

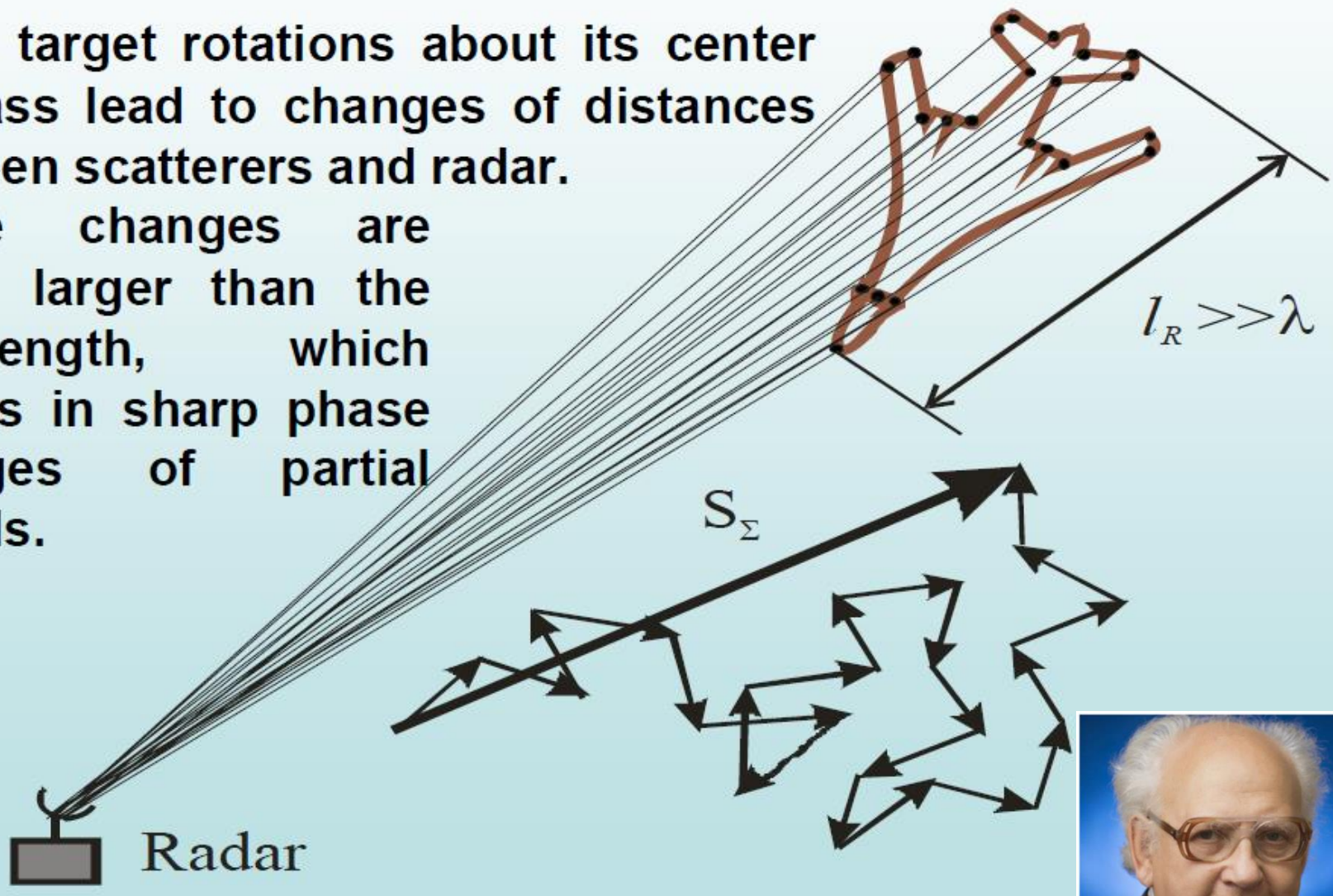
DETERMINISTIC TARGETS

Sphere, plane, corner reflector, antenna

STATISTICAL TARGETS

Multiple scatterers, fluctuating targets, distributions, Swerling models

Small target rotations about its center of mass lead to changes of distances between scatterers and radar. These changes are much larger than the wavelength, which results in sharp phase changes of partial signals.



Victor Chernyak:

Tutorial at the RADAR 2010, 10 May, 2010, Arlington VA, USA



Professor Viktor Chernyak
1929 - 2020

Signal from a Multi-Scatterer Target

- Assume N scatterers
 - individual RCS σ_i
 - fixed individual ranges R_i
- To within a constant, we have by superposition, in field units

$$y(t) = \sum_{i=1}^N \sqrt{\sigma_i} \exp\left[j2\pi f(t - 2R_i/C)\right] = \exp(j2\pi ft) \sum_{i=1}^N \sqrt{\sigma_i} \exp(-j4\pi R_i/\lambda)$$

The voltage of the envelope detected field is $\zeta = |y|$

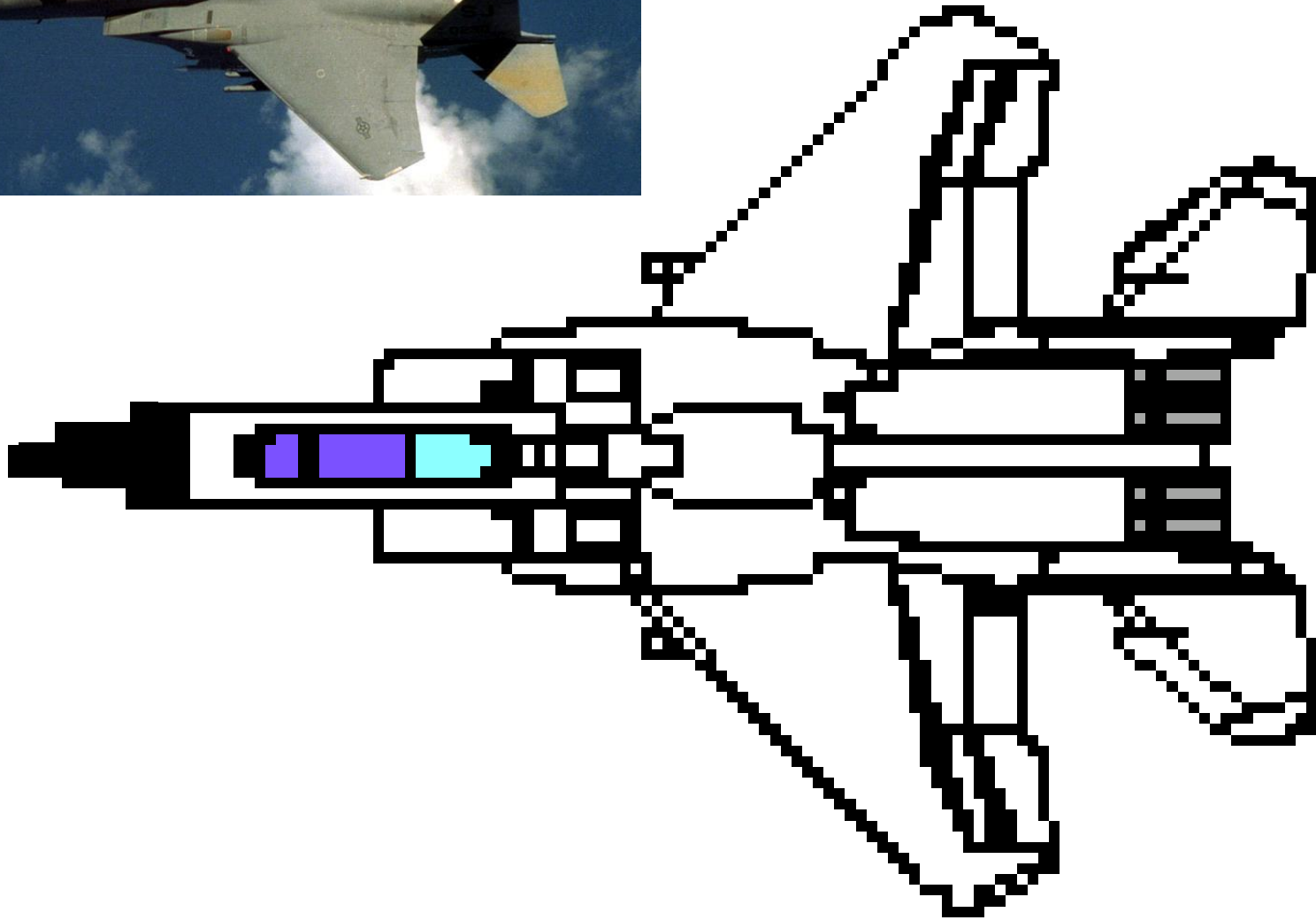
The power, which reflects the RCS is given by

$$\sigma = \zeta^2 = \left| \sum_{i=1}^N \sqrt{\sigma_i} \exp(-j4\pi R_i/\lambda) \right|^2$$

