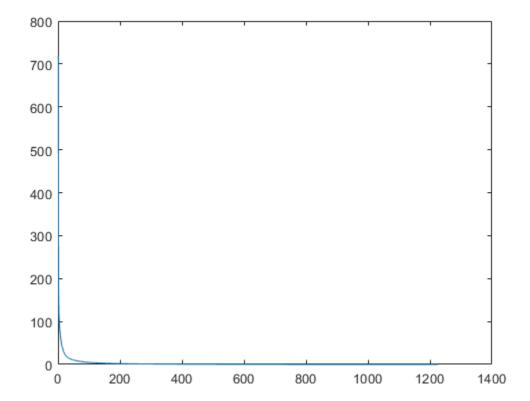
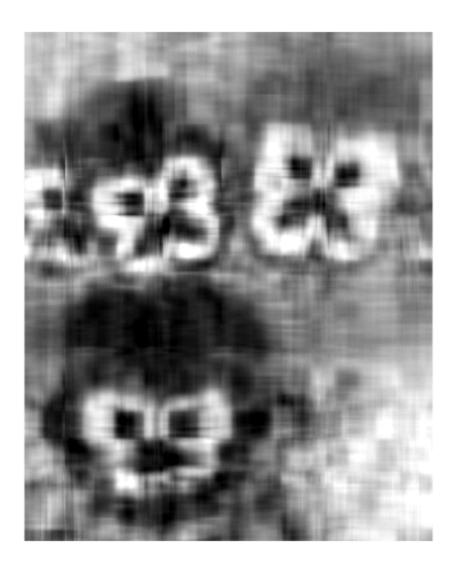
```
imdata = imread('panises.jpg'); %load the image
A = im2double(imdata); %convert from uint8 matrix to double
[U,S,V] = svd(A); %compute SVD

s = svds(A,1225); %singular values in decreasing order
plt = plot(s); %plot out all singular values of A
% The singular values of A decay extremely fast, mimicing an exponential
% decay
```



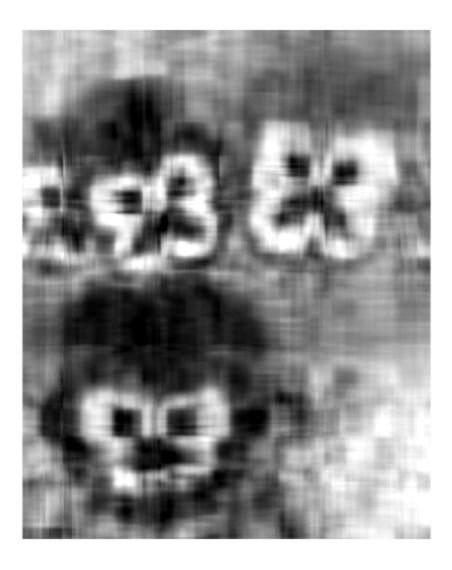
k=10

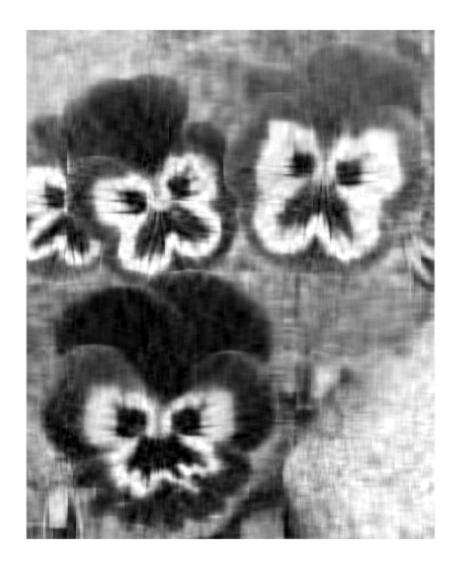
% is difficult to identify each petal of each flower.



```
%k=25
for N=25
% store the singular values in a temporary variable
C = S;
% discard the diagonal values not required for compression
C(N+1:end,:)=0;
C(:,N+1:end)=0;
% Construct an Image using the selected singular values
Newim25=U*C*V';
% display
figure;
imshow(Newim25);
end
```

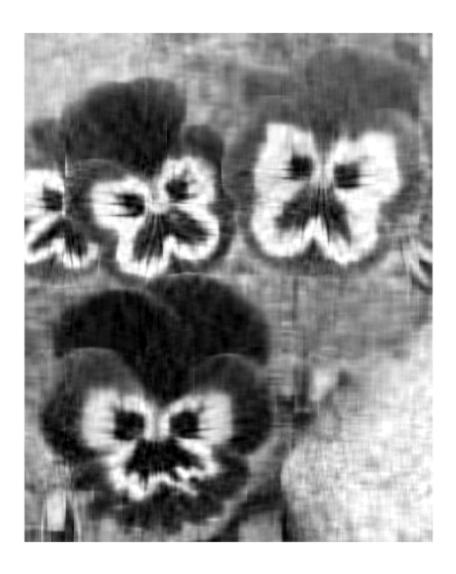
- $\mbox{\ensuremath{\mbox{$\kappa$}}}$ Compared to that using $k=10\,,$ this image is slightly clearer. Lines for
- % parts of the petals are sharper.





```
%k=50
for N=50
% store the singular values in a temporary variable
C = S;
% discard the diagonal values not required for compression
C(N+1:end,:)=0;
C(:,N+1:end)=0;
% Construct an Image using the selected singular values
Newim50=U*C*V';
% display
figure;
imshow(Newim50);
end
```

- $\mbox{\ensuremath{\mbox{$\kappa$}}}$ Compared to the results $k{=}10$ and $k{=}25\,,$ outlines for each petal are now
- % much clearer. However, the stems are still difficult to identify





```
%k=100
for N=100
% store the singular values in a temporary variable
C = S;
% discard the diagonal values not required for compression
C(N+1:end,:)=0;
C(:,N+1:end)=0;
% Construct an Image using the selected singular values
Newim100=U*C*V';
% display
figure;
imshow(Newim100);
end
```

- % Most clear image among all results. Both the outlines of the petals and
- % stems, little leaves on the stems can be identified. Somewhat smooth % transition between different shades of gray too.





- $\mbox{\ensuremath{\$}}$ One way to identify the optimal parameter k could via the computation of
- $\mbox{\ensuremath{\$}}$ error from the approximated matrix. As the decrease in error comes at the
- $\mbox{\ensuremath{\$}}$ price of increasing k (and thus increasing cost of memory and computing
- % power), we want to choose k right before it introduce a significant drop
- % in error.
- % initiate an empty array to contain errors
 error = zeros([1 100]);
 for N=1:1:100 %check the error for k = 1: 100
 % store the singular values in a temporary variable
 C = S;

```
% discard the diagonal values not required for compression
C(N+1:end,:)=0;
C(:,N+1:end)=0;
% Construct an Image using the selected singular values
ErrIm=U*C*V';
% finding errors and add to the error array
error(N) = sum(sum((ErrIm - A).^2));
plterr=plot(error);
end
% from the error plot, it seems that the plot tends to have a linear
% decrease after k=40. The optimal k = 40.
```

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