0038 -0.0378 0.0182 0.0113 0.0117 -0.0542 -0.0305 -0.0245 0.0253 -0.0304 -0.0196 -0.0116 0.0298 0.0126 -0.0371 0.0104 -0.0375 -0.0484 0.0158 0.0255 -0.0458 -0.0269 -0.0275 -0.0421 0.0139 -0.0296 -0.0255 -0.0365 -0.0607 -0.0426 -0.0278 -0.0387 </pre>	
<pre> ans (:,:,3,10) = </pre>	
<pre> 0.0094 0.0266 -0.0348 -0.0083 0.0165 0.0070 -0.0180 -0.0442 0.0301 -0.0514 0.0277 0.0422 0.0237 -0.0318 0.0265 -0.0569 0.0093 -0.0026 0.0121 0.0148 0.0076 -0.0456 0.0126 -0.0081 0.0284 0.0096 -0.0135 0.0228 -0.0499 -0.0382 -0.0138 -0.0369 -0.0191 0.0011 -0.0140 -0.0075 0.0284 -0.0137 -0.0190 -0.0437 0.0247 0.0246 0.0196 -0.0091 0.0027 -0.0370 0.0028 0.0114 0.0113 </pre>	
<pre> ans (:,:,1,11) = </pre>	
<pre> 0.0079 -0.0066 0.0554 0.0145 0.0378 0.0049 -0.0517 -0.0172 0.0352 0.0161 0.0416 0.0318 -0.0501 0.0216 -0.0048 -0.0031 0.0541 0.0214 0.0200 0.0238 -0.0134 0.0153 0.0432 -0.0059 0.0291 0.0266 -0.0174 -0.0172 0.0147 -0.0360 0.0293 0.0294 -0.0251 -0.0081 0.0201 -0.0111 -0.0216 0.0374 0.0141 0.0242 0.0049 0.0030 </pre>	

$\begin{bmatrix} -0.0091 & 0.0303 & 0.0544 & -0.0052 \\ -0.0016 & -0.0186 & 0.0279 \end{bmatrix}$	
$\text{ans}(:,:,2,11) =$ $\begin{bmatrix} 0.0444 & 0.0514 & 0.0227 & 0.0485 \\ -0.0144 & -0.0299 & 0.0262 \\ 0.0234 & 0.0062 & 0.0593 & -0.0104 \\ -0.0072 & -0.0080 & -0.0279 \\ -0.0034 & 0.0120 & 0.0062 & -0.0307 \\ 0.0144 & -0.0203 & -0.0151 \\ -0.0167 & 0.0041 & -0.0096 & 0.0064 \\ -0.0142 & 0.0123 & 0.0176 \\ -0.0325 & -0.0108 & 0.0237 & 0.0442 \\ -0.0255 & -0.0178 & 0.0184 \\ 0.0090 & -0.0002 & 0.0148 & -0.0046 \\ 0.0041 & -0.0150 & -0.0307 \\ 0.0060 & -0.0138 & 0.0066 & 0.0283 \\ -0.0328 & 0.0121 & -0.0180 \end{bmatrix}$	
$\text{ans}(:,:,3,11) =$ $\begin{bmatrix} 0.0447 & 0.0540 & 0.0036 & 0.0314 \\ 0.0367 & 0.0057 & -0.0059 \\ -0.0051 & 0.0149 & 0.0372 & 0.0092 \\ -0.0271 & 0.0254 & -0.0274 \\ 0.0217 & 0.0365 & 0.0103 & -0.0225 \\ 0.0261 & 0.0098 & -0.0062 \\ 0.0368 & -0.0156 & 0.0439 & 0.0187 \\ 0.0277 & -0.0220 & -0.0220 \\ 0.0480 & 0.0205 & 0.0357 & 0.0324 \\ -0.0140 & -0.0343 & -0.0193 \\ 0.0448 & -0.0149 & 0.0123 & 0.0495 \\ -0.0068 & -0.0417 & -0.0315 \\ 0.0075 & 0.0392 & 0.0234 & 0.0314 \\ 0.0238 & 0.0355 & -0.0273 \end{bmatrix}$	
$\text{ans}(:,:,1,12) =$ $\begin{bmatrix} 0.0104 & 0.0320 & -0.0410 & 0.0231 \\ 0.0136 & -0.0407 & 0.0286 \\ 0.0363 & -0.0001 & 0.0277 & -0.0068 \\ 0.0018 & -0.0460 & 0.0174 \\ -0.0002 & -0.0054 & -0.0291 & 0.0030 \\ -0.0050 & 0.0170 & 0.0232 \\ 0.0416 & -0.0131 & -0.0312 & 0.0180 \\ -0.0258 & 0.0162 & -0.0047 \\ 0.0417 & 0.0117 & -0.0015 & -0.0149 \\ 0.0044 & 0.0213 & -0.0198 \\ 0.0137 & 0.0069 & -0.0078 & 0.0406 \\ 0.0388 & 0.0168 & 0.0178 \\ -0.0226 & 0.0243 & 0.0106 & 0.0327 \\ 0.0293 & -0.0298 & 0.0195 \end{bmatrix}$	

<pre>ans (:,:,2,12) =</pre> $\begin{matrix} 0.0190 & 0.0360 & -0.0049 & 0.0166 \\ -0.0104 & -0.0406 & 0.0105 & \\ -0.0318 & 0.0191 & 0.0382 & 0.0332 \\ 0.0100 & 0.0014 & 0.0283 & \\ -0.0093 & -0.0012 & 0.0077 & 0.0471 \\ -0.0015 & 0.0355 & -0.0216 & \\ 0.0009 & 0.0218 & -0.0023 & 0.0425 \\ 0.0198 & -0.0240 & 0.0330 & \\ 0.0130 & 0.0223 & 0.0491 & 0.0422 \\ 0.0163 & -0.0129 & 0.0134 & \\ 0.0243 & -0.0047 & 0.0627 & -0.0031 \\ 0.0467 & -0.0031 & 0.0389 & \\ 0.0105 & -0.0165 & -0.0096 & 0.0440 \\ 0.0531 & -0.0001 & 0.0007 & \end{matrix}$	
<pre>ans (:,:,3,12) =</pre> $\begin{matrix} 0.0192 & 0.0128 & -0.0277 & 0.0281 \\ -0.0098 & 0.0237 & 0.0056 & \\ -0.0150 & 0.0217 & -0.0106 & -0.0180 \\ -0.0353 & -0.0442 & -0.0288 & \\ 0.0001 & 0.0209 & 0.0341 & 0.0320 \\ -0.0052 & 0.0187 & 0.0409 & \\ 0.0526 & 0.0429 & 0.0233 & 0.0325 \\ 0.0089 & 0.0110 & -0.0032 & \\ 0.0399 & 0.0014 & 0.0437 & -0.0117 \\ 0.0482 & -0.0143 & 0.0050 & \\ -0.0121 & 0.0608 & 0.0473 & 0.0242 \\ 0.0372 & -0.0240 & -0.0234 & \\ -0.0160 & 0.0177 & 0.0245 & 0.0184 \\ 0.0371 & -0.0222 & 0.0471 & \end{matrix}$	
<pre>ans (:,:,1,13) =</pre> $\begin{matrix} 0.0230 & -0.0096 & -0.0346 & 0.0263 \\ 0.0069 & 0.0244 & -0.0387 & \\ -0.0400 & 0.0065 & 0.0093 & 0.0528 \\ -0.0054 & 0.0436 & -0.0007 & \\ -0.0122 & 0.0352 & 0.0103 & 0.0349 \\ 0.0041 & -0.0160 & 0.0032 & \\ -0.0041 & -0.0512 & 0.0042 & 0.0124 \\ 0.0105 & 0.0228 & -0.0340 & \\ 0.0219 & -0.0042 & -0.0637 & 0.0021 \\ 0.0033 & -0.0395 & 0.0226 & \\ -0.0403 & 0.0115 & -0.0236 & -0.0819 \\ -0.0682 & 0.0296 & -0.0336 & \\ -0.0675 & 0.0077 & -0.0437 & 0.0024 \\ 0.0214 & 0.0442 & 0.0513 & \end{matrix}$	
<pre>ans (:,:,2,13) =</pre>	

-0.0070	0.0070	-0.0439	0.0007
0.0078	-0.0080	-0.0149	
-0.0039	0.0358	0.0131	0.0117
0.0595	0.0247	-0.0082	
-0.0094	-0.0174	-0.0373	-0.0168
0.0116	0.0424	0.0332	
-0.0416	0.0211	0.0000	-0.0354
-0.0487	0.0186	0.0135	
-0.0582	0.0206	-0.0328	-0.0171
-0.0341	0.0065	-0.0082	
0.0175	-0.0275	-0.0647	-0.0642
-0.0294	-0.0335	0.0230	
-0.0540	-0.0723	-0.0603	-0.0241
-0.0255	0.0312	0.0277	
ans (:,:,3,13) =			
0.0117	-0.0156	0.0076	-0.0243
0.0257	0.0042	-0.0138	
0.0132	0.0083	0.0122	0.0129
0.0066	0.0032	0.0197	
-0.0351	-0.0407	0.0139	-0.0117
0.0443	0.0397	-0.0151	
-0.0477	-0.0012	-0.0440	-0.0330
0.0127	-0.0462	0.0164	
-0.0245	0.0150	-0.0268	-0.0265
-0.0050	-0.0127	0.0382	
-0.0457	-0.0480	-0.0103	-0.0507
-0.0096	0.0109	0.0127	
-0.0045	-0.0671	-0.0005	-0.0067
-0.0389	0.0289	0.0456	
ans (:,:,1,14) =			
-0.0691	-0.0480	-0.0541	-0.0170
0.0326	0.0199	0.0048	
-0.0174	-0.0298	0.0044	0.0282
0.0299	0.0586	0.0555	
-0.0398	-0.0161	-0.0145	0.0112
0.0480	0.0146	0.0463	
0.0110	-0.0226	-0.0416	-0.0157
0.0211	-0.0168	0.0067	
-0.0117	0.0337	-0.0286	0.0262
0.0383	0.0413	-0.0050	
-0.0224	0.0207	0.0325	0.0402
0.0286	0.0233	0.0406	
-0.0582	0.0095	0.0248	0.0401
0.0158	-0.0177	0.0062	
ans (:,:,2,14) =			
-0.0467	-0.0464	0.0030	0.0320
0.0493	0.0679	0.0603	

-0.0620	-0.0193	-0.0092	-0.0260
0.0362	-0.0020	0.0295	
	-0.0299	-0.0191	-0.0377
0.0122	-0.0127	-0.0145	-0.0448
	-0.0271	0.0280	0.0227
-0.0387	0.0260	0.0328	-0.0086
	-0.0284	0.0183	-0.0161
0.0340	0.0117	0.0547	0.0314
	-0.0415	0.0224	0.0367
0.0146	0.0467	0.0090	-0.0003
	-0.0010	-0.0284	-0.0134
0.0059	0.0651	0.0129	0.0202

ans (:,:,3,14) =

-0.0174	0.0161	0.0082	-0.0249
0.0038	0.0657	0.0173	
	0.0087	-0.0001	0.0111
0.0349	0.0414	0.0357	-0.0387
	-0.0031	-0.0202	0.0240
-0.0248	0.0118	0.0638	-0.0120
	-0.0122	-0.0295	-0.0488
0.0280	0.0249	0.0519	0.0274
	-0.0020	0.0158	0.0134
0.0022	0.0152	0.0516	-0.0055
	-0.0442	-0.0051	0.0216
-0.0008	0.0010	0.0029	-0.0017
	-0.0544	-0.0417	-0.0103
0.0511	0.0657	0.0443	-0.0236

ans (:,:,1,15) =

-0.0014	0.0245	0.0272	0.0054
0.0225	0.0338	0.0091	
	0.0217	-0.0032	0.0026
0.0220	0.0186	0.0304	0.0416
	-0.0168	-0.0484	0.0210
0.0351	-0.0037	0.0275	0.0144
	0.0083	0.0029	0.0413
0.0280	0.0241	-0.0117	0.0287
	-0.0456	-0.0451	0.0127
-0.0353	-0.0206	0.0224	0.0299
	-0.0083	-0.0203	0.0139
-0.0010	-0.0237	-0.0178	0.0103
	-0.0435	-0.0190	-0.0081
-0.0199	0.0263	0.0581	-0.0102

ans (:,:,2,15) =

-0.0407	-0.0354	-0.0345	-0.0033
0.0366	-0.0200	0.0419	
	0.0046	-0.0276	0.0233
0.0428	0.0304	0.0555	0.0008

-0.0057 -0.0129 -0.0385 -0.0377
-0.0083 0.0424 0.0284
-0.0464 0.0317 -0.0423 -0.0097
-0.0302 -0.0126 0.0312
-0.0359 0.0050 -0.0369 0.0184
-0.0412 -0.0115 0.0582
0.0008 0.0135 0.0092 -0.0382
-0.0039 0.0411 0.0044
-0.0496 -0.0125 -0.0588 0.0074
0.0168 0.0207 0.0011
ans (:,:,3,15) =
-0.0388 -0.0420 -0.0095 0.0431
-0.0084 0.0317 0.0185
-0.0480 -0.0028 0.0318 0.0284
-0.0353 0.0445 0.0351
-0.0200 -0.0242 -0.0176 0.0216
0.0132 0.0117 0.0580
0.0118 0.0102 -0.0313 -0.0160
0.0202 0.0378 0.0451
-0.0225 -0.0193 0.0328 -0.0033
0.0076 -0.0055 0.0655
-0.0260 0.0076 -0.0596 -0.0332
-0.0458 -0.0306 0.0303
-0.0203 -0.0386 -0.0341 0.0099
-0.0148 -0.0203 0.0407
ans (:,:,1,16) =
0.0105 -0.0556 0.0148 -0.0043
-0.0108 0.0149 -0.0129
0.0143 -0.0090 0.0242 0.0067
0.0402 0.0444 0.0315
-0.0284 -0.0370 0.0271 0.0388
-0.0071 -0.0008 0.0028
-0.0355 0.0178 0.0232 0.0206
-0.0222 -0.0323 0.0150
-0.0575 -0.0089 -0.0358 0.0125
0.0032 -0.0170 0.0221
-0.0167 -0.0146 -0.0552 0.0208
0.0374 0.0095 0.0012
-0.0826 -0.0840 -0.0061 -0.0102
0.0329 0.0195 0.0346
ans (:,:,2,16) =
0.0193 -0.0336 -0.0595 -0.0313
0.0142 0.0552 0.0346
0.0261 -0.0126 0.0066 -0.0245
-0.0188 0.0316 0.0480
-0.0275 -0.0102 0.0235 -0.0240
0.0141 0.0218 -0.0138

-0.0572 -0.0220 -0.0037 -0.0272
0.0463 0.0018 -0.0533
-0.0174 -0.0168 -0.0080 0.0253
0.0367 0.0192 -0.0586
-0.0541 -0.0308 0.0080 0.0204
0.0327 -0.0306 0.0262
-0.0573 -0.0769 -0.0232 -0.0088
0.0250 0.0341 0.0005
ans (:,:,3,16) =
-0.0232 -0.0146 -0.0338 -0.0480
-0.0113 0.0358 0.0240
-0.0277 0.0197 -0.0276 -0.0177
0.0547 0.0014 -0.0142
-0.0339 -0.0307 -0.0047 0.0496
0.0229 0.0316 0.0371
-0.0158 0.0061 0.0070 0.0086
0.0278 -0.0134 -0.0437
-0.0523 -0.0379 0.0255 -0.0207
0.0378 -0.0333 -0.0135
0.0080 -0.0262 -0.0462 -0.0406
0.0083 0.0428 -0.0177
-0.0715 -0.0641 -0.0058 0.0383
-0.0218 0.0013 0.0051
ans (:,:,1,17) =
-0.0482 -0.0231 0.0177 -0.0258
0.0036 -0.0235 0.0287
-0.0596 0.0121 0.0106 -0.0148
-0.0251 0.0191 0.0284
0.0068 -0.0173 0.0162 0.0244
-0.0238 0.0081 -0.0212
-0.0418 -0.0014 -0.0284 0.0253
-0.0327 -0.0438 -0.0111
0.0048 0.0013 -0.0126 -0.0300
0.0257 0.0082 0.0102
-0.0309 -0.0433 -0.0450 -0.0198
0.0038 -0.0548 -0.0435
-0.0290 0.0038 0.0202 -0.0249
0.0036 -0.0104 -0.0249
ans (:,:,2,17) =
0.0026 -0.0659 0.0090 -0.0293
0.0216 0.0481 -0.0324
0.0054 -0.0022 0.0216 -0.0251
-0.0309 0.0185 -0.0094
-0.0315 -0.0077 -0.0018 0.0363
-0.0031 0.0280 0.0152
0.0136 0.0238 0.0172 0.0137
-0.0441 -0.0014 -0.0054