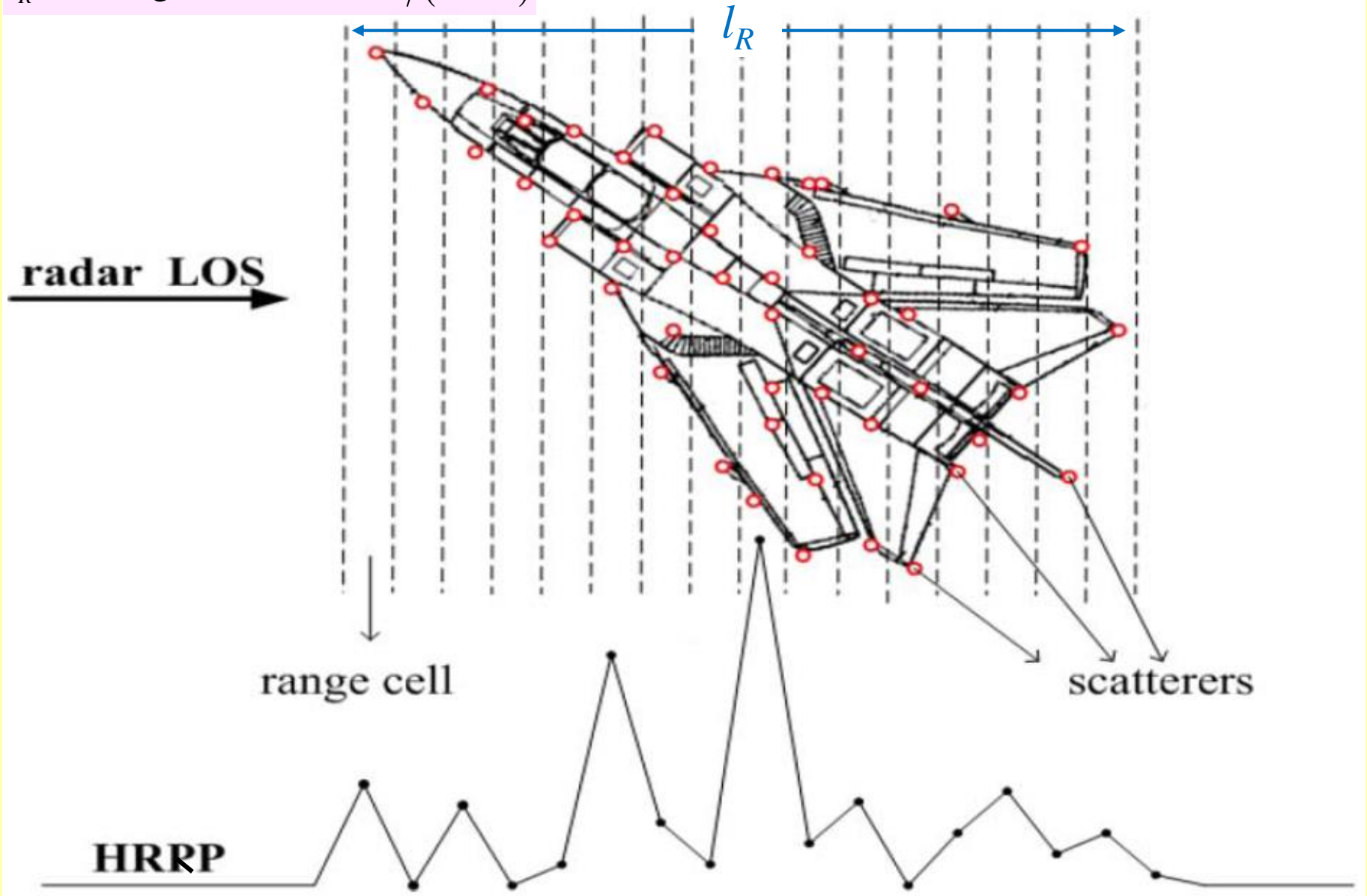
RCS of Multi-Scatterer Target

- RCS is proportional to $|\bar{y}(t)|^2$
 - Define the “voltage” $\zeta \equiv |\bar{y}| = \left| \sum_{i=1}^N \sqrt{\sigma_i} e^{-j4\pi R_i/\lambda} \right|$
- and “power” (RCS) $\sigma = \zeta^2 = \left| \sum_{i=1}^N \sqrt{\sigma_i} e^{-j4\pi R_i/\lambda} \right|^2$

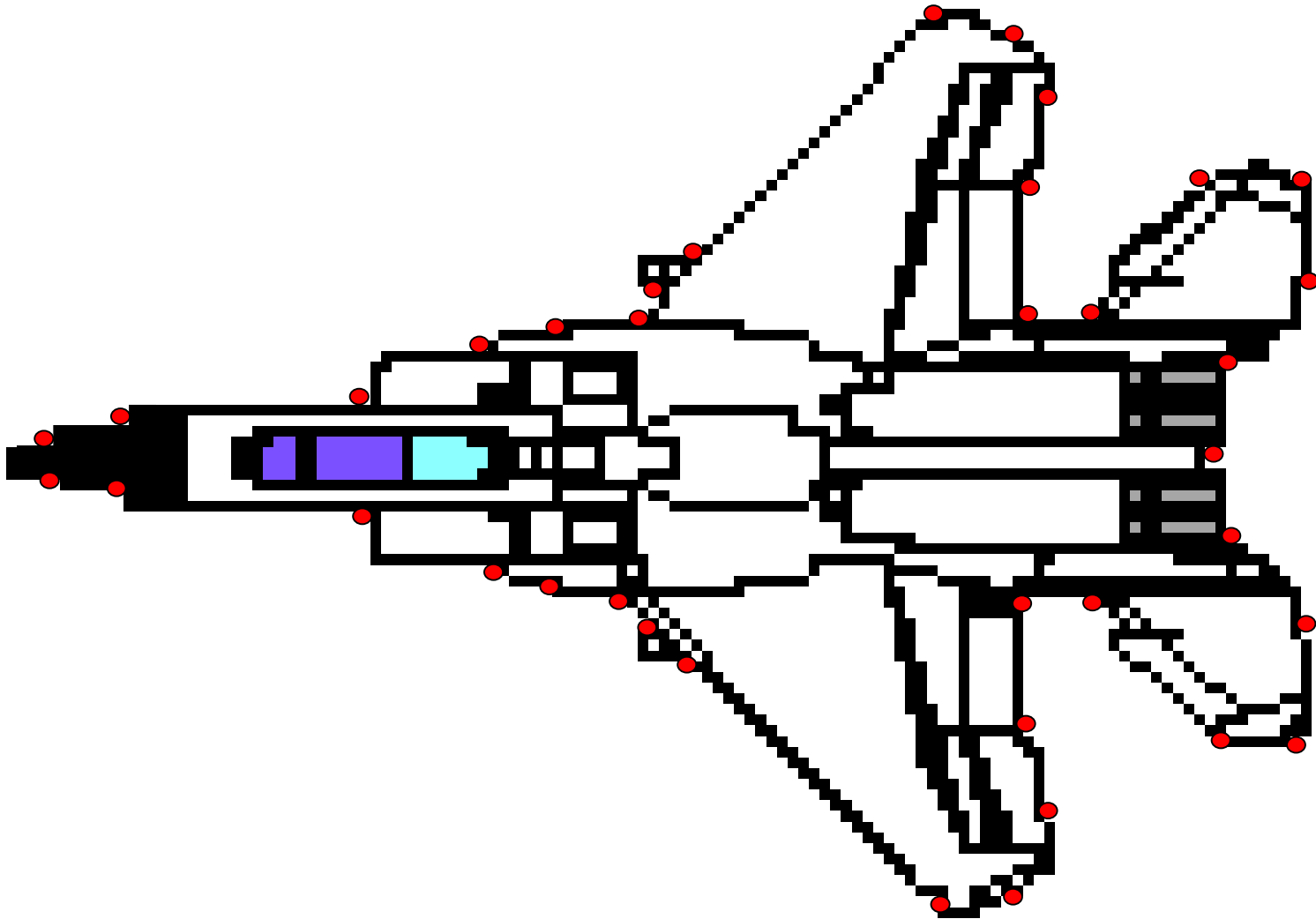
F-15

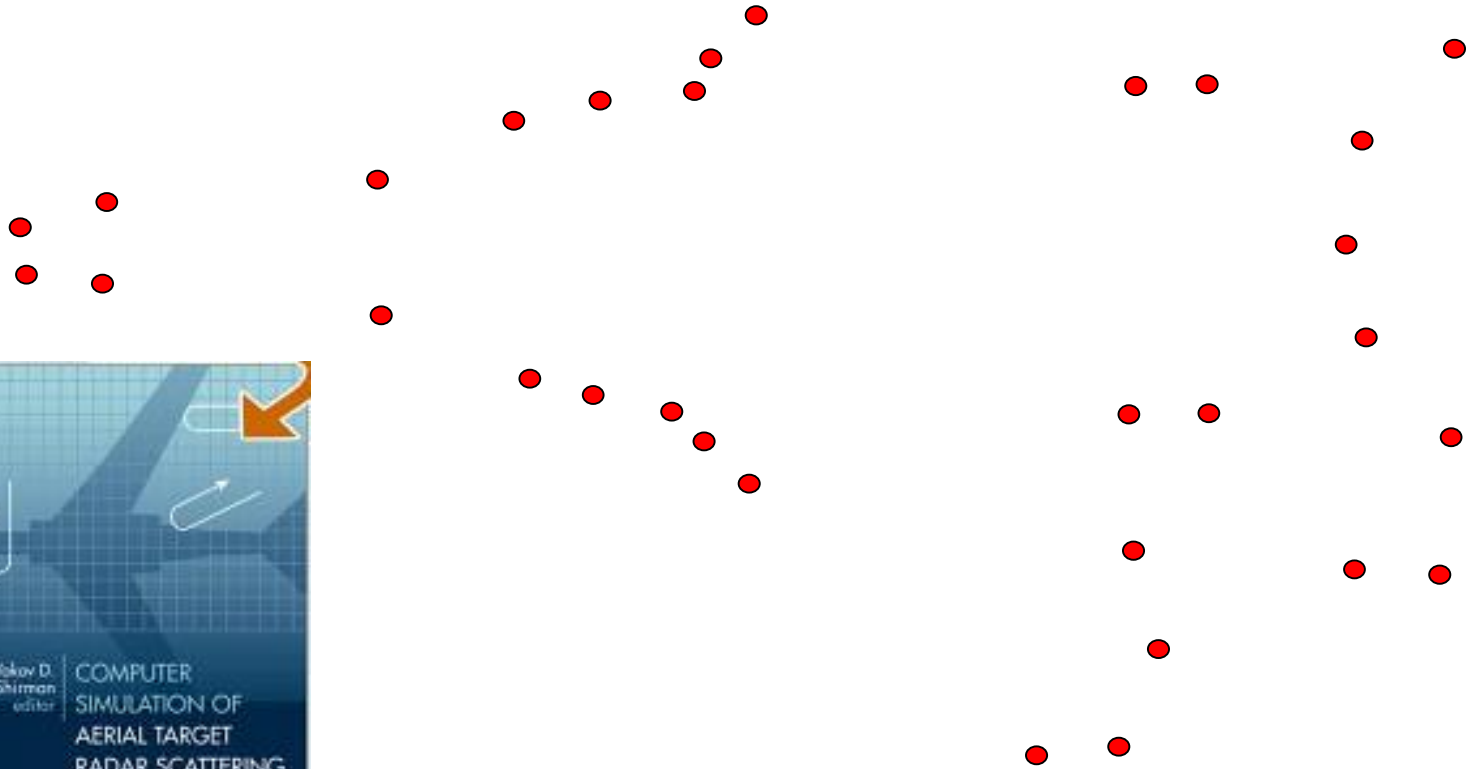
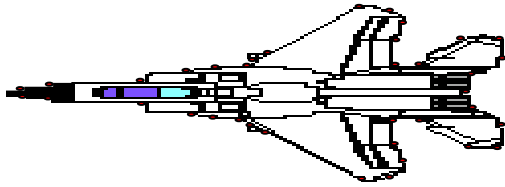


