## Some More Matrix and Statistical Results

Definition: If $\mathbf{y}$ is a vector random variable of length $n$ with mean vector $E(\mathbf{y})=\mu$, covariance matrix $V(\mathbf{y})=\boldsymbol{\Sigma}$, and has a probability density function of:

$$
f(\mathbf{y})=|2 \pi \boldsymbol{\Sigma}|^{-1 / 2} \exp \left\{-\frac{1}{2}(\mathbf{y}-\boldsymbol{\mu})^{\prime} \boldsymbol{\Sigma}^{-1}(\mathbf{y}-\boldsymbol{\mu})\right\}
$$

then $\mathbf{y}$ has a multivariate normal distribution, denoted by $\mathbf{y} \sim N_{n}(\boldsymbol{\mu}, \boldsymbol{\Sigma})$.
Note that this is a slightly more general version than shown on page 197 of the text, since here $V(\mathbf{y})=\boldsymbol{\Sigma}$ whereas on that page they assume $V(\mathbf{y})=\sigma_{\varepsilon}^{2} \mathbf{I}_{n}$.

Result 3: Let $\mathbf{y} \sim N_{n}(\boldsymbol{\mu}, \boldsymbol{\Sigma})$, let $\mathbf{C}$ be an $m \mathbf{x} n$ matrix, and $\mathbf{d}$ be an $m \mathbf{x} 1$ vector. Then $\mathbf{C y}+\mathbf{d} \sim N_{m}\left(\mathbf{C} \boldsymbol{\mu}+\mathbf{d}, \mathbf{C} \boldsymbol{\Sigma} \mathbf{C}^{\prime}\right)$.

## References

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