<pre> -0.0199 0.0222 -0.0301 0.0030 -0.0520 -0.0497 -0.0645 -0.0257 -0.0331 0.0184 -0.0362 -0.0437 -0.0546 0.0085 -0.0093 -0.0240 0.0258 -0.0510 -0.0230 -0.0183 -0.0089 </pre>
<pre> ans (:,:,3,17) = </pre>
<pre> -0.0208 -0.0527 -0.0391 0.0161 0.0085 0.0037 -0.0126 0.0073 0.0089 0.0110 0.0444 0.0517 -0.0137 0.0167 -0.0558 -0.0019 0.0130 0.0123 0.0293 -0.0283 -0.0285 -0.0356 0.0226 0.0044 -0.0431 0.0258 0.0125 -0.0565 -0.0455 0.0242 -0.0052 -0.0148 0.0242 -0.0521 -0.0623 0.0241 -0.0048 -0.0497 -0.0512 -0.0175 -0.0435 0.0083 0.0267 -0.0150 0.0061 -0.0092 -0.0361 -0.0064 -0.0691 </pre>
<pre> ans (:,:,1,18) = </pre>
<pre> 0.0433 -0.0197 -0.0461 0.0072 -0.0383 -0.0122 -0.0518 -0.0225 -0.0207 -0.0103 0.0133 -0.0306 -0.0610 -0.0645 -0.0208 0.0354 0.0382 -0.0359 -0.0040 -0.0345 -0.0576 0.0461 0.0034 0.0175 0.0301 -0.0435 0.0023 0.0052 0.0290 -0.0376 -0.0204 0.0103 -0.0281 0.0007 0.0033 -0.0425 -0.0450 -0.0047 -0.0163 0.0352 -0.0348 -0.0215 -0.0296 0.0004 0.0108 0.0150 0.0276 -0.0090 -0.0113 </pre>
<pre> ans (:,:,2,18) = </pre>
<pre> -0.0106 0.0193 -0.0267 -0.0255 0.0165 0.0060 -0.0300 0.0239 0.0058 0.0091 -0.0041 -0.0249 -0.0007 -0.0540 0.0438 0.0401 -0.0314 0.0224 -0.0364 -0.0131 0.0093 0.0014 -0.0293 -0.0047 0.0342 -0.0057 0.0220 -0.0406 0.0271 -0.0085 -0.0117 -0.0068 -0.0300 -0.0359 -0.0469 </pre>

<pre> -0.0208 -0.0303 0.0010 -0.0399 -0.0491 -0.0108 -0.0120 -0.0045 -0.0169 -0.0122 -0.0097 -0.0206 -0.0238 0.0213 </pre>	
<pre> ans (:,:,3,18) = </pre>	
<pre> 0.0440 -0.0252 -0.0439 -0.0269 0.0000 0.0031 -0.0273 0.0356 -0.0061 0.0170 -0.0542 0.0085 -0.0303 -0.0587 -0.0225 0.0338 -0.0147 -0.0264 0.0118 0.0196 0.0212 -0.0361 -0.0026 -0.0250 -0.0381 -0.0196 -0.0009 0.0103 -0.0314 -0.0382 -0.0424 -0.0238 -0.0175 -0.0354 0.0209 -0.0247 -0.0152 -0.0079 0.0272 -0.0451 -0.0333 0.0186 -0.0202 -0.0016 -0.0240 0.0243 -0.0251 0.0100 -0.0351 </pre>	
<pre> ans (:,:,1,19) = </pre>	
<pre> -0.0316 -0.0346 -0.0056 0.0325 0.0301 0.0062 0.0659 -0.0169 -0.0130 0.0171 -0.0148 0.0304 0.0420 0.0076 0.0281 0.0115 0.0517 0.0317 0.0574 0.0507 0.0507 0.0310 0.0087 0.0418 0.0336 -0.0155 -0.0236 0.0520 -0.0387 -0.0389 -0.0021 0.0227 -0.0123 0.0293 -0.0080 -0.0172 -0.0107 -0.0101 0.0174 0.0489 -0.0264 0.0087 -0.0086 0.0032 0.0111 0.0383 -0.0126 0.0093 -0.0317 </pre>	
<pre> ans (:,:,2,19) = </pre>	
<pre> 0.0072 0.0343 0.0064 0.0304 0.0283 -0.0045 0.0132 0.0384 0.0114 0.0140 0.0228 -0.0032 0.0563 0.0421 -0.0060 0.0375 0.0229 0.0208 -0.0066 0.0613 0.0374 0.0149 0.0256 -0.0050 0.0067 0.0511 0.0090 0.0387 -0.0407 -0.0346 -0.0236 0.0324 -0.0012 0.0181 -0.0216 -0.0284 0.0060 -0.0424 0.0217 0.0182 -0.0290 0.0295 </pre>	

<pre> 0.0078 0.0026 0.0264 0.0412 0.0248 0.0146 0.0077 </pre>	
<pre> ans (:,:,3,19) = </pre>	
<pre> 0.0178 0.0040 -0.0001 0.0293 -0.0152 0.0008 0.0305 0.0078 0.0192 0.0035 0.0546 -0.0158 -0.0001 -0.0113 -0.0308 0.0102 0.0491 -0.0157 -0.0033 0.0524 -0.0110 -0.0423 -0.0205 -0.0169 0.0058 -0.0045 0.0496 -0.0104 -0.0205 0.0033 -0.0007 0.0386 0.0239 0.0098 0.0232 -0.0138 0.0328 0.0033 0.0400 -0.0250 0.0092 0.0264 -0.0391 -0.0137 0.0102 0.0006 0.0161 0.0225 -0.0118 </pre>	
<pre> ans (:,:,1,20) = </pre>	
<pre> 0.0191 -0.0001 0.0483 0.0132 -0.0006 0.0074 0.0022 -0.0163 0.0305 0.0137 -0.0230 0.0384 0.0205 -0.0183 0.0098 0.0360 -0.0344 -0.0110 -0.0331 0.0183 0.0495 -0.0092 -0.0384 -0.0352 0.0074 0.0242 -0.0039 0.0272 -0.0279 -0.0039 0.0185 0.0022 0.0156 0.0053 -0.0142 0.0411 0.0078 -0.0202 0.0481 0.0066 0.0198 0.0620 -0.0029 -0.0160 0.0497 0.0466 0.0257 0.0335 0.0101 </pre>	
<pre> ans (:,:,2,20) = </pre>	
<pre> -0.0384 -0.0332 0.0229 0.0295 -0.0118 0.0190 0.0161 0.0228 -0.0187 0.0116 0.0151 -0.0314 0.0353 -0.0278 0.0322 -0.0275 0.0312 -0.0050 0.0211 0.0078 0.0489 0.0057 0.0199 0.0101 -0.0326 -0.0335 0.0480 0.0201 -0.0275 -0.0384 0.0502 -0.0142 0.0280 0.0107 0.0390 0.0158 0.0059 0.0512 0.0603 -0.0105 0.0234 0.0069 0.0131 0.0239 0.0050 -0.0180 0.0100 -0.0027 0.0402 </pre>	

<pre>ans (:,:,3,20) =</pre> $\begin{matrix} -0.0077 & 0.0367 & -0.0198 & 0.0212 \\ -0.0059 & 0.0144 & -0.0063 & \\ 0.0204 & 0.0047 & 0.0050 & 0.0405 \\ -0.0183 & 0.0333 & 0.0113 & \\ 0.0224 & 0.0060 & 0.0308 & -0.0208 \\ 0.0012 & -0.0142 & 0.0126 & \\ 0.0290 & -0.0023 & -0.0175 & 0.0470 \\ -0.0143 & 0.0164 & -0.0048 & \\ -0.0360 & -0.0268 & 0.0404 & 0.0179 \\ 0.0228 & 0.0240 & 0.0132 & \\ -0.0083 & -0.0142 & 0.0360 & 0.0475 \\ 0.0092 & 0.0533 & 0.0511 & \\ -0.0004 & 0.0032 & 0.0016 & 0.0001 \\ 0.0057 & 0.0466 & 0.0027 & \end{matrix}$	
<pre>ans (:,:,1,21) =</pre> $\begin{matrix} -0.0227 & 0.0232 & -0.0339 & -0.0186 \\ 0.0242 & -0.0314 & 0.0096 & \\ -0.0123 & 0.0098 & -0.0214 & -0.0206 \\ -0.0139 & -0.0087 & -0.0310 & \\ -0.0337 & -0.0310 & 0.0059 & 0.0223 \\ -0.0261 & 0.0170 & 0.0264 & \\ 0.0149 & 0.0005 & -0.0353 & 0.0071 \\ -0.0072 & -0.0219 & 0.0145 & \\ -0.0343 & -0.0135 & 0.0042 & -0.0400 \\ 0.0349 & 0.0187 & 0.0171 & \\ -0.0428 & -0.0146 & -0.0445 & 0.0267 \\ 0.0073 & -0.0346 & 0.0047 & \\ 0.0052 & -0.0465 & -0.0482 & 0.0287 \\ -0.0366 & -0.0267 & -0.0340 & \end{matrix}$	
<pre>ans (:,:,2,21) =</pre> $\begin{matrix} -0.0331 & -0.0033 & 0.0280 & 0.0322 \\ 0.0092 & -0.0494 & -0.0162 & \\ -0.0275 & -0.0071 & 0.0042 & 0.0095 \\ -0.0399 & 0.0098 & -0.0176 & \\ 0.0321 & 0.0406 & -0.0132 & -0.0174 \\ -0.0009 & 0.0204 & -0.0354 & \\ -0.0342 & 0.0105 & 0.0018 & -0.0186 \\ -0.0232 & 0.0167 & -0.0355 & \\ -0.0109 & -0.0294 & 0.0337 & 0.0094 \\ -0.0299 & -0.0169 & 0.0073 & \\ -0.0169 & -0.0453 & 0.0197 & -0.0288 \\ 0.0262 & 0.0042 & -0.0447 & \\ 0.0094 & -0.0047 & -0.0217 & -0.0223 \\ -0.0023 & 0.0241 & -0.0482 & \end{matrix}$	
<pre>ans (:,:,3,21) =</pre>	

0.0144 -0.0192 -0.0335 -0.0457
0.0029 0.0130 -0.0465
-0.0306 -0.0286 -0.0312 -0.0022
-0.0291 -0.0462 -0.0182
0.0330 -0.0281 0.0385 -0.0279
0.0063 0.0102 -0.0306
0.0115 -0.0194 -0.0189 0.0170
-0.0001 0.0335 -0.0272
-0.0213 -0.0290 -0.0323 -0.0247
-0.0163 0.0096 0.0277
-0.0293 0.0201 -0.0110 -0.0064
-0.0435 0.0069 -0.0359
0.0047 -0.0359 -0.0219 -0.0343
0.0293 -0.0167 -0.0234
ans (:,:,1,22) =
-0.0292 -0.0010 -0.0366 -0.0482
0.0184 -0.0039 0.0174
0.0036 0.0070 -0.0389 -0.0360
0.0047 -0.0514 0.0389
-0.0195 -0.0515 -0.0251 0.0077
-0.0363 -0.0415 0.0106
0.0156 -0.0448 -0.0072 -0.0075
-0.0357 -0.0113 0.0352
0.0338 -0.0301 -0.0019 -0.0426
0.0251 -0.0356 -0.0227
-0.0172 0.0227 0.0187 0.0079
0.0173 0.0162 -0.0197
-0.0160 -0.0243 -0.0093 0.0218
0.0001 0.0236 0.0117
ans (:,:,2,22) =
-0.0487 0.0052 -0.0408 -0.0207
-0.0480 0.0099 0.0287
-0.0521 0.0125 -0.0574 -0.0357
-0.0046 -0.0141 -0.0109
-0.0010 0.0010 -0.0208 0.0194
-0.0131 -0.0408 -0.0353
-0.0007 0.0200 -0.0098 0.0297
0.0238 -0.0394 -0.0217
0.0059 -0.0222 -0.0126 0.0299
-0.0380 -0.0390 -0.0359
-0.0196 0.0041 0.0152 -0.0273
-0.0090 -0.0385 -0.0364
-0.0406 -0.0312 -0.0330 -0.0284
0.0121 0.0073 0.0434
ans (:,:,3,22) =
0.0158 -0.0065 -0.0030 0.0225
-0.0477 -0.0287 -0.0410

0.0186	-0.0096	-0.0342	0.0084
-0.0213	-0.0515	-0.0336	
-0.0620	-0.0212	-0.0223	-0.0058
0.0002	-0.0141	0.0005	
0.0110	0.0024	-0.0441	0.0118
0.0104	-0.0399	-0.0220	
-0.0164	0.0173	0.0028	0.0292
0.0305	0.0278	0.0262	
0.0217	-0.0436	-0.0403	-0.0120
0.0183	0.0195	-0.0177	
-0.0446	-0.0346	-0.0477	-0.0182
0.0085	0.0202	0.0286	

ans (:,:,1,23) =

0.0004	0.0125	0.0118	0.0249
0.0108	0.0290	-0.0169	
-0.0244	-0.0030	-0.0103	0.0501
-0.0184	0.0512	0.0029	
0.0391	0.0396	-0.0264	0.0073
-0.0251	-0.0110	0.0267	
-0.0119	-0.0207	0.0194	0.0228
-0.0073	-0.0062	0.0406	
-0.0024	0.0193	0.0559	0.0432
-0.0206	0.0022	0.0404	
0.0373	-0.0051	0.0588	-0.0207
0.0307	-0.0065	0.0268	
0.0468	0.0391	-0.0084	-0.0259
0.0400	-0.0009	0.0227	

ans (:,:,2,23) =

-0.0074	-0.0224	-0.0034	0.0487
0.0054	0.0097	0.0275	
-0.0105	-0.0159	0.0105	0.0186
-0.0226	0.0466	0.0351	
-0.0068	-0.0072	0.0209	0.0121
0.0247	0.0401	0.0432	
0.0112	-0.0154	0.0555	-0.0034
0.0255	-0.0251	-0.0205	
-0.0030	0.0159	-0.0145	0.0047
0.0367	0.0361	0.0464	
-0.0183	0.0265	0.0516	0.0129
-0.0158	-0.0093	0.0387	
0.0183	0.0365	0.0072	0.0081
0.0129	0.0157	-0.0211	

ans (:,:,3,23) =

-0.0035	0.0069	0.0001	-0.0272
0.0443	-0.0090	0.0278	
0.0130	0.0191	-0.0330	0.0054
-0.0195	0.0316	0.0386	