$l_R \gg$ Range resolution $\approx C/(2BW)$

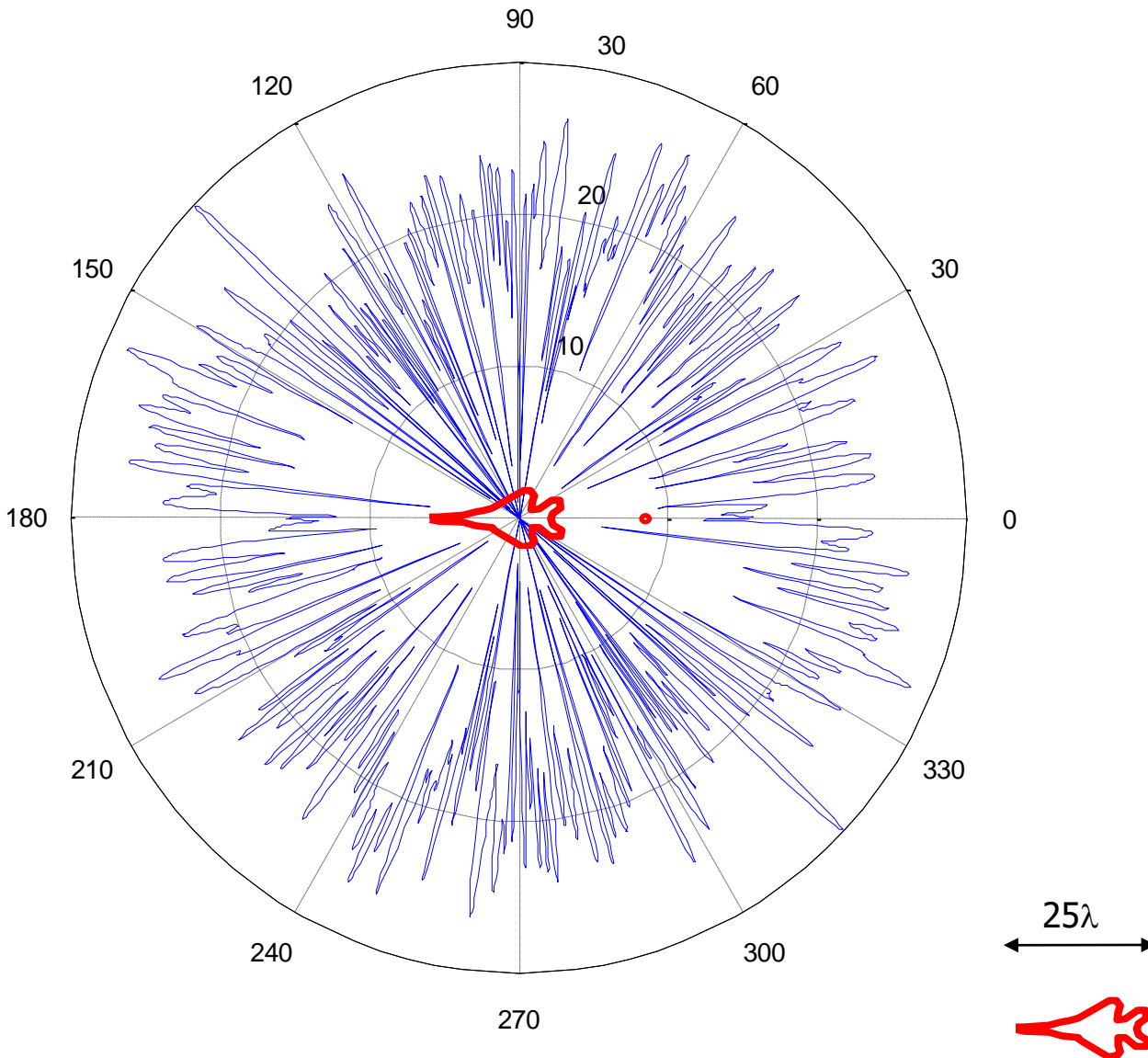


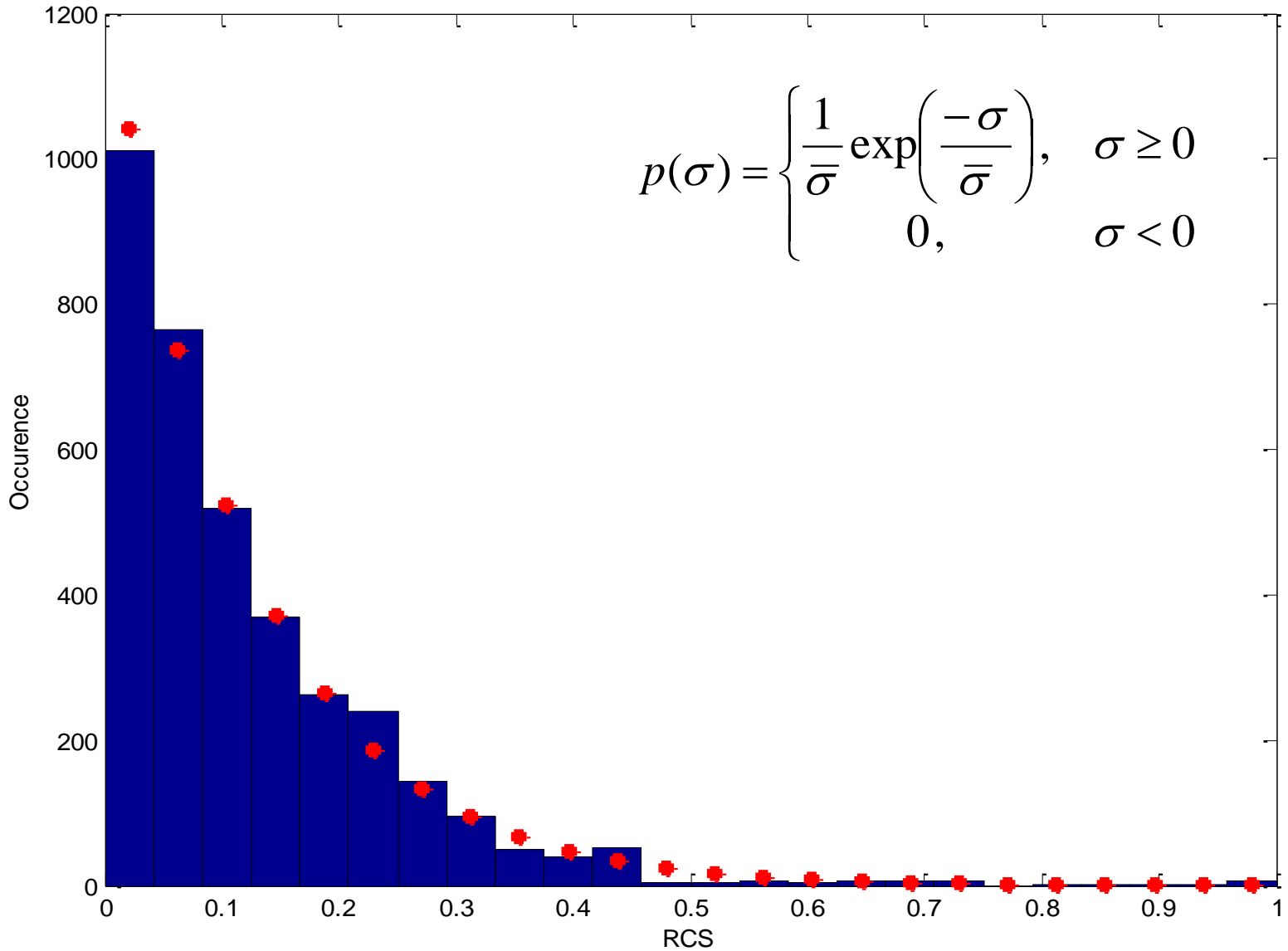
HRRP – High Resolution Range Profile





Assumptions (not realistic but simplify the model):
 Isotropical reflectors of identical σ , no shadowing, no multiple reflections





Exponential PDF (power) or Rayleigh PDF (voltage)

$$p(\sigma) = \begin{cases} \frac{1}{\bar{\sigma}} \exp\left(\frac{-\sigma}{\bar{\sigma}}\right), & \sigma \geq 0 \\ 0, & \sigma < 0 \end{cases}$$

