#### AMATH 482 Homework 5

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March 16<sup>th</sup>, 2021

#### **Abstract**

I used the Dynamic Mode Decomposition method on the video clips ski\_drop.mov and monte\_carlo.mov to separate the video stream to both the foreground video and a background.

#### Introduction and Overview

This assignment poses a problem when separating the DMD terms into approximate low-rank and sparse reconstructions: the DMD reconstruction is complex, but the real valued outputs are desired and knowing how to handle the complex elements can make a significant difference in the accuracy of the results. Doing all the maths, only negative values turned out. However, it does not make any sense of having negative pixel intensities. Finding a way to make pixel intensities to be higher or equal to zero is important. This could make the method work well.

#### Theoretical Background

In this project includes the implementation of Dynamic Mode Decomposition:

$$\mathbf{U}^*\mathbf{A}\mathbf{U} = \underbrace{\mathbf{U}^*\mathbf{X}_2^M\mathbf{V}\Sigma^{-1}}_{=:\tilde{\mathbf{S}}}.$$

# Algorithm Implementation and Development

I first calculated two submatrices for X and then compute the SVD of  $\mathbf{x}_1^{M-1}$ . Then I created a matrix  $\tilde{\mathbf{s}} = \mathbf{U}^* \mathbf{x}_2^M \mathbf{v} \mathbf{\Sigma}^{-1}$  to find the eigenvalues and eigenvectors. After that, I used the initial snapshot x1 to find the coefficients  $b_k$ . Therefore, I could compute the solution at any future time using the DMD modes along with their projection to the initial conditions and the time dynamics by using the eigenvalues of  $\tilde{\mathbf{S}}$ .

## Computational Results

I was finding a hard time to calculate the S, V, D out, so there are no results available. I put the reasonable plot codes in the appendix.

# **Summary and Conclusions**

Same as above.

### Appendix A

v = VideoReader(filename) creates object v to read video data from the file named filename.

real(z) returns the real part of z. If z is a matrix, real acts elementwise on z.

Y =  $\exp(X)$  returns the exponential  $e^x$  for each element in array X. For complex elements z = x + iy, it returns the complex exponential

# Appendix B

```
clc; clear all
mtcl = VideoReader('monte carlo.mov')
skdp = VideoReader('ski drop.mov')
numFrames = 0
data = zeros(,1)
col = zeros(540*960, 379)
[U, S, V] = svd(mtcl)
for col < 379
    numFrames = numFrames + 1
   mtcl2 = readFrame(mtcl)
end
subplot(2, 1, 1)
plot(diag(S), 'ko', 'Linewidth', 2)
ylabel('\sigma j')
set(gca, 'Fontsize', 16, 'Xlim', [0.9 delays+0.1])
subplot(2,1,2)
plot(t(1:end-delays+1), V(:,1), 'r', 'Linewidth', 2)
plot(t(1:end-delays+1)), V(:,2), 'b--', 'Linewidth', 2)
xlabel('t')
ylabel('v_j(t)')
```

```
set(gca,'Fontsize',16)

X1 = V(1:end-1,1:2)';
X2 = V(2:end-1,1:2)';

[U2, S2, V2] = svd(X1, 'econ');
Stilde = U2'*X*V2*diag(1./diag(S2));
[eV, D] = eig(D);
omega = log(mu)/dt;
Phi = U2*eV;
y0 = Phi\X1(:1);
u_modes = zeros(length(y0), length(t));
for iter = 1:length(t)
    u_modes(:,iter) = y0.*exp(omega*t(iter));
end
u_dmd = real(Phi*u_modes);
```