-0.0184	-0.0029	-0.0208	0.0112
-0.0089	0.0427	-0.0293	
0.0190	-0.0041	0.0038	0.0215
-0.0198	0.0274	-0.0349	
0.0360	-0.0048	0.0308	0.0320
0.0500	0.0270	0.0430	
0.0561	0.0074	0.0606	0.0133
0.0214	0.0176	0.0134	
0.0461	0.0023	-0.0272	0.0371
0.0545	0.0071	0.0109	
ans (:,:,1,24) =			
-0.0139	0.0072	-0.0172	0.0234
0.0626	0.0087	0.0044	
-0.0033	0.0486	-0.0216	0.0013
-0.0085	-0.0025	0.0151	
0.0175	-0.0211	-0.0005	0.0203
0.0196	-0.0188	-0.0108	
0.0392	0.0474	0.0497	0.0081
-0.0220	-0.0144	0.0356	
0.0393	0.0001	-0.0330	0.0120
0.0093	0.0094	-0.0332	
-0.0282	-0.0117	-0.0467	0.0059
0.0028	-0.0132	0.0201	
-0.0096	0.0053	-0.0228	0.0167
-0.0035	-0.0242	0.0368	
ans (:,:,2,24) =			
0.0626	0.0347	-0.0042	0.0360
0.0120	-0.0085	0.0127	
0.0280	0.0563	0.0487	-0.0011
0.0645	0.0243	0.0370	
-0.0209	0.0471	0.0061	0.0353
-0.0133	0.0321	0.0443	
0.0230	0.0480	0.0217	-0.0102
-0.0245	0.0039	-0.0156	
0.0114	0.0432	-0.0355	-0.0127
0.0247	0.0182	0.0053	
-0.0113	-0.0458	0.0152	-0.0079
0.0329	0.0008	0.0374	
0.0443	-0.0211	0.0340	0.0093
-0.0063	0.0161	0.0356	
ans (:,:,3,24) =			
0.0481	0.0587	0.0431	0.0469
-0.0021	0.0450	0.0407	
0.0341	0.0223	0.0375	0.0259
-0.0153	-0.0120	0.0427	
-0.0077	0.0029	-0.0044	0.0329
0.0298	0.0489	-0.0220	

0.0003 -0.0071 0.0251 -0.0166
0.0187 0.0431 0.0145
-0.0329 0.0246 -0.0245 -0.0033
0.0001 -0.0208 -0.0222
-0.0255 0.0109 0.0277 -0.0453
0.0204 -0.0291 -0.0211
0.0318 0.0059 0.0191 -0.0232
-0.0296 -0.0092 0.0011
ans (:,:,1,25) =
0.0158 -0.0227 0.0242 0.0196
-0.0384 0.0391 0.0012
-0.0052 -0.0104 0.0005 0.0359
-0.0356 0.0145 0.0160
0.0441 0.0529 0.0320 0.0237
0.0440 0.0032 0.0290
-0.0321 0.0197 0.0135 -0.0012
-0.0080 0.0418 0.0239
-0.0389 0.0131 0.0102 0.0183
0.0139 0.0385 0.0056
0.0156 0.0328 0.0296 0.0039
0.0386 -0.0120 0.0157
0.0575 0.0167 -0.0149 -0.0375
-0.0201 0.0103 0.0211
ans (:,:,2,25) =
-0.0020 -0.0228 0.0214 0.0206
0.0338 0.0325 0.0477
0.0196 -0.0101 -0.0428 0.0124
-0.0161 0.0349 -0.0068
0.0517 -0.0190 -0.0227 0.0040
0.0233 0.0064 0.0097
-0.0357 0.0441 0.0458 0.0377
-0.0135 -0.0141 0.0264
0.0160 0.0442 0.0557 0.0543
0.0364 -0.0066 0.0407
-0.0090 -0.0368 0.0251 -0.0258
-0.0018 -0.0191 0.0124
0.0083 0.0398 -0.0065 -0.0238
0.0021 -0.0041 -0.0080
ans (:,:,3,25) =
-0.0538 0.0097 -0.0443 0.0140
0.0215 0.0096 -0.0043
0.0280 0.0004 0.0295 -0.0190
0.0330 -0.0004 0.0268
0.0457 -0.0005 -0.0306 -0.0357
-0.0218 -0.0175 0.0064
0.0022 -0.0008 -0.0211 -0.0132
-0.0334 -0.0183 0.0469

-0.0098 0.0490 0.0579 0.0439
0.0358 -0.0168 -0.0207
0.0485 -0.0097 0.0146 -0.0271
0.0235 0.0424 0.0296
0.0116 0.0552 0.0389 -0.0022
0.0384 -0.0067 0.0118
ans (:,:,1,26) =
0.0017 -0.0298 0.0477 0.0222
-0.0263 0.0318 0.0376
-0.0272 -0.0343 0.0440 -0.0103
0.0426 0.0555 0.0074
0.0145 0.0151 -0.0277 0.0138
0.0304 0.0100 -0.0143
-0.0055 0.0334 0.0422 -0.0081
0.0391 -0.0030 0.0525
-0.0012 -0.0064 0.0205 0.0227
-0.0209 0.0286 0.0475
0.0101 -0.0293 -0.0214 0.0401
-0.0175 0.0443 -0.0051
0.0184 -0.0288 0.0542 -0.0147
0.0018 0.0384 -0.0185
ans (:,:,2,26) =
0.0383 -0.0194 0.0360 0.0344
0.0519 0.0055 0.0265
0.0032 0.0080 0.0498 -0.0334
0.0014 0.0093 0.0234
0.0032 0.0412 0.0415 0.0288
0.0230 0.0254 0.0356
-0.0403 0.0380 0.0039 -0.0259
0.0361 0.0332 -0.0135
0.0013 -0.0409 0.0127 0.0195
-0.0090 0.0288 0.0455
-0.0445 0.0076 0.0376 0.0273
-0.0233 -0.0155 -0.0082
-0.0129 -0.0121 0.0075 0.0178
-0.0006 -0.0022 0.0497
ans (:,:,3,26) =
0.0311 0.0410 0.0182 0.0504
-0.0238 0.0385 0.0540
0.0064 -0.0335 -0.0103 0.0328
0.0117 -0.0035 0.0562
0.0232 0.0359 0.0397 0.0052
0.0168 -0.0037 -0.0214
-0.0023 0.0386 0.0436 -0.0049
0.0177 0.0403 -0.0113
-0.0040 -0.0383 -0.0154 -0.0207
0.0063 0.0068 0.0428

0.0130	0.0340	0.0368	-0.0066
-0.0201	-0.0176	0.0379	
-0.0469	0.0015	0.0272	0.0381
0.0550	-0.0189	-0.0236	
 ans (:,:,1,27) =			
0.0094	-0.0175	-0.0089	0.0062
-0.0111	0.0041	-0.0400	
0.0144	0.0320	-0.0026	-0.0015
0.0098	0.0018	0.0087	
0.0446	0.0391	0.0303	0.0366
0.0590	0.0484	-0.0284	
0.0461	0.0641	0.0393	-0.0075
0.0079	0.0041	0.0358	
0.0024	0.0356	0.0454	-0.0077
0.0314	0.0329	-0.0324	
0.0285	0.0238	0.0176	0.0266
0.0711	0.0582	-0.0054	
0.0608	0.0568	-0.0129	0.0141
-0.0013	0.0722	0.0195	
 ans (:,:,2,27) =			
0.0560	-0.0025	0.0146	0.0273
-0.0153	0.0115	0.0004	
-0.0240	-0.0300	-0.0272	0.0173
0.0046	-0.0264	-0.0010	
0.0010	0.0212	-0.0054	0.0127
-0.0188	0.0080	0.0035	
0.0355	-0.0078	0.0046	-0.0016
0.0579	-0.0316	-0.0263	
0.0535	0.0525	0.0482	0.0108
-0.0115	-0.0263	0.0320	
0.0168	0.0321	0.0215	0.0386
0.0385	0.0096	-0.0318	
-0.0152	0.0480	0.0486	-0.0132
0.0227	0.0565	0.0410	
 ans (:,:,3,27) =			
0.0497	0.0420	0.0239	0.0011
0.0423	-0.0138	-0.0338	
-0.0231	0.0081	-0.0352	0.0045
0.0117	0.0181	0.0141	
0.0044	-0.0028	-0.0154	0.0174
-0.0079	0.0299	-0.0221	
0.0468	0.0612	0.0224	0.0173
-0.0227	0.0024	-0.0025	
0.0128	0.0046	0.0325	0.0048
0.0410	0.0218	0.0070	
-0.0059	-0.0050	0.0409	0.0124
0.0291	0.0066	0.0054	

<pre> 0.0047 -0.0036 0.0288 0.0063 0.0475 0.0273 0.0317 </pre>	
<pre> ans (:,:,1,28) = </pre>	
<pre> -0.0273 -0.0116 -0.0137 -0.0237 -0.0072 -0.0506 -0.0313 0.0017 -0.0424 0.0015 0.0036 -0.0130 0.0246 -0.0047 0.0141 -0.0421 -0.0248 -0.0414 -0.0505 -0.0038 -0.0168 0.0293 -0.0119 0.0170 -0.0162 -0.0297 -0.0173 -0.0296 -0.0536 -0.0102 -0.0328 -0.0053 0.0138 0.0236 -0.0410 -0.0360 0.0178 -0.0183 0.0129 0.0133 0.0041 -0.0365 -0.0364 -0.0417 0.0137 -0.0034 -0.0378 0.0146 0.0195 </pre>	
<pre> ans (:,:,2,28) = </pre>	
<pre> -0.0431 0.0149 -0.0406 0.0331 -0.0276 0.0159 -0.0051 0.0281 0.0105 0.0090 0.0265 -0.0075 -0.0492 -0.0150 0.0033 0.0211 -0.0003 0.0043 -0.0103 0.0279 -0.0424 0.0226 -0.0229 -0.0185 0.0199 0.0196 -0.0451 -0.0128 -0.0130 -0.0046 -0.0355 -0.0178 -0.0421 -0.0058 -0.0342 0.0250 -0.0459 -0.0019 -0.0237 0.0225 -0.0423 -0.0135 0.0326 -0.0484 0.0216 -0.0431 0.0150 -0.0330 0.0152 </pre>	
<pre> ans (:,:,3,28) = </pre>	
<pre> -0.0404 0.0231 -0.0418 -0.0118 0.0093 -0.0232 0.0313 -0.0175 -0.0311 -0.0432 0.0212 0.0206 -0.0381 0.0190 0.0250 0.0317 -0.0055 -0.0163 -0.0223 -0.0386 -0.0163 -0.0056 0.0157 0.0183 0.0001 0.0085 -0.0074 -0.0209 -0.0221 0.0062 -0.0436 -0.0087 -0.0093 0.0085 0.0072 -0.0374 -0.0555 -0.0043 -0.0372 -0.0100 -0.0313 -0.0102 -0.0224 -0.0115 0.0202 -0.0443 -0.0347 -0.0416 -0.0388 </pre>	

```

ans (:,:,1,29) =

```

-0.0391	-0.0294	0.0321	-0.0298
0.0199	0.0107	-0.0009	
-0.0422	0.0104	-0.0191	-0.0381
-0.0127	0.0310	0.0465	
-0.0175	-0.0290	0.0089	-0.0284
-0.0231	-0.0163	-0.0202	
-0.0281	-0.0463	0.0419	0.0497
-0.0205	-0.0230	0.0169	
0.0217	0.0318	0.0181	-0.0129
0.0307	0.0222	-0.0182	
-0.0104	0.0181	-0.0207	-0.0197
0.0191	0.0573	0.0096	
0.0072	0.0295	-0.0303	0.0506
0.0192	0.0071	0.0254	


```

ans (:,:,2,29) =

```

0.0010	0.0102	-0.0192	0.0251
0.0067	0.0534	0.0038	
0.0351	0.0189	0.0323	0.0353
-0.0050	0.0111	0.0023	
0.0147	-0.0028	0.0338	-0.0156
-0.0195	0.0435	-0.0169	
-0.0445	-0.0377	0.0362	0.0507
0.0430	-0.0200	0.0145	
-0.0254	0.0313	0.0082	0.0486
0.0282	-0.0184	-0.0221	
-0.0187	0.0021	0.0153	0.0005
0.0356	0.0402	-0.0159	
0.0201	0.0109	0.0423	0.0540
0.0456	0.0380	0.0480	


```

ans (:,:,3,29) =

```

0.0168	0.0413	0.0001	-0.0000
0.0350	0.0234	0.0095	
0.0023	0.0236	-0.0240	-0.0161
-0.0144	0.0163	0.0382	
-0.0262	0.0319	-0.0051	-0.0133
0.0263	-0.0230	-0.0186	
0.0190	0.0216	0.0060	-0.0216
0.0569	0.0368	0.0123	
-0.0327	-0.0165	-0.0320	0.0283
-0.0186	0.0384	0.0470	
-0.0369	-0.0424	0.0161	0.0030
0.0114	0.0214	0.0524	
0.0149	0.0063	0.0299	0.0298
0.0137	0.0489	-0.0124	


```

ans (:,:,1,30) =

```

-0.0357 -0.0193 0.0074 -0.0321
-0.0069 -0.0372 0.0252
0.0254 -0.0341 0.0208 0.0225
-0.0425 -0.0422 0.0346
-0.0397 0.0070 -0.0377 0.0245
-0.0126 -0.0259 0.0129
-0.0410 -0.0195 0.0291 -0.0102
-0.0140 -0.0164 -0.0088
0.0009 -0.0287 0.0017 -0.0151
0.0122 0.0150 -0.0466
-0.0165 -0.0081 -0.0564 -0.0383
-0.0355 -0.0532 -0.0068
0.0012 -0.0108 -0.0011 -0.0288
0.0168 -0.0456 -0.0239
ans (:,:,2,30) =
0.0303 0.0279 0.0071 -0.0299
-0.0303 0.0206 -0.0123
-0.0028 -0.0136 0.0174 0.0239
0.0125 -0.0264 -0.0152
-0.0272 -0.0384 -0.0349 -0.0532
-0.0545 -0.0483 0.0080
-0.0325 -0.0241 -0.0403 -0.0509
-0.0352 0.0052 -0.0064
-0.0246 -0.0092 0.0062 0.0223
0.0267 0.0282 -0.0300
-0.0503 0.0127 0.0249 -0.0002
-0.0506 -0.0396 0.0180
-0.0450 -0.0209 -0.0548 -0.0078
0.0012 0.0265 -0.0285
ans (:,:,3,30) =
0.0256 -0.0417 0.0071 -0.0127
-0.0393 -0.0031 0.0101
-0.0029 0.0300 -0.0355 -0.0421
-0.0524 -0.0140 -0.0469
-0.0217 0.0202 -0.0481 -0.0046
-0.0127 -0.0251 0.0007
-0.0164 -0.0196 -0.0276 -0.0280
-0.0164 0.0282 0.0307
-0.0284 -0.0400 -0.0317 -0.0239
-0.0254 -0.0161 -0.0005
-0.0453 0.0126 0.0234 -0.0319
-0.0172 -0.0393 0.0177
-0.0027 -0.0064 0.0219 -0.0330
-0.0052 -0.0319 0.0067
ans (:,:,1,31) =
-0.0420 -0.0077 -0.0074 -0.0205
0.0318 0.0082 0.0015

-0.0554	0.0325	0.0357	0.0277
0.0152	0.0320	-0.0119	
-0.0269	-0.0442	-0.0261	0.0420
0.0206	-0.0137	-0.0200	
0.0314	-0.0109	-0.0413	-0.0317
0.0522	-0.0002	0.0390	
-0.0550	-0.0434	-0.0326	-0.0100
0.0209	0.0026	-0.0305	
0.0221	-0.0206	0.0112	-0.0263
-0.0292	0.0006	-0.0176	
-0.0208	0.0136	-0.0325	0.0092
-0.0359	-0.0348	-0.0019	
ans (:,:,2,31) =			
-0.0303	-0.0413	0.0014	0.0150
-0.0186	0.0087	-0.0097	
-0.0449	-0.0190	-0.0315	0.0066
0.0503	0.0170	0.0067	
-0.0228	-0.0395	-0.0328	0.0180
0.0358	0.0368	-0.0089	
0.0156	0.0058	0.0322	-0.0283
0.0149	0.0092	0.0192	
0.0118	-0.0355	-0.0010	-0.0285
0.0145	0.0361	-0.0308	
-0.0420	0.0139	-0.0286	0.0250
0.0196	0.0013	0.0039	
-0.0398	-0.0421	0.0248	-0.0285
0.0270	-0.0419	-0.0320	
ans (:,:,3,31) =			
-0.0564	-0.0065	0.0277	0.0299
-0.0123	0.0072	0.0528	
-0.0215	-0.0154	0.0414	0.0237
0.0488	0.0212	0.0074	
-0.0121	0.0110	0.0101	-0.0168
-0.0113	-0.0154	0.0526	
-0.0490	0.0199	-0.0205	0.0485
0.0252	-0.0276	0.0264	
-0.0373	-0.0147	-0.0461	-0.0144
-0.0161	-0.0043	-0.0238	
-0.0537	0.0280	-0.0121	-0.0220
0.0111	-0.0030	0.0238	
0.0048	-0.0290	-0.0001	-0.0042
0.0194	-0.0488	-0.0364	
ans (:,:,1,32) =			
-0.0016	-0.0367	-0.0440	-0.0191
-0.0194	-0.0434	0.0123	
-0.0571	-0.0255	-0.0292	-0.0382
-0.0029	0.0063	-0.0525	