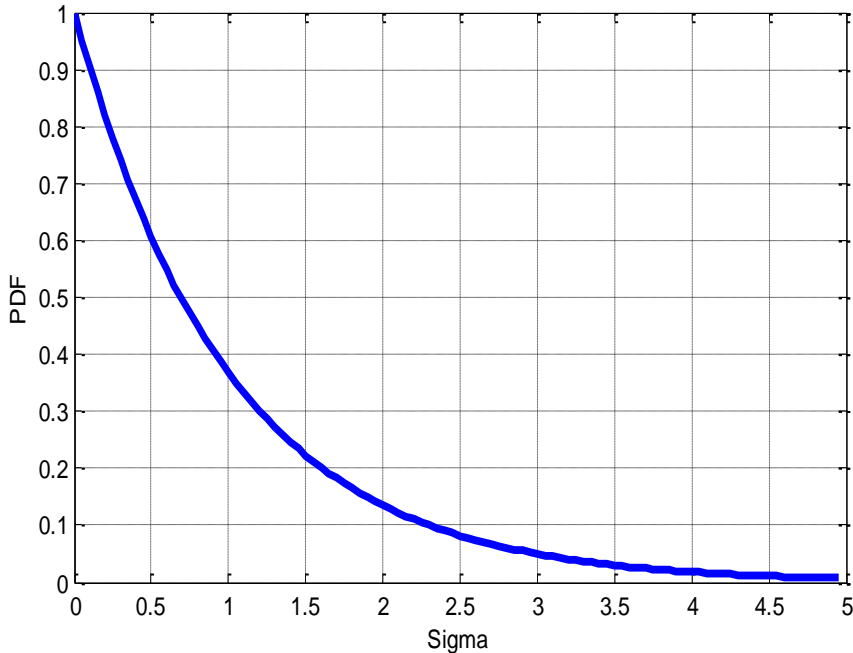
$\sigma \propto \text{RCS} \propto \text{power}$
 $A \propto \text{amplitude} \propto \text{field}$

$$p(A) = \begin{cases} \frac{A}{A_0^2} \exp\left(\frac{-A^2}{2A_0^2}\right), & A \geq 0 \\ 0, & A < 0 \end{cases}$$

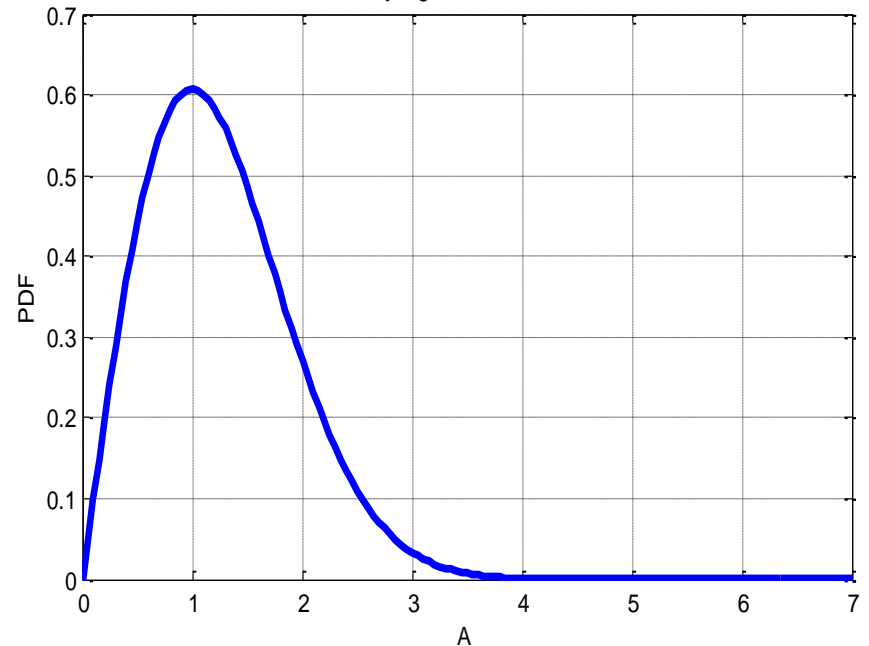
$$\sigma = \frac{A^2}{2}, \quad A_0^2 = \bar{\sigma}$$

$$\bar{A} = A_0 \sqrt{\frac{\pi}{2}}, \quad \overline{A^2} = 2A_0^2, \quad A_M = A_0 \sqrt{\ln 4}, \quad \text{Var}\{A\} = A_0^2 (2 - \pi/2)$$

Exponential PDF, avg. Sigma = 1



Rayleigh PDF, Ao = 1



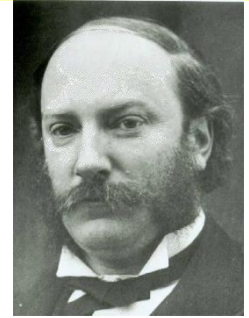
Why Rayleigh ?

Signal reflected from M scatterers of similar size

The range differences between the scatterers $\gg \lambda \Rightarrow \phi_k \gg 2\pi$

The modulo 2π nature of phase

$\therefore \phi_k$ uniformly distributed between 0 and 2π



The Rayleigh PDF was not developed by John William Strutt (Lord Rayleigh III) but was named in his honor.

$$y(t) = \text{Re} \left\{ \left[\exp(j\omega_c t + j\phi_0) \right] \sum_{k=1}^M a_k \exp(j\phi_k) \right\}$$

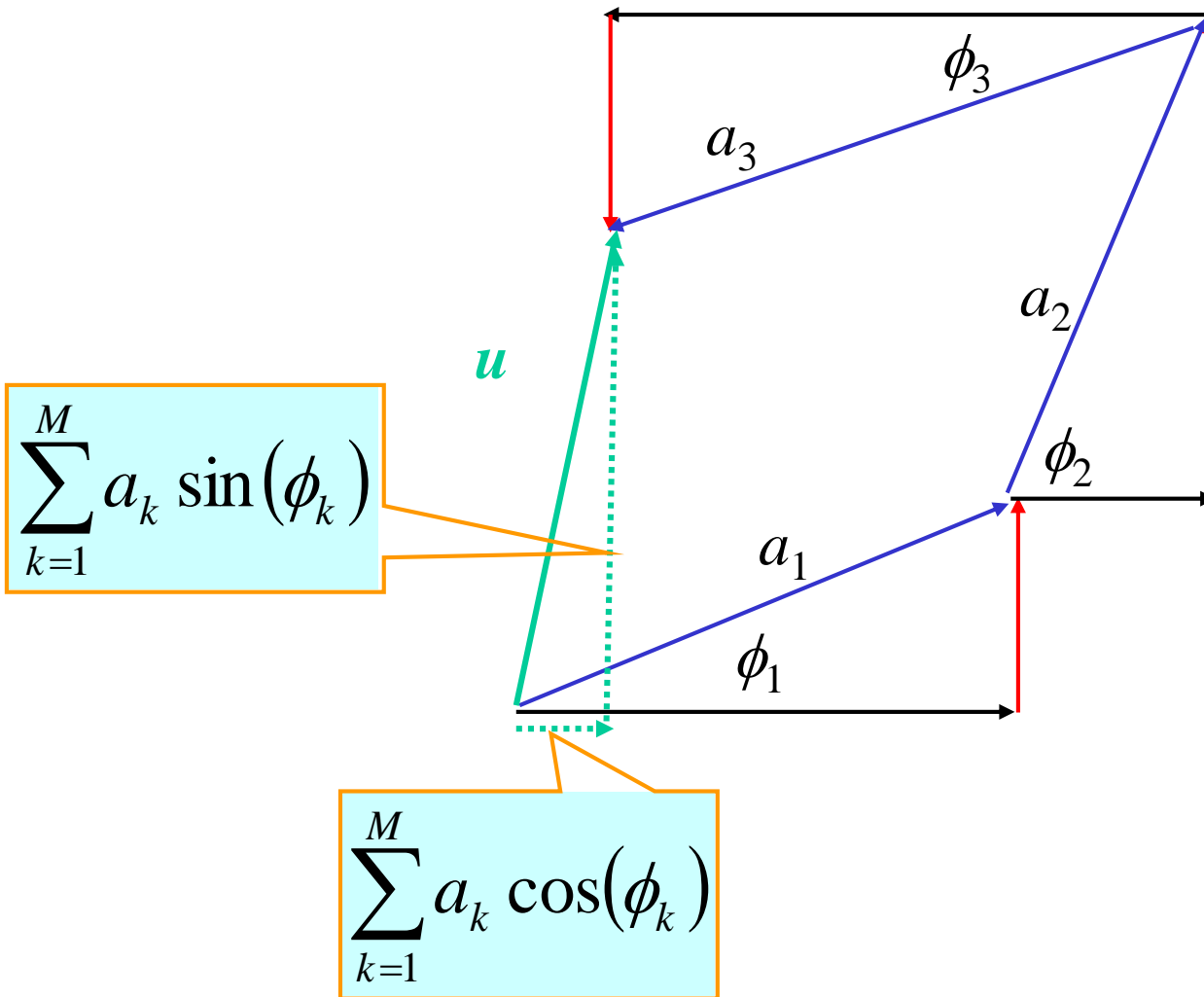
The received signal

$$\sum_{k=1}^M a_k \exp(j\phi_k) = u = r \exp(j\theta)$$

The complex envelope of the signal

$$u = \sum_{k=1}^M a_k \cos(\phi_k) + j \sum_{k=1}^M a_k \sin(\phi_k)$$

$$u = \sum_{k=1}^M a_k \cos(\phi_k) + j \sum_{k=1}^M a_k \sin(\phi_k)$$



$$\frac{u}{a} = \sum_{k=1}^M \cos(\phi_k) + j \sum_{k=1}^M \sin(\phi_k) = X + jY$$

