## Generating correlated random vectors



- Straightforward to implement KF steps in Octave
- To exercise KF code, however, also must simulate system whose state is being estimated
  - □ Produces system input/output data as input to the KF
  - □ Provides "true" state value at every point in time
- But, to simulate true system, must be able to create nonzero mean Gaussian noise with covariance  $\Sigma_{\widetilde{Y}}$ .
- That is, we want  $Y \sim \mathcal{N}(\bar{y}, \Sigma_{\widetilde{Y}})$  but randn.m returns  $X \sim \mathcal{N}(0, I)$ .
  - $\square$  How to convert X to Y?

Dr. Gregory L. Plett

Iniversity of Colorado Colorado Springs

Battery State-of-Charge (SOC) Estimation | Coming to understand the linear Kalman filter

1 of 5

3.3.2: Introducing Octave code to generate correlated random vectors

#### A needed transformation



- Suppose that we can find matrix A such that  $A^TA = \Sigma_{\widetilde{Y}}$
- Then, we generate samples x of random vector  $X \sim \mathcal{N}(0, I)$  and compute samples y of random vector  $Y = \bar{y} + A^T X$ 
  - $\square$  Since X is zero-mean,  $\mathbb{E}[Y] = \mathbb{E}[\bar{y} + A^T X] = \bar{y}$ , as desired
  - $\Box$  The covariance of Y is

$$\mathbb{E}[(Y - \bar{y})(Y - \bar{y})^T] = \mathbb{E}[(A^T X)(A^T X)^T]$$

$$= \mathbb{E}[(A^T X)(X^T A)]$$

$$= A^T \underbrace{\mathbb{E}[XX^T]}_{I} A = \Sigma_{\widetilde{Y}},$$

also as desired

Dr. Gregory L. Plet

Iniversity of Colorado Colorado Spring

Battery State-of-Charge (SOC) Estimation | Coming to understand the linear Kalman filter

2 of

3.3.2: Introducing Octave code to generate correlated random vectors

# Important matrix factorizations



- So, if we can find A such that  $A^TA = \Sigma_{\widetilde{Y}}$ , we can generate samples x of  $X \sim \mathcal{N}(0, I)$  and compute  $y = \bar{y} + A^T x$
- $\blacksquare$  Three ways to generate A
  - $\Box$  Cholesky factorization of  $\Sigma_{\widetilde{Y}}$  computes A such that  $A^TA = \Sigma_{\widetilde{Y}}$  as long as  $\Sigma_{\widetilde{Y}}$  is positive definite (all eigenvalues are strictly greater than zero)
  - $\Box$  LDL factorization of  $\Sigma_{\widetilde{Y}}$  produces L,D such that  $LDL^T=\Sigma_{\widetilde{Y}}$  and requires only that  $\Sigma_{\widetilde{Y}}$  be positive semi definite: Can then compute  $A=\sqrt{D}L^T$
  - $\square$  LU factorization of  $\Sigma_{\widetilde{Y}}$  produces L,U such that  $LU=\Sigma_{\widetilde{Y}}$  and requires only that  $\Sigma_{\widetilde{Y}}$  be positive semi definite: Can then compute  $A=\operatorname{diag}\left(\sqrt{\operatorname{diag}(U)}\right)L^T$
- Cholesky and LU are built into Octave; all are built into MATLAB

### **Example Octave code**

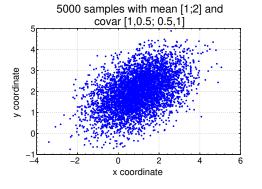


■ Suppose we wish to generate a single sample of *y* using Cholesky method:

```
ybar = [1; 2];
covar = [1, 0.5; 0.5, 1];
A = chol(covar);
x = randn([2, 1]);
y = ybar + A'*x;
```

Suppose we wish to generate 5000 samples of *y* using the LU method:

```
ybar = [1; 2];
covar = [1, 0.5; 0.5, 1];
[L,U] = lu(covar);
A = diag(sqrt(diag(U)))*L';
x = randn([2,5000]);
y = ybar(:,ones([1 5000]))+A'*x;
plot(y(1,:),y(2,:),'.');
```



Dr. Gregory L. Plett | University of Colorado Colorado Springs

Battery State-of-Charge (SOC) Estimation | Coming to understand the linear Kalman filter | 4 of 5

3.3.2: Introducing Octave code to generate correlated random vectors

### Summary



- In order to test KF code via simulation, must also simulate system whose state is being estimated
- This requires ability to simulate (possibly) correlated (possibly nonzero mean) noises
- Octave natively produces samples of zero-mean uncorrelated Gaussians only
- But, can transform these to have desired mean and correlation using a matrix determined using either the Cholesky or LU matrix decompositions, both of which are built into Octave

Dr. Gregory L. Plett | University of Colorado Colorado Springs

Battery State-of-Charge (SOC) Estimation | Coming to understand the linear Kalman filter | 5 of 5