-0.0292 -0.0539 -0.0118 -0.0482			
-0.0258 -0.0356 -0.0057			
0.0209 0.0265 0.0111 0.0259			
0.0334 0.0345 -0.0284			
-0.0039 0.0122 0.0225 0.0099			
-0.0081 -0.0125 0.0101			
0.0267 -0.0208 -0.0169 -0.0103			
-0.0218 0.0043 0.0327			
-0.0150 0.0368 -0.0144 -0.0400			
-0.0094 -0.0461 -0.0181			
ans (:,:,2,32) =			
-0.0605 0.0032 -0.0482 0.0147			
-0.0429 -0.0073 0.0034			
-0.0223 -0.0348 -0.0023 -0.0229			
0.0199 -0.0327 -0.0419			
0.0115 -0.0222 0.0320 0.0054			
0.0330 -0.0460 -0.0424			
-0.0496 0.0293 -0.0444 -0.0081			
-0.0122 0.0356 0.0153			
-0.0019 -0.0030 0.0252 -0.0407			
0.0160 -0.0194 -0.0143			
0.0071 -0.0196 -0.0269 0.0314			
-0.0030 -0.0303 -0.0384			
0.0380 0.0402 -0.0250 0.0227			
0.0319 0.0225 -0.0231			
ans (:,:,3,32) =			
-0.0599 -0.0017 -0.0032 -0.0027			
-0.0076 -0.0385 -0.0377			
-0.0691 -0.0036 -0.0500 -0.0520			
-0.0169 0.0095 -0.0146			
-0.0448 -0.0463 -0.0406 -0.0376			
-0.0308 -0.0269 -0.0003			
0.0298 0.0022 0.0323 -0.0117			
-0.0093 -0.0062 0.0060			
-0.0271 0.0219 0.0388 0.0233			
0.0349 0.0444 0.0409			
0.0076 0.0199 -0.0010 0.0251			
-0.0320 -0.0295 0.0066			
0.0085 -0.0166 -0.0259 0.0043			
0.0269 0.0036 -0.0141			
ans (:,:,1,33) =			
0.0227 -0.0003 -0.0337 0.0213			
-0.0209 -0.0262 -0.0441			
0.0247 -0.0046 -0.0480 -0.0502			
0.0015 -0.0536 -0.0100			
-0.0603 0.0035 -0.0424 -0.0079			
-0.0407 -0.0328 -0.0437			

<pre> -0.0279 -0.0538 0.0192 0.0197 -0.0279 0.0322 0.0191 -0.0297 -0.0481 -0.0461 0.0123 -0.0094 0.0109 -0.0056 0.0349 0.0187 -0.0172 0.0204 -0.0436 -0.0332 -0.0470 0.0113 -0.0318 0.0212 0.0274 0.0247 -0.0399 0.0169 </pre>
<pre>ans (:,:,2,33) =</pre> <pre> -0.0413 -0.0323 0.0008 0.0139 -0.0413 -0.0362 -0.0457 -0.0075 0.0216 -0.0358 -0.0026 -0.0306 0.0164 -0.0223 0.0067 0.0075 0.0315 -0.0357 0.0173 -0.0342 0.0258 -0.0315 -0.0359 -0.0181 -0.0508 -0.0020 -0.0114 0.0308 -0.0338 0.0142 0.0143 0.0227 -0.0613 0.0061 -0.0375 -0.0310 -0.0110 -0.0295 -0.0306 -0.0481 -0.0344 0.0265 -0.0265 -0.0222 -0.0018 0.0283 0.0050 -0.0156 -0.0310 </pre>
<pre>ans (:,:,3,33) =</pre> <pre> -0.0167 -0.0072 0.0284 0.0119 -0.0037 -0.0025 -0.0222 -0.0504 0.0194 0.0049 0.0041 0.0246 -0.0071 -0.0310 -0.0225 -0.0277 0.0333 -0.0196 -0.0112 -0.0270 -0.0308 -0.0496 -0.0194 -0.0379 -0.0285 0.0015 0.0059 0.0222 -0.0013 -0.0028 -0.0527 -0.0000 -0.0112 -0.0347 -0.0219 -0.0375 0.0005 0.0028 -0.0155 -0.0540 0.0026 -0.0281 -0.0233 -0.0323 -0.0149 -0.0155 -0.0483 0.0246 0.0176 </pre>
<pre>ans (:,:,1,34) =</pre> <pre> -0.0428 -0.0431 -0.0164 0.0138 -0.0049 0.0237 -0.0028 -0.0080 0.0176 -0.0118 -0.0446 -0.0522 -0.0303 0.0187 -0.0088 0.0207 -0.0173 -0.0512 -0.0623 -0.0491 -0.0469 0.0119 0.0322 -0.0403 -0.0288 0.0119 0.0216 0.0168 </pre>

<pre> -0.0471 0.0013 0.0270 -0.0430 -0.0153 -0.0054 -0.0044 0.0350 0.0283 0.0033 0.0107 -0.0355 0.0119 -0.0164 0.0196 0.0201 -0.0050 -0.0268 0.0187 -0.0257 -0.0510 </pre>
<pre> ans (:,:,2,34) = </pre>
<pre> 0.0325 -0.0491 -0.0122 -0.0541 -0.0232 -0.0087 -0.0455 -0.0497 0.0283 0.0115 -0.0222 -0.0375 -0.0272 0.0045 0.0289 0.0262 -0.0263 -0.0232 -0.0641 -0.0594 -0.0364 -0.0156 0.0266 -0.0452 -0.0253 0.0017 -0.0518 0.0030 -0.0126 -0.0265 -0.0315 -0.0218 0.0053 -0.0438 -0.0138 0.0016 -0.0080 0.0192 -0.0153 -0.0052 -0.0359 -0.0055 0.0393 -0.0153 -0.0288 0.0306 -0.0037 -0.0550 0.0141 </pre>
<pre> ans (:,:,3,34) = </pre>
<pre> -0.0307 0.0298 -0.0128 0.0236 -0.0245 0.0203 0.0108 -0.0065 -0.0141 -0.0449 -0.0670 0.0111 -0.0077 -0.0422 0.0294 -0.0289 0.0092 -0.0312 -0.0270 0.0118 0.0241 0.0363 -0.0003 -0.0432 -0.0381 -0.0017 -0.0113 0.0162 -0.0446 -0.0100 -0.0523 -0.0383 -0.0325 -0.0150 0.0011 -0.0414 -0.0460 0.0097 -0.0237 0.0030 0.0160 -0.0323 -0.0105 -0.0283 -0.0270 0.0324 -0.0382 -0.0328 0.0049 </pre>
<pre> ans (:,:,1,35) = </pre>
<pre> 0.0662 -0.0087 -0.0162 -0.0347 -0.0383 -0.0423 0.0200 0.0038 0.0290 0.0389 -0.0322 -0.0297 0.0126 -0.0054 0.0075 0.0003 0.0003 -0.0153 0.0488 -0.0284 -0.0303 0.0292 0.0086 0.0173 0.0517 0.0412 0.0254 -0.0201 0.0459 -0.0164 0.0125 0.0590 0.0110 0.0297 0.0135 </pre>

-0.0281	0.0426	0.0569	0.0227
-0.0082	0.0561	0.0326	
-0.0175	0.0161	-0.0103	-0.0085
0.0625	0.0361	0.0104	
 ans (:,:,2,35) =			
0.0445	-0.0046	-0.0136	0.0355
-0.0259	-0.0135	-0.0244	
0.0445	0.0419	0.0412	0.0012
-0.0225	0.0024	-0.0109	
0.0304	0.0497	-0.0259	-0.0244
0.0483	0.0105	-0.0363	
0.0229	-0.0221	0.0333	0.0202
-0.0215	0.0148	0.0120	
-0.0265	0.0195	0.0438	0.0165
0.0417	0.0612	0.0174	
-0.0310	0.0304	0.0092	-0.0181
0.0293	0.0005	-0.0144	
-0.0244	0.0158	0.0115	0.0213
0.0431	0.0662	0.0262	
 ans (:,:,3,35) =			
0.0312	0.0344	0.0489	-0.0102
-0.0363	0.0085	-0.0184	
-0.0157	0.0401	0.0368	-0.0159
-0.0138	-0.0265	0.0052	
0.0290	0.0302	0.0417	0.0446
-0.0226	0.0026	0.0149	
0.0521	-0.0255	0.0020	0.0145
0.0474	0.0386	-0.0040	
0.0380	-0.0269	0.0373	0.0031
0.0632	0.0463	-0.0046	
0.0197	0.0279	0.0184	-0.0172
0.0521	0.0379	-0.0130	
-0.0329	0.0229	-0.0016	-0.0170
0.0264	0.0636	0.0799	
 ans (:,:,1,36) =			
-0.0257	0.0120	0.0190	0.0010
0.0244	0.0276	0.0291	
0.0259	-0.0317	0.0443	-0.0112
0.0044	-0.0090	0.0160	
-0.0289	0.0115	0.0071	-0.0144
0.0259	0.0491	0.0336	
-0.0453	0.0039	0.0024	-0.0143
0.0413	-0.0092	-0.0205	
0.0182	0.0183	-0.0250	-0.0067
0.0069	-0.0111	-0.0130	
-0.0363	-0.0171	-0.0131	0.0292
0.0227	-0.0001	0.0507	

$\begin{array}{cccc} -0.0342 & -0.0247 & -0.0397 & 0.0378 \\ 0.0323 & 0.0388 & -0.0077 & \end{array}$	
$\text{ans}(:,:,2,36) =$ $\begin{array}{cccc} -0.0418 & 0.0421 & 0.0420 & -0.0243 \\ 0.0602 & -0.0010 & 0.0340 & \\ -0.0076 & 0.0440 & -0.0080 & -0.0100 \\ 0.0302 & 0.0248 & 0.0670 & \\ 0.0140 & 0.0049 & 0.0399 & -0.0015 \\ 0.0184 & -0.0109 & -0.0032 & \\ -0.0010 & -0.0170 & 0.0407 & 0.0430 \\ 0.0540 & 0.0466 & -0.0186 & \\ 0.0199 & 0.0329 & -0.0250 & -0.0058 \\ 0.0039 & -0.0054 & 0.0488 & \\ 0.0168 & 0.0045 & 0.0398 & 0.0563 \\ 0.0415 & 0.0325 & -0.0266 & \\ 0.0096 & 0.0218 & 0.0001 & -0.0293 \\ -0.0186 & 0.0239 & 0.0335 & \end{array}$	
$\text{ans}(:,:,3,36) =$ $\begin{array}{cccc} -0.0250 & -0.0204 & 0.0334 & 0.0608 \\ 0.0667 & 0.0267 & 0.0342 & \\ -0.0395 & -0.0038 & 0.0033 & 0.0399 \\ 0.0236 & 0.0686 & -0.0004 & \\ 0.0162 & 0.0422 & 0.0562 & -0.0145 \\ -0.0077 & -0.0087 & -0.0026 & \\ -0.0365 & 0.0355 & -0.0218 & 0.0017 \\ 0.0355 & -0.0088 & 0.0240 & \\ -0.0312 & -0.0186 & -0.0183 & -0.0192 \\ 0.0263 & 0.0118 & 0.0532 & \\ -0.0013 & 0.0262 & 0.0341 & -0.0196 \\ 0.0155 & 0.0162 & 0.0485 & \\ -0.0364 & 0.0011 & 0.0173 & 0.0475 \\ 0.0555 & -0.0200 & 0.0428 & \end{array}$	
$\text{ans}(:,:,1,37) =$ $\begin{array}{cccc} -0.0209 & -0.0433 & -0.0070 & 0.0255 \\ -0.0527 & 0.0085 & -0.0677 & \\ 0.0048 & -0.0437 & -0.0246 & 0.0206 \\ -0.0004 & 0.0245 & 0.0020 & \\ 0.0285 & -0.0395 & 0.0276 & -0.0399 \\ -0.0079 & 0.0041 & -0.0328 & \\ 0.0007 & -0.0198 & -0.0124 & 0.0321 \\ -0.0134 & 0.0181 & -0.0339 & \\ -0.0316 & 0.0116 & -0.0265 & -0.0040 \\ -0.0426 & -0.0421 & -0.0414 & \\ 0.0105 & -0.0376 & -0.0163 & -0.0227 \\ 0.0240 & 0.0103 & -0.0461 & \\ 0.0070 & -0.0341 & 0.0250 & 0.0177 \\ -0.0298 & 0.0341 & 0.0252 & \end{array}$	

<pre>ans (:,:,2,37) =</pre> $\begin{matrix} -0.0069 & 0.0105 & -0.0341 & -0.0466 \\ -0.0376 & -0.0272 & -0.0085 & \\ 0.0113 & -0.0227 & -0.0195 & 0.0199 \\ 0.0102 & -0.0440 & 0.0070 & \\ -0.0053 & -0.0176 & 0.0026 & -0.0427 \\ -0.0118 & 0.0168 & -0.0144 & \\ -0.0004 & 0.0424 & 0.0395 & -0.0207 \\ 0.0052 & 0.0007 & 0.0058 & \\ -0.0025 & -0.0161 & 0.0026 & -0.0034 \\ 0.0208 & 0.0094 & 0.0275 & \\ -0.0331 & 0.0046 & -0.0241 & 0.0191 \\ -0.0305 & 0.0197 & 0.0249 & \\ 0.0064 & 0.0295 & -0.0343 & 0.0099 \\ 0.0280 & -0.0025 & -0.0397 & \end{matrix}$	
<pre>ans (:,:,3,37) =</pre> $\begin{matrix} -0.0107 & -0.0240 & -0.0216 & 0.0129 \\ -0.0554 & -0.0315 & -0.0029 & \\ -0.0385 & -0.0093 & -0.0400 & 0.0120 \\ -0.0271 & -0.0120 & 0.0053 & \\ -0.0262 & -0.0006 & 0.0049 & -0.0468 \\ -0.0423 & 0.0113 & -0.0533 & \\ 0.0243 & 0.0027 & 0.0322 & -0.0058 \\ 0.0008 & -0.0365 & 0.0010 & \\ 0.0299 & -0.0148 & 0.0088 & 0.0036 \\ -0.0311 & -0.0074 & -0.0468 & \\ -0.0389 & 0.0201 & -0.0010 & -0.0175 \\ -0.0081 & -0.0321 & -0.0297 & \\ 0.0191 & -0.0391 & -0.0326 & -0.0153 \\ -0.0384 & -0.0421 & 0.0239 & \end{matrix}$	
<pre>ans (:,:,1,38) =</pre> $\begin{matrix} 0.0200 & 0.0028 & 0.0309 & 0.0584 \\ 0.0553 & -0.0339 & 0.0210 & \\ 0.0122 & -0.0188 & -0.0285 & 0.0455 \\ 0.0437 & 0.0394 & 0.0042 & \\ -0.0027 & 0.0052 & -0.0265 & -0.0060 \\ -0.0062 & -0.0249 & -0.0020 & \\ -0.0022 & 0.0257 & 0.0173 & 0.0422 \\ 0.0581 & -0.0265 & 0.0221 & \\ 0.0055 & 0.0365 & 0.0253 & 0.0021 \\ 0.0552 & -0.0313 & -0.0533 & \\ 0.0604 & 0.0602 & -0.0122 & 0.0372 \\ 0.0426 & 0.0029 & -0.0477 & \\ -0.0129 & -0.0064 & -0.0092 & 0.0126 \\ 0.0456 & -0.0489 & -0.0626 & \end{matrix}$	
<pre>ans (:,:,2,38) =</pre>	