To simplify the analysis assume
 $a_k = a$

$$p(\phi_k) = \begin{cases} \frac{1}{2\pi} & 0 \leq \phi_k \leq 2\pi \\ 0 & \text{elsewhere} \end{cases} \Rightarrow \overbrace{\sin(\phi_k)}^{E(Y)} = \overbrace{\cos(\phi_k)}^{E(X)} = 0$$

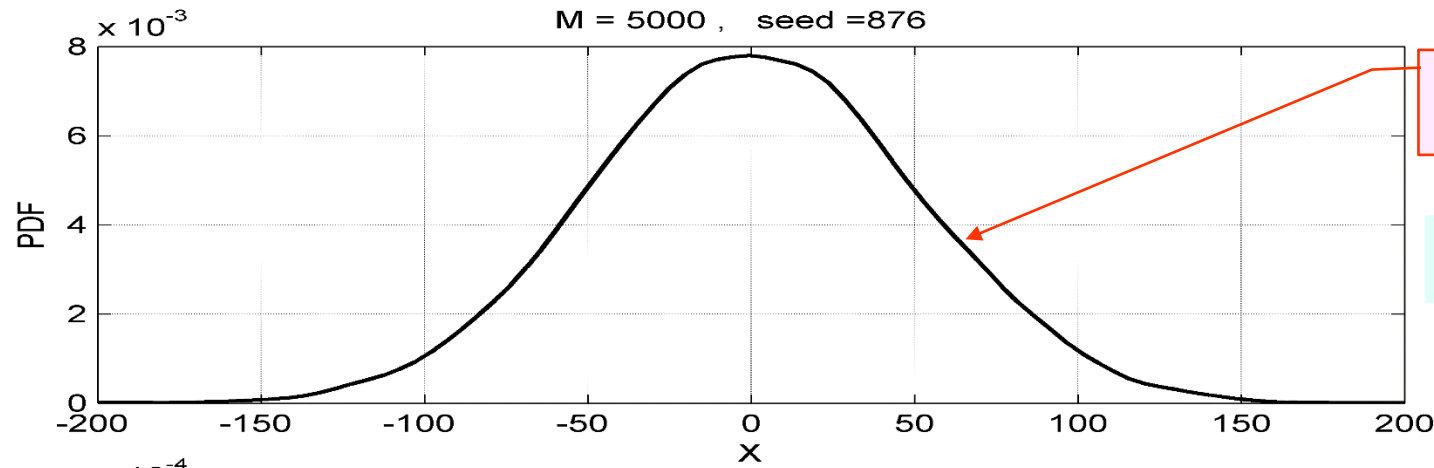
For $M \gg 1$, the central limit theory implies that X and Y are **Gaussian** distributed with zero mean and variance equal to M times the variance of $\cos(\phi_k)$

$$\text{Var } Y = \text{Var } X = M \int_0^{2\pi} \frac{1}{2\pi} \cos^2 \phi \, d\phi = \frac{M}{2}$$

Next slide

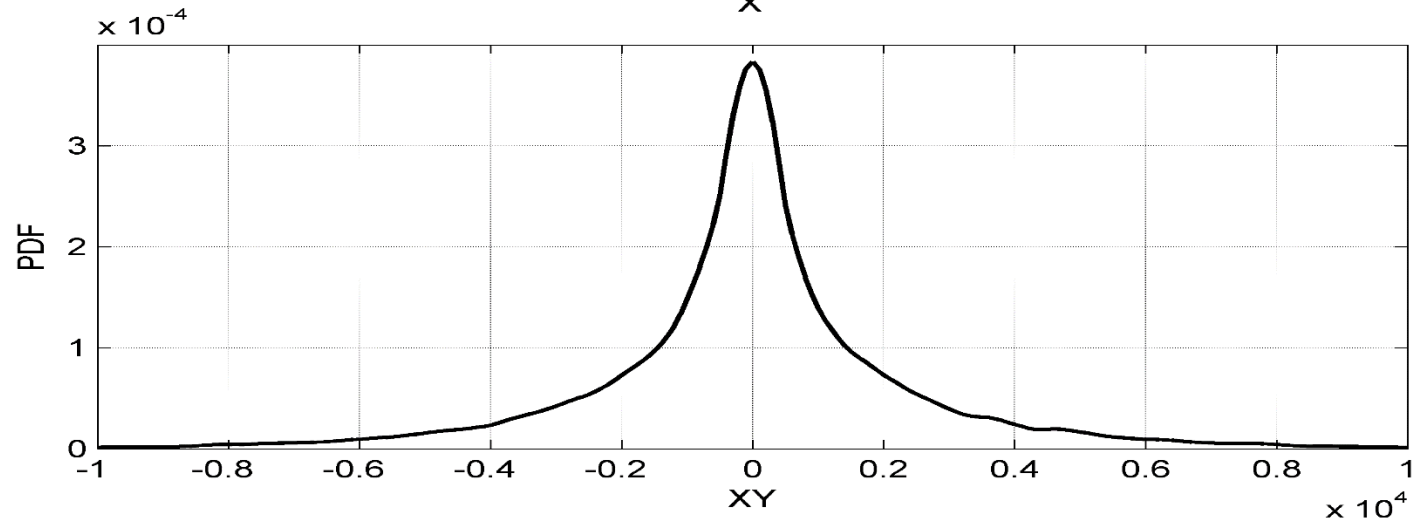
$$E\{XY\} = 0 = E\{X\}E\{Y\} \Rightarrow X \text{ and } Y \text{ uncorrelated}$$

X and Y uncorrelated and Gaussian \Rightarrow **independent**



Gaussian

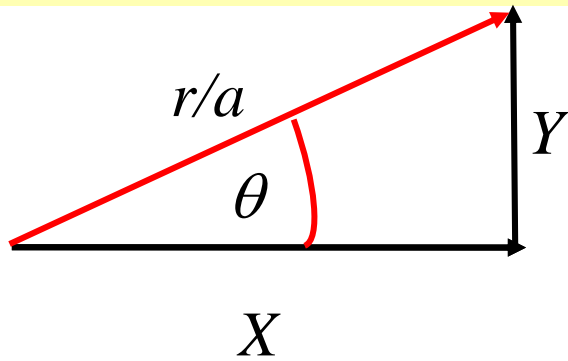
$$E\{X\} = \overline{\cos(\varphi_k)} = 0$$



$$E\{XY\} = 0$$

$$E\{XY\} = 0 = E\{X\}E\{Y\} \Rightarrow X \text{ and } Y \text{ uncorrelated}$$

X and Y uncorrelated and Gaussian \Rightarrow independent



$$\left(\frac{r}{a}\right)^2 = X^2 + Y^2, \quad \theta = \tan^{-1}\left(\frac{Y}{X}\right)$$

$$\text{Var } Y = \text{Var } X = \frac{M}{2}$$

$$\frac{M}{2} a^2 = \text{Var}(Xa)$$

$$p(X, Y) = p(X)p(Y) = \frac{1}{M\pi} \exp\left[\frac{-(X^2 + Y^2)}{M}\right]$$

$$p(r) = \frac{2r}{Ma^2} \exp\left(\frac{-r^2}{Ma^2}\right)$$

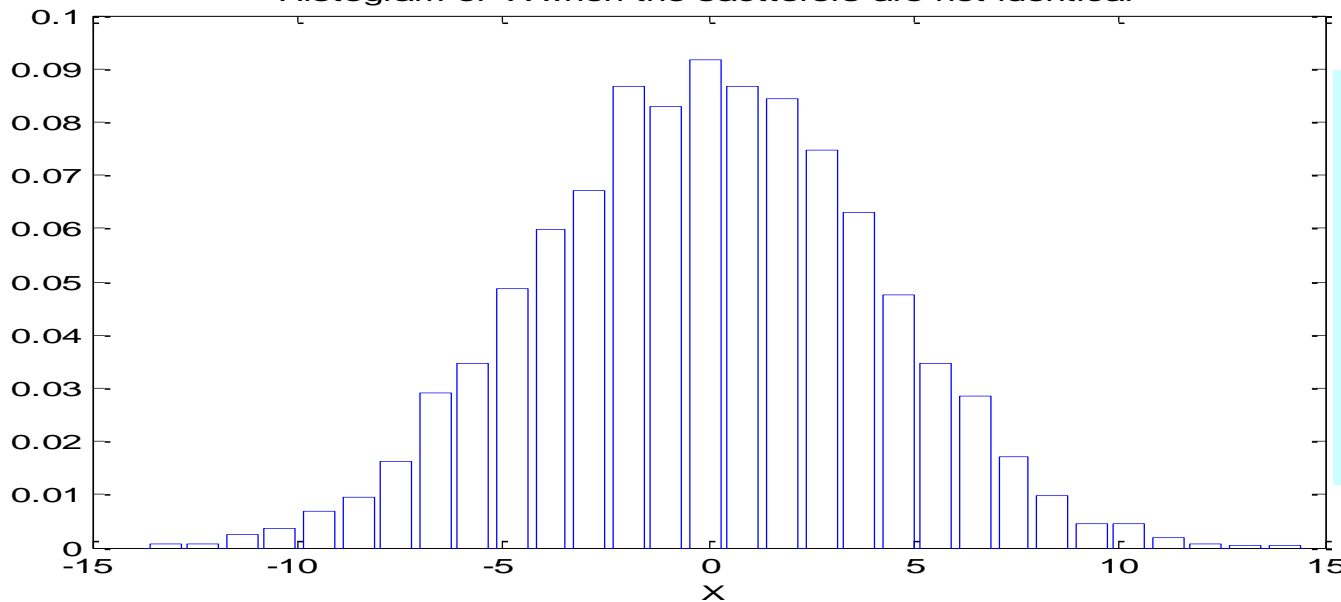
$$p(\theta) = \begin{cases} \frac{1}{2\pi} & 0 \leq \theta \leq 2\pi \\ 0 & \text{elsewhere} \end{cases}$$

$$A = r\sqrt{2}, \quad A_0^2 = Ma^2$$

$$p(A) = \frac{A}{A_0^2} \exp\left(\frac{-A^2}{2A_0^2}\right), \quad 0 \leq A$$

With the assumption that $a_k = a$, we got Rayleigh PDF of A . What happens when this assumption does not hold?