0.0665 0.0294 -0.0163 0.0244
0.0358 0.0092 -0.0102
0.0430 0.0318 0.0064 0.0074
0.0540 0.0288 -0.0533
-0.0118 0.0246 0.0155 0.0367
0.0284 0.0092 0.0198
-0.0142 -0.0093 0.0420 0.0042
0.0004 -0.0271 -0.0163
0.0426 0.0053 0.0595 -0.0119
-0.0013 -0.0255 -0.0055
0.0483 0.0163 0.0025 -0.0196
0.0347 -0.0384 -0.0001
-0.0178 0.0137 0.0169 0.0020
0.0264 -0.0482 0.0180
ans (:,:,3,38) =
-0.0135 0.0473 -0.0129 0.0174
0.0294 0.0363 -0.0230
-0.0032 -0.0090 0.0003 0.0182
0.0204 -0.0217 0.0165
0.0238 -0.0067 0.0182 0.0309
0.0115 -0.0180 -0.0074
-0.0216 -0.0073 0.0070 0.0442
0.0088 -0.0175 -0.0141
0.0304 -0.0170 0.0396 0.0390
0.0030 0.0278 -0.0322
-0.0212 0.0184 0.0657 -0.0023
0.0386 0.0283 -0.0112
0.0010 0.0523 0.0038 0.0216
0.0097 -0.0152 -0.0312
ans (:,:,1,39) =
-0.0031 -0.0067 -0.0071 -0.0117
-0.0344 -0.0182 -0.0231
0.0340 0.0114 -0.0244 0.0049
0.0178 -0.0290 0.0020
-0.0374 -0.0509 -0.0290 0.0191
-0.0072 -0.0237 -0.0179
-0.0095 -0.0263 -0.0466 -0.0085
0.0095 -0.0054 -0.0039
0.0417 -0.0060 -0.0317 0.0119
-0.0527 -0.0338 -0.0189
-0.0204 0.0122 -0.0346 0.0142
-0.0536 -0.0613 -0.0006
0.0212 0.0055 -0.0157 -0.0364
-0.0721 -0.0232 -0.0305
ans (:,:,2,39) =
0.0412 0.0009 0.0094 -0.0222
-0.0185 -0.0556 -0.0643

0.0110 -0.0280 -0.0395 0.0340
0.0238 -0.0473 0.0170
0.0342 0.0304 -0.0256 -0.0411
-0.0045 -0.0043 -0.0149
0.0138 -0.0126 0.0242 -0.0228
-0.0534 -0.0526 -0.0318
0.0299 0.0154 0.0203 -0.0410
-0.0589 -0.0184 -0.0025
-0.0333 0.0061 -0.0123 -0.0452
-0.0121 -0.0106 -0.0406
-0.0108 0.0300 -0.0555 -0.0446
-0.0373 -0.0190 -0.0127
 ans (:,:,3,39) =
0.0185 0.0004 0.0021 0.0405
-0.0215 -0.0023 -0.0610
-0.0216 -0.0022 -0.0247 0.0251
-0.0201 0.0141 -0.0538
-0.0097 0.0261 -0.0396 0.0178
-0.0487 -0.0257 0.0032
0.0255 0.0379 -0.0028 -0.0370
0.0157 -0.0136 -0.0036
0.0404 0.0407 -0.0317 -0.0035
-0.0027 -0.0336 -0.0379
-0.0313 0.0108 0.0193 -0.0506
-0.0046 0.0043 0.0083
0.0271 -0.0042 -0.0063 0.0075
-0.0559 -0.0197 -0.0655
 ans (:,:,1,40) =
-0.0257 0.0037 -0.0557 -0.0202
-0.0578 -0.0370 0.0134
0.0172 0.0221 -0.0027 -0.0488
0.0006 0.0290 0.0096
-0.0188 -0.0180 0.0191 -0.0418
0.0142 0.0242 -0.0412
-0.0122 -0.0052 0.0167 0.0555
0.0498 -0.0156 0.0136
-0.0434 0.0264 -0.0221 0.0015
0.0299 0.0206 -0.0612
-0.0328 -0.0064 -0.0373 -0.0312
-0.0349 -0.0123 -0.0722
-0.0717 -0.0240 -0.0029 0.0218
-0.0142 -0.0293 -0.0143
 ans (:,:,2,40) =
-0.0328 -0.0356 -0.0356 -0.0409
-0.0145 -0.0563 -0.0436
-0.0183 0.0228 -0.0216 -0.0089
0.0045 -0.0099 -0.0471

0.0183 -0.0132 -0.0325 -0.0366
-0.0320 0.0086 0.0102
-0.0583 0.0259 0.0416 0.0366
0.0280 -0.0297 -0.0325
-0.0086 -0.0085 0.0344 0.0411
0.0430 -0.0576 -0.0035
-0.0840 -0.0027 0.0134 0.0264
-0.0305 -0.0291 -0.0261
-0.0809 -0.0006 -0.0073 -0.0093
-0.0168 -0.0384 -0.0245
ans (:,:,3,40) =
-0.0018 0.0183 -0.0437 -0.0398
-0.0105 -0.0264 -0.0535
-0.0426 -0.0213 -0.0482 -0.0339
-0.0185 -0.0158 -0.0193
-0.0098 -0.0047 0.0366 -0.0069
0.0324 0.0065 0.0164
-0.0589 -0.0257 0.0106 -0.0053
0.0303 0.0140 -0.0126
-0.0295 0.0185 0.0040 0.0011
-0.0223 0.0017 -0.0719
-0.0197 -0.0545 0.0147 0.0034
0.0001 -0.0344 -0.0328
-0.0852 -0.0411 0.0035 0.0253
-0.0106 0.0001 -0.0384
ans (:,:,1,41) =
-0.0244 0.0379 0.0465 -0.0185
0.0116 0.0289 0.0061
0.0356 0.0055 0.0025 0.0362
0.0010 -0.0109 0.0415
0.0079 0.0358 -0.0056 0.0399
-0.0295 -0.0039 0.0376
0.0082 0.0439 0.0373 0.0293
-0.0145 -0.0218 -0.0285
0.0369 0.0215 -0.0081 0.0432
0.0519 0.0470 -0.0031
-0.0143 0.0121 0.0267 0.0341
0.0074 -0.0154 0.0327
0.0162 0.0305 0.0421 0.0029
0.0503 0.0621 0.0012
ans (:,:,2,41) =
0.0137 0.0601 0.0038 0.0034
-0.0229 0.0200 0.0033
0.0401 0.0199 0.0408 0.0527
0.0051 0.0033 -0.0002
-0.0029 -0.0154 -0.0213 -0.0054
-0.0133 0.0299 -0.0114

0.0411 -0.0098 -0.0156 0.0072
0.0076 -0.0041 -0.0326
0.0414 0.0049 0.0177 0.0229
0.0194 0.0323 0.0313
0.0403 -0.0116 0.0484 0.0365
0.0208 -0.0036 0.0456
0.0367 0.0571 0.0417 0.0015
0.0133 0.0041 0.0639
 ans (:,:,3,41) =
-0.0167 0.0183 0.0087 0.0332
0.0090 0.0273 -0.0009
-0.0225 0.0436 0.0565 0.0214
0.0341 -0.0040 0.0136
0.0024 0.0016 0.0220 0.0350
0.0494 -0.0173 0.0202
-0.0100 0.0633 0.0630 -0.0304
-0.0317 -0.0056 0.0072
-0.0103 0.0387 0.0441 0.0108
0.0252 -0.0199 -0.0247
0.0046 0.0494 0.0505 0.0088
0.0061 0.0028 0.0396
-0.0189 0.0140 0.0361 0.0474
0.0267 0.0416 0.0520
 ans (:,:,1,42) =
0.0141 0.0369 0.0285 -0.0060
-0.0079 0.0066 -0.0615
0.0206 0.0415 0.0438 0.0433
-0.0504 -0.0613 -0.0372
0.0285 0.0297 0.0366 0.0162
-0.0262 -0.0408 -0.0014
0.0064 0.0474 0.0453 -0.0146
0.0041 -0.0458 0.0011
0.0040 -0.0251 -0.0137 0.0392
-0.0054 -0.0038 -0.0044
0.0451 0.0171 0.0055 -0.0191
0.0028 0.0102 -0.0564
-0.0385 -0.0318 0.0011 -0.0514
-0.0513 0.0198 -0.0177
 ans (:,:,2,42) =
-0.0012 -0.0051 0.0736 0.0017
0.0170 0.0157 -0.0713
0.0506 0.0534 0.0337 -0.0215
-0.0162 0.0068 -0.0797
0.0569 0.0449 0.0300 -0.0070
-0.0416 -0.0583 -0.0426
-0.0159 0.0061 -0.0263 -0.0007
-0.0282 -0.0244 -0.0676

0.0151	0.0325	-0.0308	0.0213
-0.0527	0.0121	-0.0112	
-0.0203	0.0395	0.0302	-0.0286
-0.0046	-0.0122	-0.0320	
0.0344	-0.0100	-0.0070	-0.0091
-0.0229	-0.0083	-0.0070	
 ans (:,:,3,42) =			
0.0442	0.0134	0.0315	0.0172
-0.0246	-0.0236	-0.0105	
-0.0019	0.0052	0.0360	0.0094
0.0175	-0.0047	-0.0400	
0.0058	-0.0072	0.0119	-0.0095
0.0282	-0.0555	-0.0215	
0.0452	0.0598	0.0142	-0.0386
-0.0157	-0.0448	0.0099	
0.0485	-0.0048	-0.0269	-0.0080
-0.0473	0.0075	-0.0013	
0.0339	0.0412	-0.0027	0.0005
-0.0517	0.0133	-0.0339	
-0.0033	0.0403	0.0105	-0.0469
-0.0412	-0.0068	-0.0144	
 ans (:,:,1,43) =			
0.0143	0.0298	0.0279	0.0415
0.0014	-0.0275	0.0328	
0.0081	-0.0221	-0.0114	-0.0169
-0.0196	-0.0164	0.0455	
0.0289	0.0191	-0.0337	0.0222
0.0365	-0.0072	0.0363	
0.0388	0.0353	0.0098	-0.0292
0.0416	0.0364	-0.0155	
0.0273	-0.0246	0.0232	0.0215
-0.0275	0.0208	0.0080	
0.0057	0.0330	-0.0135	-0.0074
-0.0062	0.0141	0.0415	
0.0005	0.0002	0.0258	0.0377
0.0024	0.0399	0.0093	
 ans (:,:,2,43) =			
0.0007	-0.0112	0.0457	-0.0105
0.0492	0.0191	0.0013	
-0.0197	0.0068	0.0097	0.0283
0.0039	0.0062	-0.0361	
0.0272	0.0452	-0.0126	0.0354
-0.0100	-0.0184	0.0398	
-0.0180	-0.0001	0.0415	0.0216
-0.0202	0.0064	0.0537	
-0.0158	-0.0268	-0.0214	0.0418
-0.0037	0.0388	0.0051	

0.0441	0.0315	0.0385	-0.0267
-0.0040	0.0078	-0.0041	
-0.0149	0.0496	0.0010	0.0484
0.0099	0.0520	-0.0087	

ans (:,:,3,43) =

-0.0264	0.0332	0.0120	0.0252
0.0225	-0.0018	0.0433	
0.0032	0.0399	0.0559	0.0269
0.0306	0.0049	-0.0133	
0.0310	0.0356	-0.0191	0.0245
0.0260	-0.0191	-0.0221	
0.0089	-0.0223	-0.0350	0.0298
-0.0396	0.0210	0.0122	
0.0445	0.0014	0.0270	-0.0290
0.0016	0.0258	-0.0176	
0.0440	0.0443	0.0101	0.0287
0.0260	0.0524	0.0522	
-0.0171	0.0108	0.0278	0.0207
0.0392	0.0648	0.0056	

ans (:,:,1,44) =

0.0077	0.0212	-0.0245	0.0380
-0.0129	0.0390	0.0485	
-0.0188	0.0476	0.0537	0.0336
-0.0245	-0.0313	0.0277	
0.0381	0.0225	-0.0030	0.0426
0.0324	-0.0160	0.0293	
0.0080	-0.0140	-0.0230	-0.0005
0.0495	0.0020	0.0055	
-0.0245	-0.0305	0.0277	0.0167
0.0458	0.0079	0.0158	
-0.0051	0.0490	0.0272	-0.0291
-0.0082	0.0016	0.0347	
0.0569	-0.0036	0.0462	0.0508
0.0058	-0.0190	0.0457	

ans (:,:,2,44) =

0.0087	0.0084	0.0238	-0.0204
0.0038	0.0462	-0.0258	
-0.0182	0.0092	0.0184	-0.0106
0.0393	-0.0194	0.0348	
0.0475	-0.0321	0.0295	0.0161
-0.0025	0.0103	-0.0113	
0.0162	0.0358	0.0348	-0.0251
0.0519	0.0085	-0.0047	
-0.0011	-0.0137	0.0047	-0.0092
0.0315	0.0486	0.0026	
-0.0215	-0.0028	0.0257	0.0225
-0.0182	0.0194	0.0170	

$\begin{array}{cccc} -0.0143 & 0.0427 & -0.0248 & -0.0271 \\ 0.0504 & 0.0178 & -0.0307 & \end{array}$	
$\text{ans}(:,:,3,44) =$ $\begin{array}{cccc} 0.0472 & 0.0020 & 0.0487 & -0.0205 \\ 0.0168 & -0.0034 & 0.0318 & \\ -0.0039 & 0.0273 & 0.0553 & 0.0240 \\ 0.0381 & -0.0057 & 0.0465 & \\ 0.0178 & -0.0361 & 0.0376 & 0.0230 \\ 0.0348 & 0.0101 & 0.0148 & \\ 0.0095 & -0.0245 & -0.0093 & 0.0406 \\ 0.0379 & 0.0522 & 0.0093 & \\ 0.0465 & -0.0103 & 0.0263 & -0.0252 \\ 0.0007 & 0.0048 & 0.0409 & \\ 0.0380 & -0.0032 & -0.0223 & 0.0201 \\ -0.0046 & -0.0010 & 0.0492 & \\ -0.0152 & -0.0180 & -0.0054 & -0.0205 \\ 0.0087 & -0.0050 & -0.0118 & \end{array}$	
$\text{ans}(:,:,1,45) =$ $\begin{array}{cccc} 0.0138 & -0.0556 & 0.0056 & -0.0477 \\ -0.0691 & -0.0357 & -0.0299 & \\ -0.0211 & 0.0095 & 0.0088 & -0.0486 \\ 0.0139 & -0.0454 & -0.0324 & \\ -0.0250 & 0.0219 & -0.0317 & -0.0373 \\ -0.0464 & -0.0176 & 0.0221 & \\ -0.0170 & 0.0250 & 0.0390 & 0.0004 \\ 0.0151 & -0.0328 & -0.0249 & \\ 0.0227 & 0.0010 & 0.0238 & 0.0273 \\ -0.0025 & -0.0211 & -0.0183 & \\ 0.0040 & -0.0171 & -0.0096 & -0.0042 \\ -0.0291 & 0.0187 & -0.0486 & \\ 0.0216 & -0.0088 & -0.0294 & -0.0359 \\ -0.0238 & -0.0278 & -0.0344 & \end{array}$	
$\text{ans}(:,:,2,45) =$ $\begin{array}{cccc} -0.0064 & -0.0673 & -0.0035 & -0.0576 \\ -0.0506 & -0.0547 & -0.0062 & \\ 0.0149 & -0.0495 & 0.0004 & -0.0475 \\ -0.0504 & -0.0253 & 0.0024 & \\ 0.0203 & -0.0247 & 0.0011 & -0.0589 \\ -0.0343 & -0.0031 & -0.0397 & \\ -0.0228 & -0.0260 & 0.0194 & 0.0268 \\ -0.0436 & 0.0186 & 0.0103 & \\ -0.0220 & 0.0059 & 0.0431 & 0.0144 \\ 0.0202 & 0.0021 & -0.0167 & \\ -0.0181 & -0.0084 & 0.0337 & 0.0465 \\ -0.0076 & -0.0053 & 0.0094 & \\ 0.0200 & 0.0156 & 0.0003 & -0.0008 \\ 0.0330 & 0.0206 & 0.0111 & \end{array}$	