Histogram of X when the scatterers are not identical



$$p(a_k) = 1, \quad 0 \leq a_k \leq 1$$

$$X = \sum_{k=1}^M a_k \cos(\varphi_k)$$

$$Y = \sum_{k=1}^M a_k \sin(\varphi_k)$$

$$r = \sqrt{X^2 + Y^2}$$

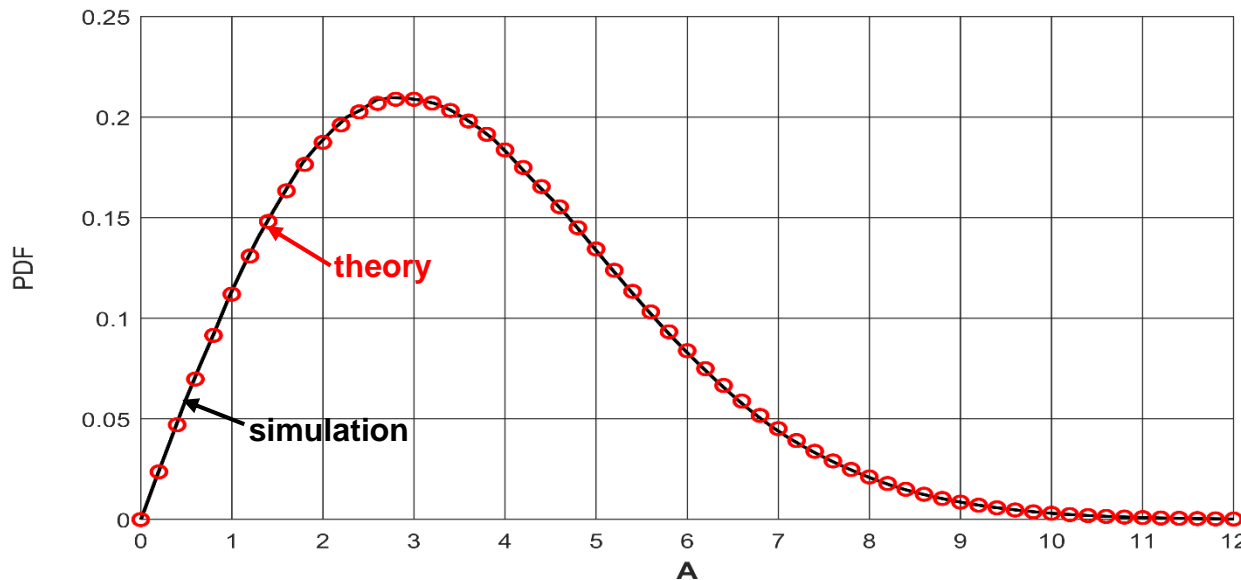
$M = 100$

mean $x = -0.0643$

mean $xy = 0.1860$

STD $x = 4.0739$

Var $x = 16.5966$



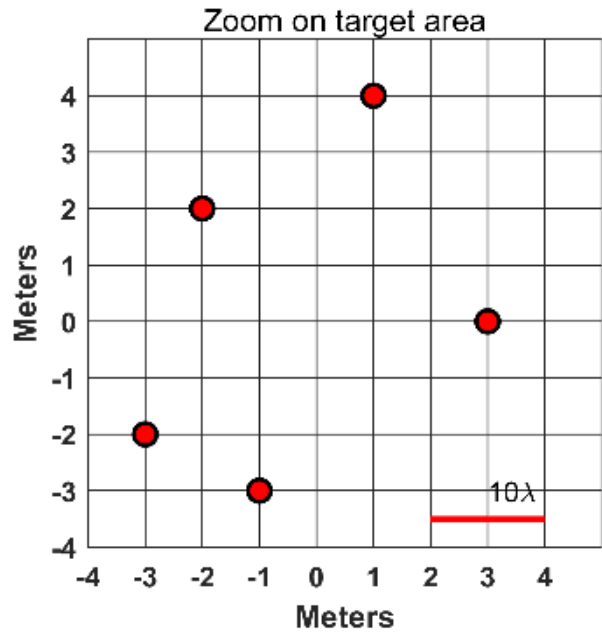
With the assumption above, we also got Rayleigh PDF of A (or r).

To plot the PDF of the simulation result, we used MATLAB's `ksdensity` function

COHERENT and NON-COHERENT RADAR

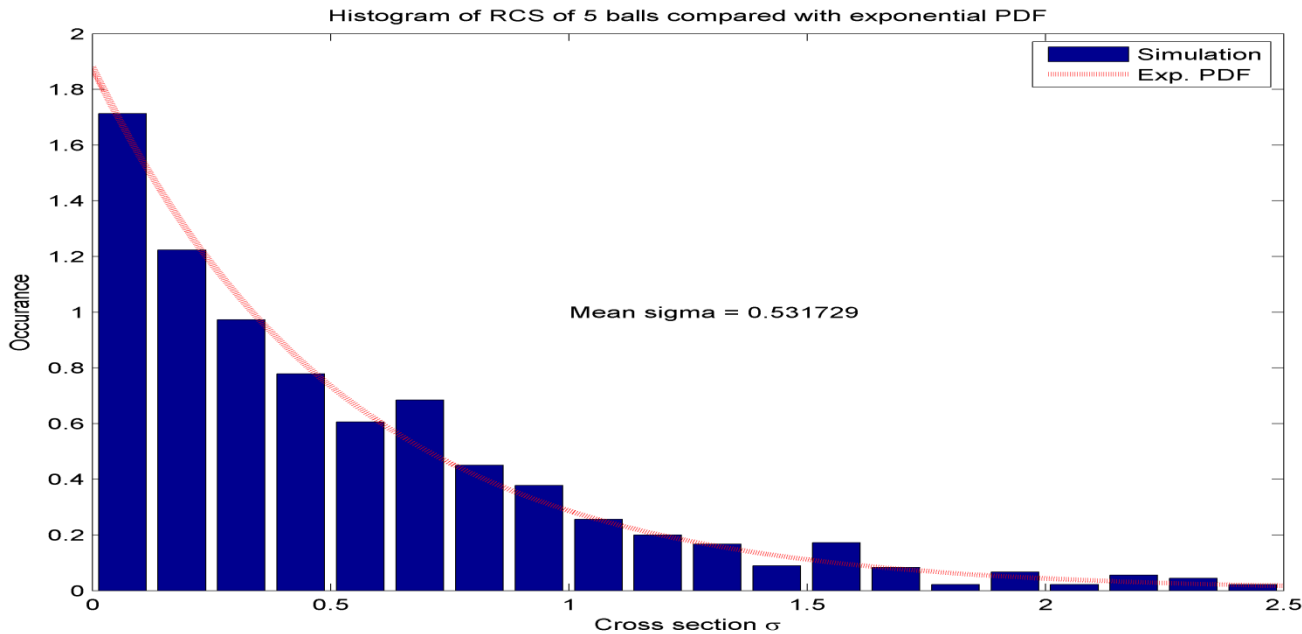
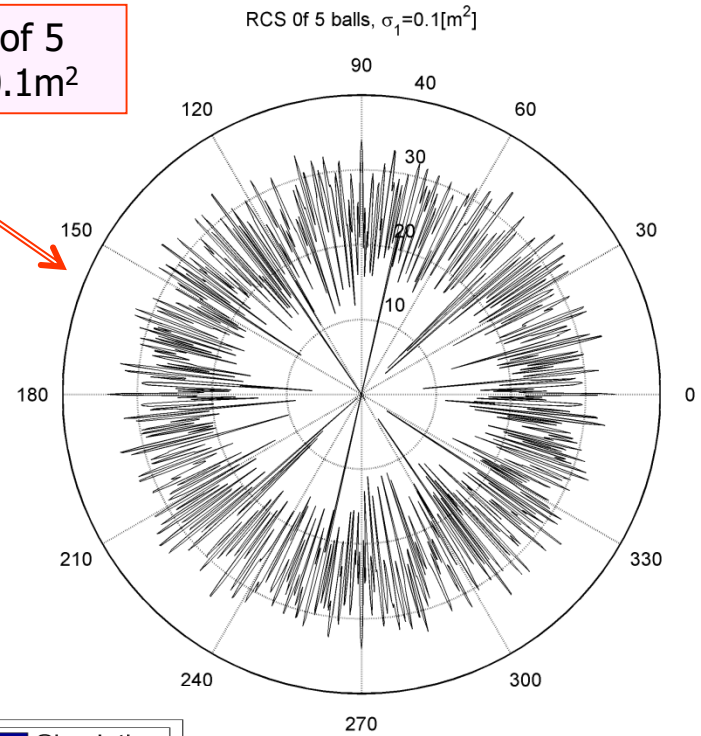
For a radar system that transmits a pulse train, to be called coherent, requires:

- The transmitted pulses must be coherent.
- The receiver needs to perform synchronous detection.
- **The target must maintain coherence for the duration of the pulse train.**



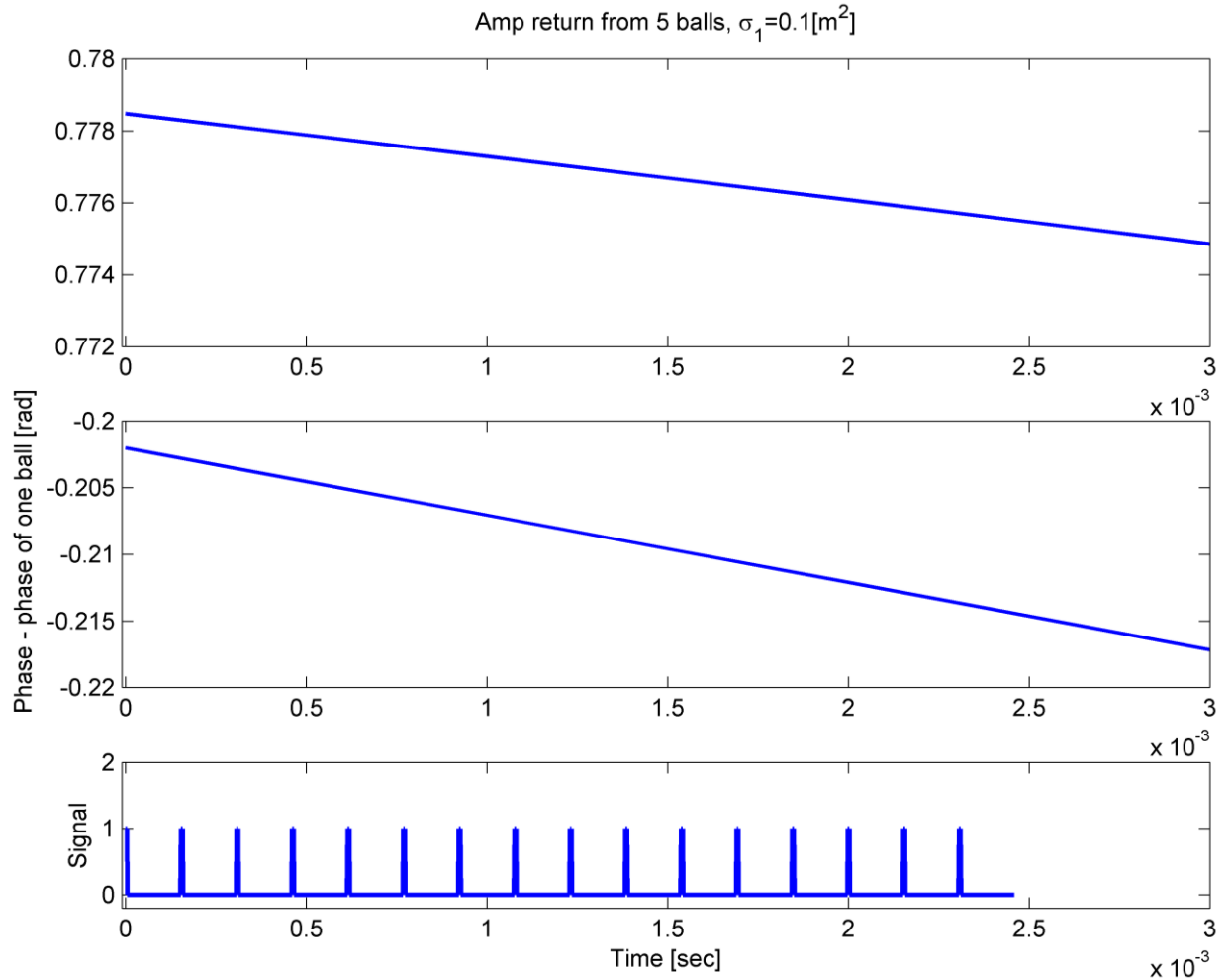
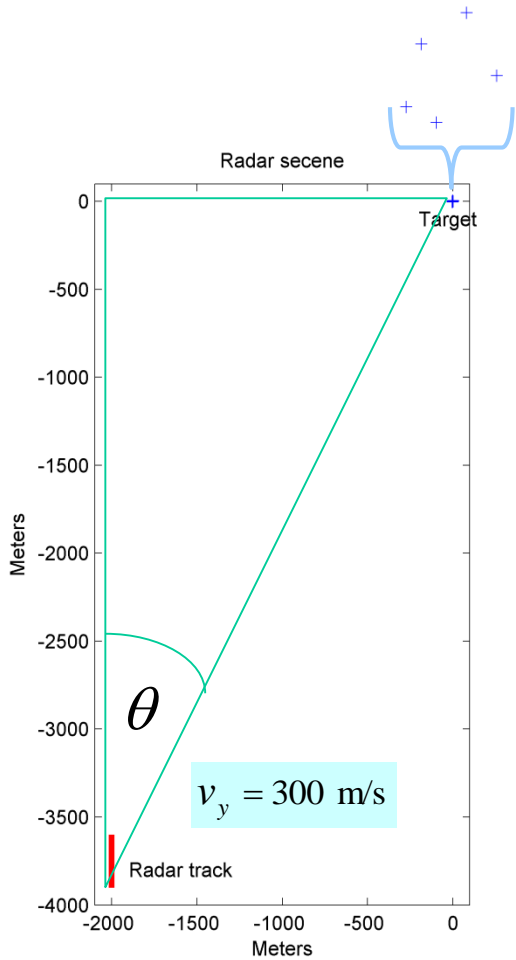
RCS angular dependence of 5 identical balls, each σ is 0.1m^2

If such a target is moving, hence changing aspect angle toward the radar, can it be a coherent target ?



$$p(\sigma) = \begin{cases} \frac{1}{\bar{\sigma}} \exp\left(-\frac{\sigma}{\bar{\sigma}}\right), & \sigma \geq 0 \\ 0, & \sigma < 0 \end{cases}$$

The target must maintain coherence for the duration of the pulse train.



Narrow bandpass signal

Natural envelope

$$s(t) = g(t) \cos[\omega_c t + \phi(t)]$$

In-phase component

Quadrature component

$$s(t) = g_c(t) \cos \omega_c t - g_s(t) \sin \omega_c t$$

$$g_c(t) = g(t) \cos \phi(t)$$

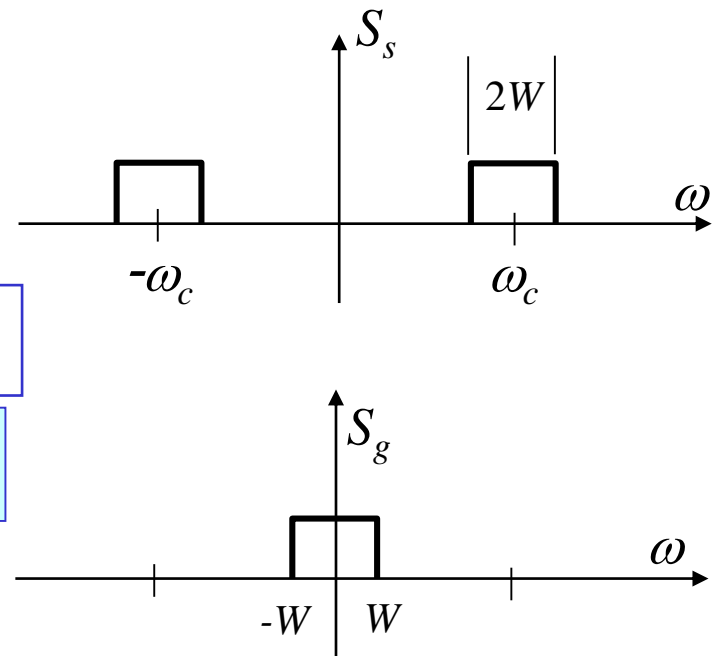
$$g_s(t) = g(t) \sin \phi(t)$$

$$u(t) = g_c(t) + jg_s(t) = \mathbf{I}(t) + j\mathbf{Q}(t)$$

Complex envelope of $s(t)$

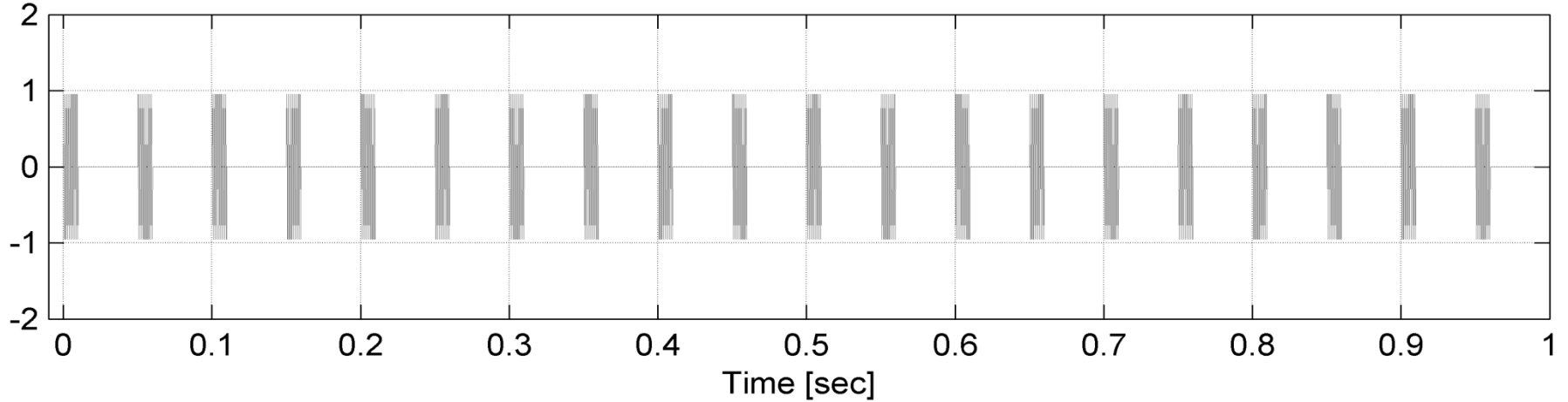
$$s(t) = \text{Re}\{u(t)e^{j\omega_c t}\}$$