<pre>ans (:,:,3,45) =</pre> $\begin{matrix} -0.0124 & -0.0124 & -0.0653 & -0.0174 \\ 0.0028 & -0.0208 & -0.0146 & \\ -0.0370 & -0.0525 & -0.0593 & -0.0644 \\ -0.0435 & -0.0157 & -0.0522 & \\ 0.0414 & -0.0318 & -0.0599 & -0.0171 \\ 0.0064 & 0.0161 & -0.0542 & \\ 0.0330 & -0.0278 & -0.0256 & -0.0455 \\ -0.0110 & -0.0228 & 0.0146 & \\ -0.0434 & -0.0329 & 0.0165 & 0.0077 \\ -0.0403 & -0.0381 & -0.0031 & \\ -0.0435 & 0.0232 & 0.0511 & -0.0030 \\ -0.0228 & 0.0004 & -0.0612 & \\ 0.0232 & -0.0260 & 0.0368 & 0.0119 \\ 0.0222 & 0.0044 & -0.0234 & \end{matrix}$	
<pre>ans (:,:,1,46) =</pre> $\begin{matrix} -0.0300 & -0.0191 & 0.0225 & 0.0183 \\ -0.0145 & 0.0380 & 0.0051 & \\ 0.0211 & -0.0068 & 0.0070 & 0.0395 \\ 0.0283 & -0.0134 & -0.0213 & \\ 0.0014 & 0.0417 & 0.0384 & 0.0191 \\ 0.0559 & 0.0222 & 0.0048 & \\ 0.0479 & -0.0154 & 0.0272 & -0.0019 \\ 0.0351 & -0.0155 & 0.0483 & \\ -0.0040 & 0.0347 & -0.0204 & 0.0485 \\ 0.0414 & -0.0016 & 0.0560 & \\ -0.0186 & 0.0433 & -0.0125 & 0.0250 \\ -0.0016 & 0.0086 & 0.0007 & \\ -0.0018 & 0.0046 & 0.0290 & 0.0385 \\ 0.0420 & 0.0022 & 0.0258 & \end{matrix}$	
<pre>ans (:,:,2,46) =</pre> $\begin{matrix} 0.0074 & -0.0238 & 0.0268 & 0.0478 \\ -0.0182 & 0.0327 & 0.0473 & \\ 0.0185 & -0.0115 & 0.0086 & -0.0050 \\ 0.0339 & -0.0289 & 0.0280 & \\ 0.0470 & 0.0263 & 0.0446 & 0.0173 \\ 0.0153 & -0.0206 & -0.0212 & \\ -0.0134 & -0.0181 & 0.0404 & -0.0115 \\ 0.0091 & -0.0247 & -0.0076 & \\ -0.0111 & 0.0224 & -0.0218 & 0.0182 \\ 0.0143 & 0.0120 & 0.0392 & \\ 0.0071 & -0.0161 & 0.0289 & 0.0356 \\ 0.0569 & 0.0425 & -0.0151 & \\ 0.0277 & -0.0122 & 0.0208 & -0.0230 \\ 0.0030 & 0.0216 & -0.0234 & \end{matrix}$	
<pre>ans (:,:,3,46) =</pre>	

0.0129	-0.0143	0.0364	0.0079
0.0300	0.0414	0.0261	
0.0372	0.0412	-0.0216	0.0158
0.0182	-0.0163	0.0302	
-0.0099	0.0422	-0.0189	-0.0104
-0.0041	0.0565	0.0603	
0.0265	-0.0008	0.0109	0.0435
0.0147	0.0095	0.0215	
0.0291	-0.0181	0.0303	0.0035
0.0329	0.0465	0.0113	
-0.0122	-0.0255	0.0146	-0.0027
0.0077	0.0126	0.0252	
-0.0210	0.0133	0.0412	0.0098
-0.0177	-0.0305	-0.0033	
ans (:,:,1,47) =			
0.0543	0.0112	-0.0295	0.0148
-0.0143	0.0481	-0.0084	
0.0034	0.0162	0.0297	-0.0132
0.0300	-0.0290	0.0087	
0.0350	0.0374	0.0390	-0.0184
-0.0343	-0.0022	-0.0133	
0.0246	-0.0063	0.0288	-0.0305
-0.0171	-0.0047	-0.0191	
0.0266	-0.0265	-0.0044	0.0469
-0.0123	-0.0028	0.0552	
0.0411	0.0092	-0.0015	0.0254
-0.0375	0.0029	0.0270	
0.0539	0.0433	-0.0011	0.0149
-0.0068	0.0120	0.0059	
ans (:,:,2,47) =			
0.0191	0.0393	-0.0207	0.0005
-0.0387	0.0434	0.0155	
0.0204	0.0657	0.0239	-0.0339
0.0142	0.0266	0.0114	
0.0687	0.0568	-0.0088	0.0455
0.0349	0.0193	-0.0186	
-0.0021	0.0590	0.0251	-0.0148
-0.0295	-0.0148	0.0400	
0.0110	-0.0241	-0.0254	0.0198
0.0424	0.0188	0.0445	
-0.0169	0.0363	0.0435	-0.0312
-0.0358	-0.0164	0.0138	
0.0115	0.0436	0.0096	-0.0142
-0.0269	0.0370	0.0431	
ans (:,:,3,47) =			
0.0373	-0.0151	0.0395	0.0137
0.0219	0.0303	-0.0053	

0.0727	0.0584	0.0431	0.0205
0.0328	0.0241	0.0466	
0.0495	0.0604	-0.0049	-0.0267
0.0034	0.0142	0.0549	
0.0461	-0.0021	-0.0123	-0.0299
-0.0136	0.0276	0.0351	
0.0569	-0.0223	0.0162	0.0128
0.0358	0.0473	-0.0235	
0.0030	-0.0176	0.0085	0.0142
-0.0323	0.0474	0.0042	
-0.0213	-0.0107	0.0466	-0.0156
0.0170	0.0040	0.0442	
ans (:,:,1,48) =			
0.0180	-0.0210	-0.0135	-0.0424
0.0070	0.0265	-0.0193	
0.0209	-0.0307	0.0110	-0.0129
-0.0408	-0.0186	-0.0398	
-0.0347	-0.0506	-0.0232	0.0038
-0.0390	-0.0448	-0.0282	
-0.0002	-0.0034	-0.0277	0.0161
0.0075	0.0161	-0.0113	
0.0238	-0.0301	-0.0499	-0.0246
-0.0312	0.0295	-0.0061	
-0.0143	0.0394	-0.0289	-0.0234
0.0096	0.0057	-0.0318	
-0.0117	-0.0144	0.0257	0.0044
-0.0385	-0.0301	-0.0119	
ans (:,:,2,48) =			
-0.0067	0.0247	0.0207	0.0157
0.0158	-0.0340	0.0254	
0.0155	-0.0288	0.0050	-0.0408
-0.0491	-0.0283	-0.0072	
0.0025	0.0189	-0.0112	-0.0130
0.0022	-0.0375	-0.0168	
0.0310	0.0050	-0.0355	-0.0427
-0.0448	-0.0261	-0.0261	
0.0187	0.0277	-0.0113	0.0180
0.0252	-0.0152	0.0085	
-0.0224	0.0248	0.0072	-0.0118
0.0288	-0.0091	-0.0161	
0.0311	0.0141	-0.0377	0.0294
0.0145	0.0062	-0.0137	
ans (:,:,3,48) =			
-0.0464	-0.0226	-0.0432	0.0099
-0.0069	-0.0273	-0.0025	
-0.0260	0.0264	0.0040	-0.0021
-0.0179	-0.0474	-0.0328	

-0.0057 0.0251 -0.0301 -0.0403
0.0344 0.0009 0.0180
0.0006 -0.0242 0.0128 -0.0151
-0.0407 -0.0496 0.0072
-0.0310 -0.0403 0.0067 -0.0336
-0.0076 0.0294 -0.0352
-0.0187 -0.0194 -0.0331 0.0089
-0.0316 -0.0172 0.0231
0.0310 0.0408 -0.0337 -0.0381
0.0202 -0.0313 -0.0278
ans (:,:,1,49) =
0.0224 -0.0532 -0.0035 -0.0368
-0.0590 -0.0712 -0.0598
-0.0074 -0.0610 0.0122 -0.0720
-0.0194 -0.0691 0.0027
0.0123 -0.0095 -0.0187 0.0061
-0.0431 -0.0372 0.0162
-0.0332 -0.0395 -0.0511 0.0058
-0.0103 -0.0365 -0.0045
0.0244 -0.0060 -0.0382 -0.0093
-0.0114 -0.0128 0.0169
0.0082 0.0186 0.0192 0.0179
0.0174 0.0081 -0.0355
0.0084 0.0117 0.0010 0.0295
0.0304 -0.0368 0.0294
ans (:,:,2,49) =
-0.0407 -0.0224 -0.0343 -0.0200
-0.0031 -0.0461 -0.0654
0.0062 -0.0417 -0.0416 0.0054
-0.0648 -0.0655 0.0016
-0.0553 0.0026 -0.0410 -0.0382
-0.0497 -0.0397 -0.0364
-0.0043 -0.0115 -0.0073 0.0279
-0.0083 0.0140 -0.0414
-0.0275 -0.0039 -0.0032 -0.0222
0.0019 -0.0052 0.0158
-0.0200 -0.0199 0.0231 0.0307
0.0251 0.0049 -0.0078
0.0243 -0.0098 -0.0247 -0.0273
0.0168 0.0020 -0.0273
ans (:,:,3,49) =
-0.0031 0.0100 -0.0728 -0.0440
-0.0620 -0.0057 -0.0174
-0.0171 -0.0385 0.0005 -0.0070
-0.0390 -0.0337 -0.0422
-0.0434 0.0120 -0.0103 -0.0264
0.0082 -0.0459 0.0010

0.0240 -0.0518 -0.0309 0.0184
-0.0090 0.0039 -0.0078
-0.0392 0.0201 -0.0344 0.0407
-0.0219 -0.0356 0.0265
0.0041 -0.0029 0.0507 -0.0097
0.0334 0.0421 -0.0300
0.0370 -0.0234 -0.0194 -0.0191
0.0136 -0.0317 0.0321
ans (:,:,1,50) =
0.0113 0.0194 0.0173 -0.0274
0.0434 0.0391 0.0359
0.0046 0.0290 0.0239 0.0266
-0.0375 0.0336 0.0246
-0.0325 0.0421 -0.0264 0.0040
0.0085 -0.0136 -0.0190
0.0293 -0.0218 0.0526 0.0096
0.0745 -0.0104 0.0513
-0.0380 -0.0027 0.0091 0.0348
0.0311 0.0361 0.0223
-0.0490 0.0079 0.0084 0.0240
0.0291 0.0252 -0.0014
0.0171 -0.0265 0.0132 -0.0271
0.0188 0.0323 0.0683
ans (:,:,2,50) =
0.0032 0.0054 0.0484 0.0378
0.0206 0.0114 -0.0309
0.0289 -0.0257 0.0167 0.0013
0.0147 -0.0195 -0.0238
-0.0280 -0.0220 0.0265 0.0365
0.0442 0.0447 0.0206
-0.0107 0.0428 -0.0166 0.0427
-0.0076 0.0060 -0.0104
-0.0156 0.0266 0.0168 0.0095
0.0554 0.0177 0.0085
-0.0184 -0.0466 0.0213 -0.0126
0.0213 0.0412 0.0557
-0.0198 -0.0073 0.0091 -0.0352
0.0435 0.0044 0.0175
ans (:,:,3,50) =
0.0236 -0.0204 -0.0284 -0.0305
-0.0052 -0.0026 -0.0209
0.0152 -0.0101 0.0284 -0.0149
0.0265 0.0174 0.0224
0.0031 0.0319 -0.0361 0.0382
0.0631 0.0162 0.0199
-0.0337 -0.0221 0.0519 0.0468
0.0576 0.0636 -0.0184

-0.0215	0.0208	0.0225	0.0082
0.0387	0.0364	0.0182	
-0.0591	0.0272	0.0047	0.0337
0.0262	0.0774	0.0231	
-0.0519	-0.0169	-0.0373	0.0074
-0.0272	-0.0061	0.0108	
 ans (:,:,1,51) =			
0.0170	0.0118	-0.0028	0.0243
0.0354	-0.0106	-0.0317	
-0.0064	-0.0124	0.0256	0.0262
0.0451	0.0427	-0.0089	
-0.0002	0.0371	-0.0316	-0.0209
-0.0324	0.0131	-0.0389	
-0.0105	-0.0064	0.0249	0.0187
-0.0312	-0.0221	-0.0011	
0.0082	-0.0040	0.0090	0.0126
-0.0248	-0.0142	0.0118	
0.0530	0.0268	-0.0224	0.0398
0.0480	0.0322	-0.0223	
0.0449	0.0282	-0.0082	0.0264
-0.0044	0.0547	0.0167	
 ans (:,:,2,51) =			
0.0407	0.0403	0.0496	-0.0154
-0.0090	0.0369	0.0326	
0.0580	0.0469	-0.0262	-0.0040
0.0464	0.0188	-0.0406	
-0.0080	0.0091	-0.0264	-0.0010
-0.0175	-0.0158	-0.0041	
-0.0270	-0.0113	0.0036	0.0109
0.0207	0.0260	0.0151	
0.0352	0.0409	0.0129	0.0103
-0.0219	0.0220	0.0253	
0.0554	0.0407	0.0247	0.0127
0.0235	0.0282	-0.0282	
-0.0038	0.0164	-0.0187	0.0007
0.0310	0.0628	-0.0083	
 ans (:,:,3,51) =			
0.0482	0.0042	0.0205	-0.0187
0.0119	-0.0103	-0.0306	
0.0397	-0.0065	0.0140	0.0444
-0.0234	0.0156	-0.0432	
0.0526	-0.0299	0.0135	-0.0216
-0.0172	-0.0348	0.0093	
0.0322	0.0134	0.0467	-0.0342
-0.0132	0.0021	0.0360	
0.0487	0.0174	0.0121	0.0155
-0.0092	-0.0158	0.0010	

<pre> 0.0364 0.0112 0.0114 0.0139 0.0431 0.0438 0.0087 -0.0033 0.0514 -0.0107 0.0402 0.0159 0.0582 0.0607 </pre>
<pre> ans (:,:,1,52) = </pre>
<pre> 0.0385 0.0111 0.0225 0.0352 -0.0294 -0.0100 -0.0207 -0.0243 0.0048 -0.0145 0.0379 -0.0035 -0.0081 0.0177 -0.0272 -0.0160 0.0028 -0.0148 0.0290 -0.0038 -0.0290 -0.0116 -0.0120 -0.0317 0.0279 -0.0363 -0.0237 0.0212 -0.0499 -0.0503 -0.0205 -0.0296 -0.0126 -0.0382 0.0102 -0.0359 -0.0113 0.0078 0.0037 -0.0042 0.0112 0.0139 0.0050 -0.0326 -0.0284 -0.0101 -0.0242 -0.0452 0.0153 </pre>
<pre> ans (:,:,2,52) = </pre>
<pre> -0.0242 0.0183 -0.0283 -0.0329 0.0113 -0.0388 0.0279 -0.0107 -0.0030 -0.0091 0.0033 0.0025 -0.0338 0.0081 0.0450 0.0133 -0.0285 -0.0016 -0.0249 -0.0079 -0.0307 -0.0046 0.0245 0.0310 -0.0034 -0.0047 0.0220 0.0194 -0.0104 -0.0284 -0.0462 -0.0316 0.0073 -0.0145 -0.0256 -0.0055 0.0171 -0.0290 -0.0357 -0.0477 0.0034 -0.0367 -0.0442 -0.0400 0.0077 0.0323 0.0247 -0.0305 -0.0096 </pre>
<pre> ans (:,:,3,52) = </pre>
<pre> -0.0295 0.0310 -0.0214 0.0188 -0.0249 -0.0139 -0.0191 -0.0069 -0.0389 -0.0139 -0.0047 0.0375 -0.0301 -0.0050 0.0360 0.0083 -0.0154 -0.0150 0.0376 -0.0058 0.0133 0.0275 0.0100 -0.0096 -0.0175 -0.0240 -0.0334 0.0150 0.0017 -0.0188 0.0205 -0.0352 -0.0130 -0.0321 -0.0426 0.0124 -0.0189 0.0149 -0.0257 -0.0176 -0.0017 -0.0153 </pre>

$\begin{array}{cccc} -0.0072 & 0.0062 & -0.0377 & -0.0352 \\ -0.0470 & -0.0407 & 0.0211 & \end{array}$	
$\text{ans}(:,:,1,53) =$ $\begin{array}{cccc} 0.0056 & -0.0359 & -0.0220 & -0.0330 \\ 0.0053 & -0.0508 & -0.0512 & \\ -0.0014 & -0.0180 & -0.0050 & -0.0351 \\ -0.0042 & -0.0074 & 0.0205 & \\ -0.0428 & -0.0198 & -0.0122 & 0.0132 \\ 0.0276 & 0.0116 & 0.0081 & \\ 0.0207 & -0.0223 & -0.0398 & 0.0230 \\ -0.0431 & -0.0349 & -0.0277 & \\ 0.0011 & 0.0374 & -0.0183 & 0.0279 \\ 0.0081 & 0.0170 & -0.0609 & \\ 0.0334 & -0.0055 & -0.0071 & -0.0200 \\ 0.0254 & -0.0241 & -0.0506 & \\ 0.0012 & 0.0499 & -0.0208 & -0.0009 \\ -0.0434 & -0.0404 & -0.0450 & \end{array}$	
$\text{ans}(:,:,2,53) =$ $\begin{array}{cccc} -0.0542 & -0.0269 & -0.0045 & -0.0295 \\ 0.0128 & -0.0259 & -0.0074 & \\ -0.0481 & 0.0245 & 0.0089 & -0.0314 \\ -0.0115 & 0.0234 & -0.0573 & \\ 0.0102 & 0.0000 & 0.0219 & 0.0065 \\ -0.0172 & 0.0183 & -0.0072 & \\ -0.0085 & -0.0357 & 0.0228 & -0.0381 \\ 0.0303 & -0.0291 & -0.0107 & \\ 0.0255 & 0.0347 & 0.0178 & 0.0178 \\ 0.0249 & -0.0165 & 0.0112 & \\ -0.0305 & -0.0250 & 0.0044 & -0.0347 \\ -0.0038 & 0.0011 & 0.0070 & \\ 0.0185 & 0.0121 & -0.0065 & -0.0408 \\ 0.0309 & -0.0358 & -0.0050 & \end{array}$	
$\text{ans}(:,:,3,53) =$ $\begin{array}{cccc} -0.0114 & -0.0174 & 0.0105 & 0.0150 \\ -0.0154 & 0.0251 & -0.0121 & \\ -0.0465 & -0.0362 & -0.0319 & -0.0416 \\ -0.0475 & -0.0559 & 0.0053 & \\ -0.0226 & 0.0118 & 0.0343 & -0.0061 \\ -0.0325 & 0.0084 & -0.0357 & \\ -0.0212 & -0.0329 & -0.0304 & -0.0321 \\ 0.0363 & -0.0338 & -0.0213 & \\ -0.0204 & -0.0371 & -0.0418 & -0.0006 \\ -0.0048 & -0.0469 & -0.0040 & \\ 0.0035 & 0.0092 & 0.0326 & -0.0331 \\ 0.0340 & -0.0053 & -0.0590 & \\ 0.0225 & -0.0255 & 0.0394 & -0.0348 \\ -0.0303 & -0.0259 & -0.0219 & \end{array}$	

<pre> ans (:,:,1,54) = 0.0004 -0.0188 -0.0298 0.0106 -0.0439 -0.0248 0.0183 -0.0314 0.0191 -0.0564 -0.0278 -0.0326 -0.0166 0.0315 0.0097 -0.0042 -0.0477 -0.0046 0.0067 -0.0015 -0.0307 0.0314 -0.0350 -0.0017 0.0036 -0.0167 -0.0124 0.0041 0.0227 0.0333 0.0304 -0.0018 -0.0343 -0.0065 0.0533 0.0274 -0.0482 -0.0179 0.0121 0.0022 0.0143 0.0210 0.0262 0.0109 -0.0416 -0.0138 0.0082 -0.0013 0.0216 </pre>	
<pre> ans (:,:,2,54) = -0.0088 -0.0247 -0.0112 -0.0319 0.0203 0.0060 -0.0136 0.0197 0.0106 -0.0253 -0.0389 -0.0357 -0.0222 -0.0195 -0.0132 -0.0347 -0.0230 0.0046 -0.0229 0.0295 -0.0166 -0.0129 0.0224 0.0328 -0.0506 -0.0086 -0.0263 -0.0109 -0.0177 -0.0392 0.0136 0.0316 -0.0073 0.0088 0.0264 -0.0473 -0.0074 -0.0395 -0.0470 0.0157 -0.0211 0.0605 0.0122 -0.0430 -0.0496 -0.0488 -0.0351 -0.0258 0.0230 </pre>	
<pre> ans (:,:,3,54) = -0.0153 -0.0306 -0.0054 -0.0030 0.0033 -0.0293 -0.0331 0.0078 0.0148 -0.0074 -0.0222 0.0212 -0.0522 0.0014 0.0169 0.0170 -0.0177 0.0045 -0.0373 -0.0302 0.0258 -0.0386 -0.0190 -0.0448 -0.0508 0.0071 -0.0329 -0.0052 -0.0011 -0.0338 0.0194 0.0240 0.0054 0.0039 -0.0085 -0.0374 -0.0279 -0.0313 -0.0112 0.0025 0.0121 0.0353 -0.0324 -0.0391 -0.0343 -0.0103 -0.0111 0.0277 0.0349 </pre>	
<pre> ans (:,:,1,55) = </pre>	

-0.0197 -0.0014 -0.0303 -0.0029			
0.0649 0.0280 -0.0049			
0.0266 -0.0071 -0.0046 0.0132			
-0.0050 0.0014 0.0597			
-0.0044 0.0483 0.0415 -0.0169			
0.0078 0.0410 -0.0009			
0.0339 0.0465 0.0327 -0.0167			
0.0270 0.0524 -0.0006			
-0.0155 0.0347 -0.0019 -0.0338			
0.0275 0.0005 0.0463			
-0.0166 0.0107 0.0194 -0.0130			
0.0046 0.0362 -0.0371			
0.0133 -0.0172 0.0056 0.0068			
-0.0090 -0.0311 -0.0302			
ans (:,:,2,55) =			
0.0114 0.0254 0.0321 0.0329			
0.0582 -0.0084 0.0020			
0.0072 -0.0270 0.0050 0.0528			
0.0535 0.0656 0.0590			
0.0366 0.0169 -0.0205 0.0551			
0.0517 0.0218 0.0001			
0.0098 0.0019 -0.0067 -0.0249			
0.0300 -0.0280 0.0124			
-0.0302 0.0228 -0.0009 0.0063			
-0.0442 -0.0316 0.0360			
-0.0250 -0.0080 -0.0044 -0.0152			
0.0151 -0.0048 -0.0339			
0.0142 0.0229 0.0040 0.0400			
0.0362 0.0337 0.0256			
ans (:,:,3,55) =			
0.0383 0.0260 0.0488 0.0400			
0.0594 0.0157 0.0365			
0.0135 0.0395 0.0360 -0.0022			
0.0557 0.0183 0.0553			
0.0075 0.0545 0.0427 0.0402			
-0.0107 -0.0032 0.0531			
0.0040 -0.0202 -0.0101 0.0315			
0.0177 0.0376 0.0559			
-0.0363 0.0003 -0.0064 0.0287			
0.0065 0.0105 0.0177			
0.0440 -0.0039 -0.0280 -0.0206			
0.0169 -0.0277 -0.0138			
0.0246 0.0303 0.0010 -0.0331			
-0.0417 0.0158 0.0359			
ans (:,:,1,56) =			
0.0489 0.0129 -0.0089 -0.0209			
-0.0172 -0.0398 -0.0281			

0.0237	0.0103	-0.0177	-0.0358
0.0003	-0.0253	0.0122	
-0.0203	0.0327	0.0017	-0.0296
-0.0086	-0.0264	-0.0486	
0.0426	-0.0395	-0.0255	0.0074
-0.0615	-0.0182	-0.0124	
0.0306	0.0264	-0.0079	0.0133
0.0080	0.0105	-0.0191	
0.0359	0.0159	0.0456	0.0220
0.0337	0.0009	0.0039	
0.0470	0.0465	-0.0069	0.0457
0.0391	0.0131	0.0062	
ans (:,:,2,56) =			
0.0366	-0.0131	0.0316	0.0132
-0.0000	-0.0509	-0.0151	
0.0387	0.0189	0.0448	-0.0541
-0.0409	0.0126	-0.0468	
-0.0226	0.0409	-0.0354	-0.0357
-0.0368	-0.0285	0.0065	
-0.0242	0.0347	0.0048	-0.0264
-0.0615	-0.0303	-0.0637	
-0.0270	-0.0075	0.0254	-0.0094
0.0022	0.0056	0.0139	
-0.0067	0.0340	0.0071	-0.0347
-0.0200	0.0283	0.0310	
0.0343	0.0231	-0.0169	0.0269
-0.0096	0.0219	0.0113	
ans (:,:,3,56) =			
0.0175	0.0086	-0.0252	-0.0460
0.0206	0.0181	-0.0089	
-0.0291	0.0104	0.0448	-0.0092
0.0033	-0.0138	0.0046	
0.0116	-0.0327	-0.0435	-0.0536
-0.0372	-0.0341	-0.0468	
0.0475	-0.0110	-0.0145	-0.0435
-0.0350	-0.0519	-0.0202	
-0.0140	0.0232	-0.0141	0.0278
0.0178	-0.0388	-0.0025	
-0.0255	-0.0161	0.0246	-0.0071
-0.0311	0.0320	-0.0016	
-0.0190	0.0339	0.0446	0.0117
0.0282	-0.0314	-0.0174	
ans (:,:,1,57) =			
0.0179	-0.0031	0.0158	0.0424
0.0597	0.0004	0.0044	
0.0204	0.0412	0.0552	0.0061
-0.0182	-0.0188	-0.0225	

-0.0196	0.0156	-0.0227	-0.0122
0.0295	0.0450	0.0159	
0.0509	0.0435	0.0417	-0.0259
-0.0144	0.0490	0.0529	
-0.0107	0.0135	0.0258	-0.0430
-0.0052	0.0037	0.0578	
-0.0293	0.0204	-0.0049	0.0391
0.0396	0.0399	0.0402	
-0.0026	-0.0009	0.0036	-0.0399
-0.0315	-0.0148	0.0048	
ans (:,:,2,57) =			
0.0146	0.0583	0.0256	0.0539
0.0302	0.0301	0.0461	
0.0339	0.0467	0.0388	0.0171
0.0552	-0.0249	-0.0021	
0.0382	0.0324	0.0414	-0.0263
0.0267	0.0325	-0.0191	
0.0168	0.0523	0.0023	-0.0049
0.0045	0.0346	0.0484	
0.0027	0.0258	-0.0402	-0.0252
0.0172	-0.0161	0.0181	
0.0115	0.0379	-0.0010	-0.0079
0.0444	-0.0168	-0.0001	
-0.0059	-0.0328	0.0214	-0.0117
0.0476	-0.0058	0.0618	
ans (:,:,3,57) =			
-0.0014	0.0251	-0.0126	0.0546
0.0442	-0.0016	0.0227	
0.0674	0.0353	-0.0205	0.0066
-0.0115	-0.0009	0.0160	
0.0252	0.0087	-0.0139	0.0082
0.0282	0.0319	-0.0059	
0.0235	-0.0168	-0.0059	0.0050
0.0130	0.0547	0.0267	
0.0040	0.0367	-0.0149	-0.0357
-0.0010	0.0125	0.0176	
0.0264	0.0192	-0.0241	-0.0338
-0.0048	-0.0127	-0.0146	
0.0085	0.0470	0.0026	-0.0326
-0.0051	-0.0138	0.0089	
ans (:,:,1,58) =			
-0.0079	0.0033	-0.0247	-0.0057
0.0307	0.0321	-0.0191	
-0.0183	0.0037	-0.0041	-0.0096
-0.0191	0.0123	-0.0015	
0.0512	0.0225	0.0496	-0.0058
-0.0216	-0.0174	-0.0368	

-0.0152 -0.0091 -0.0194 0.0435
-0.0196 -0.0262 -0.0288
-0.0295 -0.0136 0.0403 0.0057
0.0257 -0.0233 0.0056
-0.0311 0.0286 0.0530 0.0157
0.0327 0.0142 0.0192
-0.0053 0.0272 0.0295 -0.0248
-0.0170 0.0187 0.0344
ans (:,:,2,58) =
0.0420 -0.0213 0.0125 0.0293
0.0191 0.0424 -0.0160
0.0357 0.0231 0.0192 0.0608
0.0449 0.0285 0.0337
-0.0046 -0.0128 0.0327 0.0187
-0.0082 0.0418 -0.0112
0.0264 -0.0093 0.0342 0.0228
-0.0203 0.0309 -0.0134
0.0522 0.0369 -0.0200 0.0351
0.0337 0.0172 0.0260
0.0434 0.0427 -0.0268 -0.0216
0.0197 0.0021 0.0340
-0.0045 0.0274 0.0365 0.0059
-0.0168 0.0041 0.0178
ans (:,:,3,58) =
0.0345 0.0375 -0.0125 -0.0051
0.0285 0.0485 0.0152
0.0529 -0.0033 0.0426 0.0537
-0.0212 0.0558 -0.0321
0.0269 -0.0274 0.0346 0.0165
0.0144 0.0471 -0.0057
-0.0335 0.0194 0.0206 0.0013
-0.0138 0.0449 0.0002
0.0201 0.0108 0.0285 0.0027
-0.0081 0.0394 -0.0276
-0.0215 -0.0104 -0.0314 0.0436
0.0140 0.0288 0.0206
-0.0368 0.0361 -0.0266 -0.0200
-0.0279 0.0212 0.0133
ans (:,:,1,59) =
0.0381 0.0047 -0.0158 0.0525
-0.0112 0.0348 0.0442
0.0647 0.0487 0.0290 0.0586
-0.0295 0.0060 -0.0320
-0.0082 -0.0115 0.0001 0.0416
0.0377 -0.0144 0.0452
0.0044 -0.0243 -0.0113 -0.0363
0.0252 -0.0046 0.0082

<pre> -0.0385 -0.0252 -0.0430 -0.0458 0.0327 0.0337 0.0119 0.0323 -0.0055 -0.0028 0.0295 0.0316 0.0164 0.0276 0.0426 0.0318 0.0147 -0.0358 -0.0161 0.0383 -0.0070 </pre>																																																								
<pre> ans (:,:,2,59) = </pre> <table> <tbody> <tr><td>0.0134</td><td>0.0531</td><td>0.0288</td><td>-0.0169</td></tr> <tr><td>-0.0137</td><td>-0.0052</td><td>0.0166</td><td></td></tr> <tr><td>0.0603</td><td>0.0480</td><td>-0.0069</td><td>0.0524</td></tr> <tr><td>0.0028</td><td>-0.0196</td><td>0.0000</td><td></td></tr> <tr><td>0.0361</td><td>-0.0101</td><td>-0.0078</td><td>0.0045</td></tr> <tr><td>0.0220</td><td>0.0503</td><td>0.0125</td><td></td></tr> <tr><td>0.0238</td><td>0.0339</td><td>-0.0024</td><td>-0.0311</td></tr> <tr><td>0.0249</td><td>0.0135</td><td>0.0468</td><td></td></tr> <tr><td>-0.0092</td><td>0.0226</td><td>-0.0234</td><td>-0.0127</td></tr> <tr><td>-0.0122</td><td>-0.0314</td><td>-0.0316</td><td></td></tr> <tr><td>0.0194</td><td>0.0079</td><td>-0.0182</td><td>-0.0321</td></tr> <tr><td>0.0088</td><td>-0.0372</td><td>0.0085</td><td></td></tr> <tr><td>-0.0298</td><td>-0.0015</td><td>-0.0064</td><td>0.0047</td></tr> <tr><td>-0.0162</td><td>0.0091</td><td>0.0213</td><td></td></tr> </tbody> </table>	0.0134	0.0531	0.0288	-0.0169	-0.0137	-0.0052	0.0166		0.0603	0.0480	-0.0069	0.0524	0.0028	-0.0196	0.0000		0.0361	-0.0101	-0.0078	0.0045	0.0220	0.0503	0.0125		0.0238	0.0339	-0.0024	-0.0311	0.0249	0.0135	0.0468		-0.0092	0.0226	-0.0234	-0.0127	-0.0122	-0.0314	-0.0316		0.0194	0.0079	-0.0182	-0.0321	0.0088	-0.0372	0.0085		-0.0298	-0.0015	-0.0064	0.0047	-0.0162	0.0091	0.0213	
0.0134	0.0531	0.0288	-0.0169																																																					
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<pre> ans (:,:,3,59) = </pre> <table> <tbody> <tr><td>0.0169</td><td>0.0055</td><td>0.0103</td><td>-0.0206</td></tr> <tr><td>-0.0222</td><td>0.0382</td><td>0.0281</td><td></td></tr> <tr><td>0.0252</td><td>-0.0156</td><td>0.0496</td><td>0.0423</td></tr> <tr><td>0.0056</td><td>0.0162</td><td>0.0209</td><td></td></tr> <tr><td>-0.0061</td><td>0.0336</td><td>0.0275</td><td>0.0158</td></tr> <tr><td>0.0436</td><td>0.0266</td><td>0.0412</td><td></td></tr> <tr><td>0.0296</td><td>0.0525</td><td>-0.0334</td><td>0.0047</td></tr> <tr><td>-0.0388</td><td>-0.0227</td><td>-0.0184</td><td></td></tr> <tr><td>-0.0314</td><td>-0.0225</td><td>0.0127</td><td>0.0379</td></tr> <tr><td>0.0224</td><td>-0.0150</td><td>0.0206</td><td></td></tr> <tr><td>0.0270</td><td>0.0320</td><td>0.0197</td><td>-0.0317</td></tr> <tr><td>-0.0295</td><td>0.0345</td><td>-0.0314</td><td></td></tr> <tr><td>-0.0220</td><td>0.0376</td><td>0.0176</td><td>0.0087</td></tr> <tr><td>-0.0396</td><td>0.0298</td><td>0.0085</td><td></td></tr> </tbody> </table>	0.0169	0.0055	0.0103	-0.0206	-0.0222	0.0382	0.0281		0.0252	-0.0156	0.0496	0.0423	0.0056	0.0162	0.0209		-0.0061	0.0336	0.0275	0.0158	0.0436	0.0266	0.0412		0.0296	0.0525	-0.0334	0.0047	-0.0388	-0.0227	-0.0184		-0.0314	-0.0225	0.0127	0.0379	0.0224	-0.0150	0.0206		0.0270	0.0320	0.0197	-0.0317	-0.0295	0.0345	-0.0314		-0.0220	0.0376	0.0176	0.0087	-0.0396	0.0298	0.0085	
0.0169	0.0055	0.0103	-0.0206																																																					
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-0.0396	0.0298	0.0085																																																						
<pre> ans (:,:,1,60) = </pre> <table> <tbody> <tr><td>0.0121</td><td>0.0077</td><td>0.0024</td><td>0.0086</td></tr> <tr><td>0.0348</td><td>-0.0012</td><td>0.0235</td><td></td></tr> <tr><td>0.0064</td><td>-0.0236</td><td>-0.0060</td><td>-0.0338</td></tr> <tr><td>-0.0231</td><td>0.0060</td><td>-0.0216</td><td></td></tr> <tr><td>0.0239</td><td>0.0126</td><td>-0.0360</td><td>0.0255</td></tr> <tr><td>0.0301</td><td>0.0228</td><td>-0.0033</td><td></td></tr> <tr><td>-0.0452</td><td>-0.0322</td><td>-0.0573</td><td>-0.0220</td></tr> <tr><td>-0.0191</td><td>-0.0350</td><td>-0.0314</td><td></td></tr> <tr><td>-0.0435</td><td>-0.0519</td><td>0.0084</td><td>-0.0036</td></tr> <tr><td>0.0281</td><td>-0.0080</td><td>0.0134</td><td></td></tr> </tbody> </table>	0.0121	0.0077	0.0024	0.0086	0.0348	-0.0012	0.0235		0.0064	-0.0236	-0.0060	-0.0338	-0.0231	0.0060	-0.0216		0.0239	0.0126	-0.0360	0.0255	0.0301	0.0228	-0.0033		-0.0452	-0.0322	-0.0573	-0.0220	-0.0191	-0.0350	-0.0314		-0.0435	-0.0519	0.0084	-0.0036	0.0281	-0.0080	0.0134																	
0.0121	0.0077	0.0024	0.0086																																																					
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-0.0435	-0.0519	0.0084	-0.0036																																																					
0.0281	-0.0080	0.0134																																																						

<pre> -0.0507 0.0216 -0.0034 0.0257 -0.0174 0.0241 0.0042 -0.0087 0.0197 -0.0080 0.0172 -0.0457 -0.0009 0.0041 </pre>
<pre> ans (:,:,2,60) = </pre>
<pre> -0.0263 -0.0152 0.0117 0.0078 -0.0261 -0.0142 -0.0276 0.0026 -0.0328 -0.0130 0.0334 -0.0415 -0.0442 0.0213 -0.0158 -0.0300 -0.0313 0.0207 0.0040 -0.0155 0.0307 -0.0020 0.0143 -0.0598 -0.0507 0.0344 -0.0040 0.0244 0.0249 -0.0573 -0.0064 -0.0009 -0.0340 0.0311 -0.0206 -0.0493 -0.0461 -0.0505 -0.0242 0.0235 0.0051 -0.0348 0.0113 -0.0558 -0.0508 -0.0097 0.0222 0.0037 -0.0289 </pre>
<pre> ans (:,:,3,60) = </pre>
<pre> 0.0143 -0.0178 0.0317 -0.0215 0.0047 0.0303 -0.0479 -0.0003 -0.0014 -0.0298 0.0049 0.0256 -0.0272 0.0277 -0.0342 -0.0096 -0.0050 -0.0045 0.0333 -0.0279 -0.0419 0.0265 0.0127 -0.0565 -0.0288 0.0027 0.0045 0.0185 -0.0007 0.0198 -0.0432 -0.0211 -0.0450 -0.0076 -0.0006 -0.0447 -0.0189 -0.0272 -0.0491 -0.0311 0.0159 0.0320 0.0099 -0.0196 -0.0543 -0.0086 -0.0318 0.0264 -0.0287 </pre>
<pre> ans (:,:,1,61) = </pre>
<pre> 0.0065 0.0423 -0.0156 0.0118 -0.0038 0.0298 0.0179 0.0382 -0.0323 0.0304 0.0046 0.0175 0.0429 0.0157 0.0483 -0.0159 -0.0326 0.0345 0.0076 -0.0198 -0.0202 0.0184 -0.0017 0.0273 0.0014 0.0039 -0.0016 -0.0004 0.0075 0.0331 0.0309 -0.0183 0.0150 -0.0265 0.0244 0.0588 -0.0204 -0.0167 0.0299 -0.0286 0.0417 0.0567 </pre>

$\begin{array}{cccc} -0.0288 & -0.0228 & 0.0342 & -0.0164 \\ 0.0256 & 0.0125 & 0.0651 & \end{array}$	
$\text{ans}(:,:,2,61) =$ $\begin{array}{cccc} 0.0417 & 0.0213 & -0.0191 & -0.0004 \\ 0.0468 & 0.0401 & 0.0457 & \\ 0.0301 & 0.0029 & 0.0172 & 0.0176 \\ 0.0065 & 0.0481 & 0.0213 & \\ 0.0359 & 0.0099 & -0.0307 & -0.0051 \\ -0.0132 & -0.0047 & -0.0097 & \\ 0.0284 & -0.0203 & -0.0162 & -0.0158 \\ 0.0444 & 0.0087 & 0.0563 & \\ -0.0022 & -0.0148 & -0.0260 & -0.0261 \\ 0.0079 & 0.0075 & -0.0070 & \\ 0.0520 & -0.0028 & 0.0358 & 0.0177 \\ 0.0269 & -0.0163 & 0.0015 & \\ -0.0093 & 0.0153 & 0.0073 & -0.0108 \\ -0.0009 & 0.0175 & 0.0288 & \end{array}$	
$\text{ans}(:,:,3,61) =$ $\begin{array}{cccc} -0.0108 & 0.0151 & 0.0195 & 0.0007 \\ 0.0050 & 0.0267 & -0.0178 & \\ 0.0255 & 0.0363 & 0.0138 & -0.0143 \\ 0.0236 & 0.0516 & 0.0172 & \\ -0.0204 & 0.0063 & 0.0315 & 0.0122 \\ -0.0201 & 0.0419 & 0.0328 & \\ 0.0084 & 0.0436 & 0.0361 & 0.0211 \\ 0.0367 & -0.0219 & 0.0455 & \\ 0.0430 & 0.0341 & -0.0071 & -0.0419 \\ -0.0386 & 0.0366 & -0.0046 & \\ 0.0152 & -0.0150 & 0.0303 & -0.0252 \\ -0.0033 & -0.0004 & 0.0638 & \\ -0.0106 & 0.0267 & -0.0025 & 0.0388 \\ 0.0160 & 0.0174 & 0.0214 & \end{array}$	
$\text{ans}(:,:,1,62) =$ $\begin{array}{cccc} -0.0152 & -0.0035 & -0.0651 & -0.0625 \\ -0.0477 & -0.0459 & 0.0047 & \\ -0.0573 & 0.0160 & -0.0130 & -0.0632 \\ -0.0044 & -0.0210 & 0.0188 & \\ -0.0399 & 0.0025 & -0.0075 & -0.0304 \\ 0.0136 & -0.0333 & -0.0144 & \\ -0.0169 & 0.0218 & -0.0484 & 0.0138 \\ -0.0419 & 0.0174 & -0.0092 & \\ -0.0443 & -0.0380 & 0.0133 & -0.0365 \\ -0.0304 & 0.0100 & -0.0010 & \\ 0.0147 & 0.0134 & -0.0606 & -0.0021 \\ -0.0367 & -0.0096 & -0.0111 & \\ -0.0416 & -0.0515 & 0.0011 & -0.0135 \\ -0.0124 & 0.0302 & -0.0275 & \end{array}$	

```

ans (:,:,2,62) =
    0.0321   -0.0328    0.0015    0.0157
-0.0546   -0.0381   -0.0227
    -0.0220   -0.0439    0.0120    0.0137
-0.0590    0.0088   -0.0261
    -0.0303    0.0073   -0.0051    0.0040
-0.0244   -0.0107    0.0372
    -0.0375   -0.0266   -0.0284   -0.0473
-0.0411   -0.0164   -0.0002
    -0.0363   -0.0161   -0.0109   -0.0178
-0.0298   -0.0119   -0.0093
    0.0073   -0.0471    0.0047    0.0128
-0.0320   -0.0476    0.0212
    -0.0433    0.0187   -0.0299    0.0167
-0.0294    0.0160    0.0102

ans (:,:,3,62) =
   -0.0446    0.0261   -0.0411    0.0024
-0.0490   -0.0192    0.0272
    0.0002   -0.0585    0.0111   -0.0078
-0.0300   -0.0436   -0.0382
    -0.0042   -0.0049   -0.0107   -0.0127
-0.0284    0.0218    0.0047
    0.0221   -0.0115   -0.0406    0.0246
-0.0480    0.0066    0.0043
    -0.0410   -0.0168   -0.0106   -0.0502
0.0194   -0.0579    0.0076
    0.0065   -0.0298   -0.0243   -0.0086
-0.0389   -0.0396    0.0289
    0.0070   -0.0012    0.0134   -0.0144
0.0231    0.0285    0.0392

ans (:,:,1,63) =
    0.0269    0.0218   -0.0201    0.0054
0.0404    0.0113   -0.0195
    0.0121   -0.0170    0.0098    0.0112
-0.0235    0.0201   -0.0397
    -0.0052   -0.0373   -0.0282   -0.0323
0.0480   -0.0402    0.0141
    -0.0288    0.0131   -0.0087   -0.0237
0.0405    0.0378   -0.0201
    -0.0355   -0.0156    0.0041   -0.0345
-0.0065   -0.0306   -0.0453
    -0.0567   -0.0354   -0.0605    0.0103
0.0215    0.0202   -0.0144
    0.0174    0.0094    0.0150   -0.0407
-0.0598   -0.0395    0.0014

ans (:,:,2,63) =

```

-0.0271 -0.0040 0.0296 -0.0120
-0.0095 -0.0422 -0.0347
0.0454 -0.0066 -0.0280 -0.0285
-0.0301 0.0115 0.0156
0.0201 -0.0344 0.0151 -0.0093
0.0418 0.0210 -0.0107
-0.0196 -0.0420 -0.0042 0.0075
0.0149 -0.0193 -0.0222
-0.0013 -0.0411 -0.0429 0.0198
-0.0119 -0.0316 0.0141
0.0076 -0.0079 -0.0087 -0.0180
0.0091 0.0266 0.0127
-0.0367 0.0120 0.0032 -0.0527
-0.0023 -0.0174 0.0057
ans (:,:,3,63) =
0.0030 -0.0373 0.0251 0.0040
-0.0244 -0.0047 -0.0363
0.0332 -0.0292 0.0015 0.0027
0.0025 -0.0058 -0.0517
-0.0281 -0.0080 -0.0214 0.0085
-0.0182 -0.0431 -0.0403
-0.0007 -0.0048 0.0252 0.0086
-0.0078 -0.0188 0.0317
-0.0185 -0.0055 -0.0182 0.0081
0.0325 0.0255 0.0035
-0.0423 -0.0602 -0.0082 -0.0390
-0.0473 -0.0398 0.0046
-0.0387 -0.0038 -0.0534 -0.0554
-0.0183 -0.0461 -0.0039
ans (:,:,1,64) =
0.0222 0.0096 -0.0160 -0.0101
0.0202 0.0103 0.0414
0.0109 0.0124 -0.0176 0.0269
-0.0016 -0.0178 0.0065
0.0283 -0.0073 -0.0036 -0.0031
-0.0106 0.0043 0.0165
-0.0138 -0.0071 0.0377 0.0029
-0.0250 -0.0503 0.0153
0.0080 0.0413 0.0468 0.0495
0.0446 0.0269 -0.0293
0.0555 -0.0096 0.0427 0.0645
-0.0127 0.0243 0.0243
0.0227 0.0240 0.0189 0.0463
0.0087 -0.0235 0.0082
ans (:,:,2,64) =
0.0076 0.0058 0.0564 -0.0165
0.0220 -0.0047 -0.0127

-0.0260	0.0222	0.0231	0.0215
-0.0094	0.0004	0.0111	
-0.0305	-0.0074	-0.0272	0.0048
0.0161	-0.0178	-0.0288	
-0.0191	-0.0247	0.0145	0.0122
0.0398	0.0271	-0.0194	
-0.0255	0.0360	0.0053	0.0180
-0.0122	0.0401	0.0117	
0.0017	0.0189	0.0340	0.0561
0.0493	-0.0015	-0.0162	
0.0010	0.0459	0.0201	0.0321
-0.0032	-0.0098	0.0116	
 ans (:,:,3,64) =			
0.0405	-0.0114	-0.0147	0.0004
0.0199	0.0110	-0.0297	
0.0253	0.0549	-0.0065	-0.0312
0.0157	-0.0168	-0.0216	
0.0468	0.0375	-0.0257	-0.0155
-0.0075	0.0276	0.0374	
0.0150	0.0155	0.0535	-0.0128
0.0022	-0.0114	-0.0338	
0.0336	-0.0095	0.0411	-0.0113
0.0095	0.0249	0.0277	
0.0028	0.0433	0.0588	0.0647
0.0360	-0.0276	-0.0023	
0.0036	0.0180	0.0175	-0.0099
-0.0139	0.0315	-0.0074	