$$g(t) = |u(t)|$$

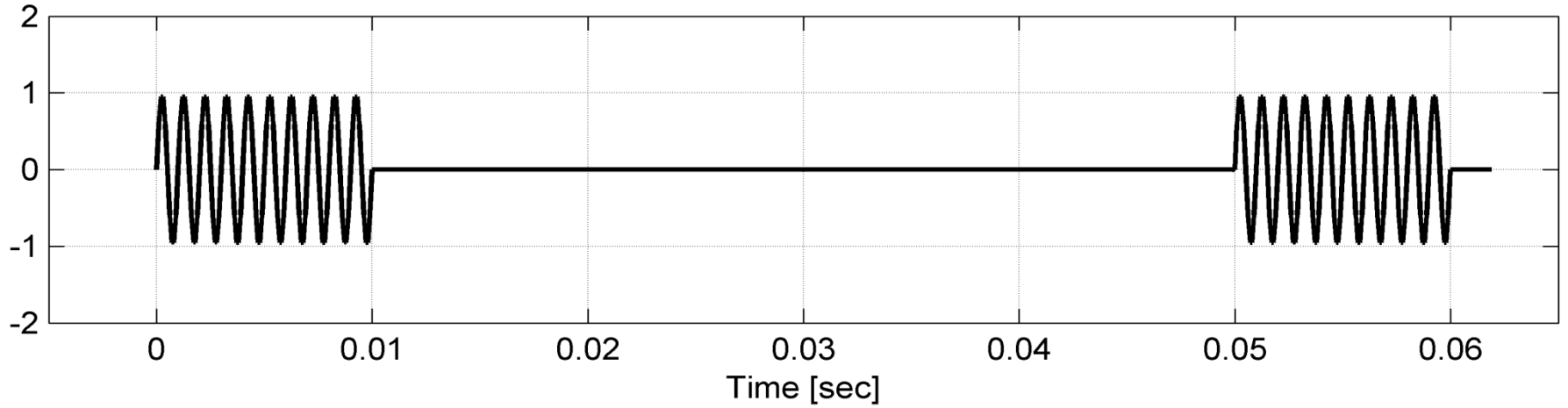


The transmitted pulses must be coherent.

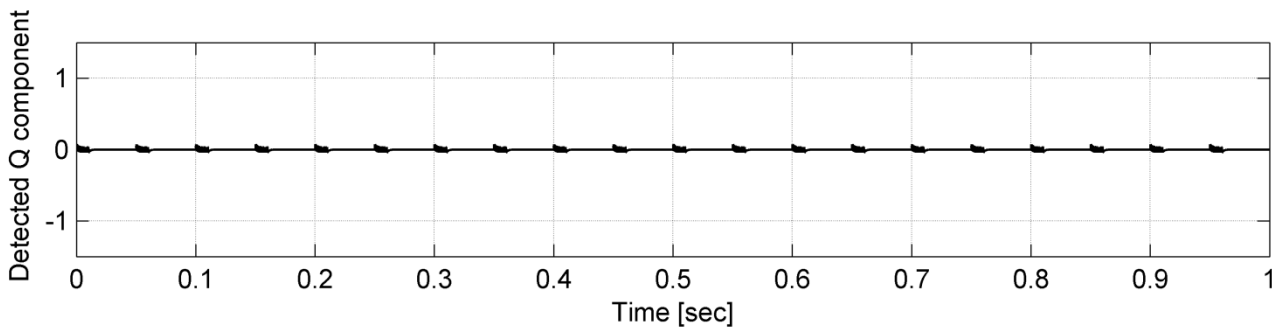
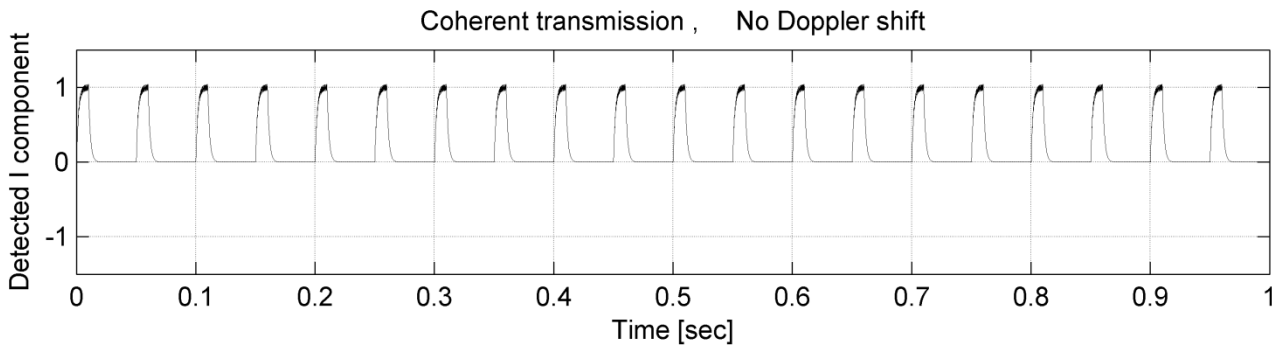
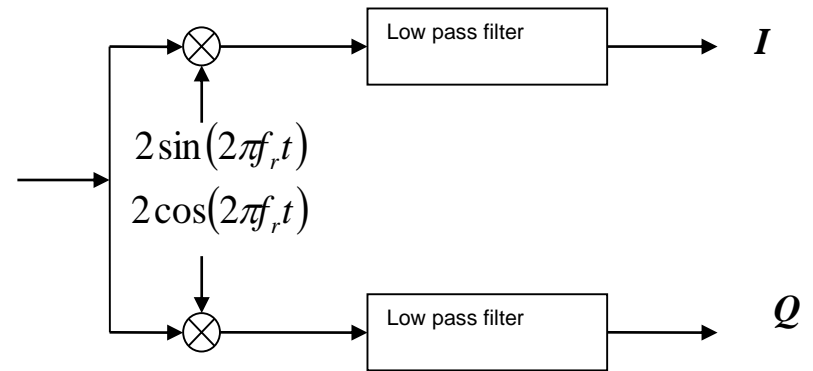
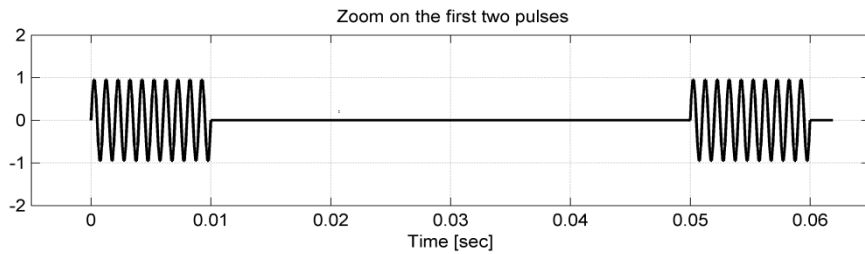
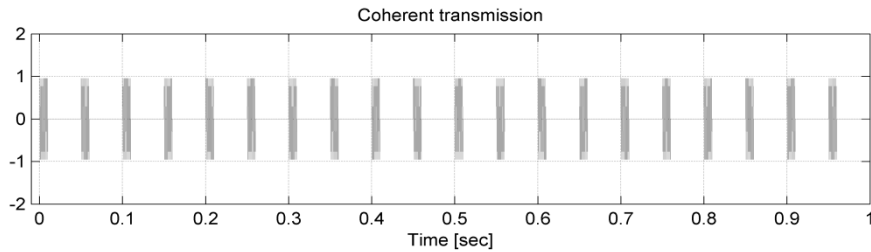
Coherent transmission

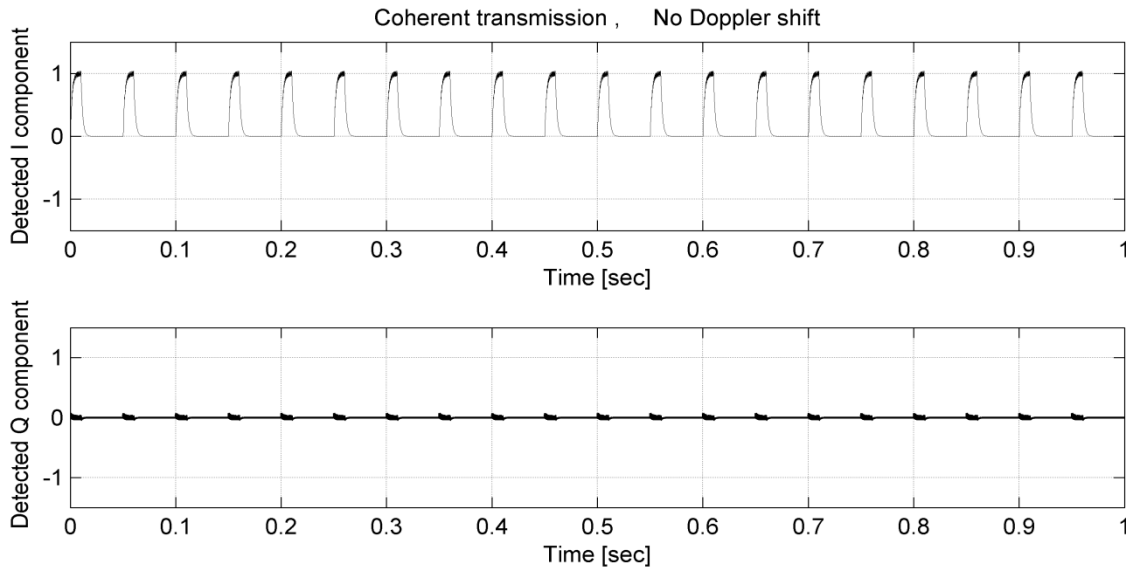


Zoom on the first two pulses

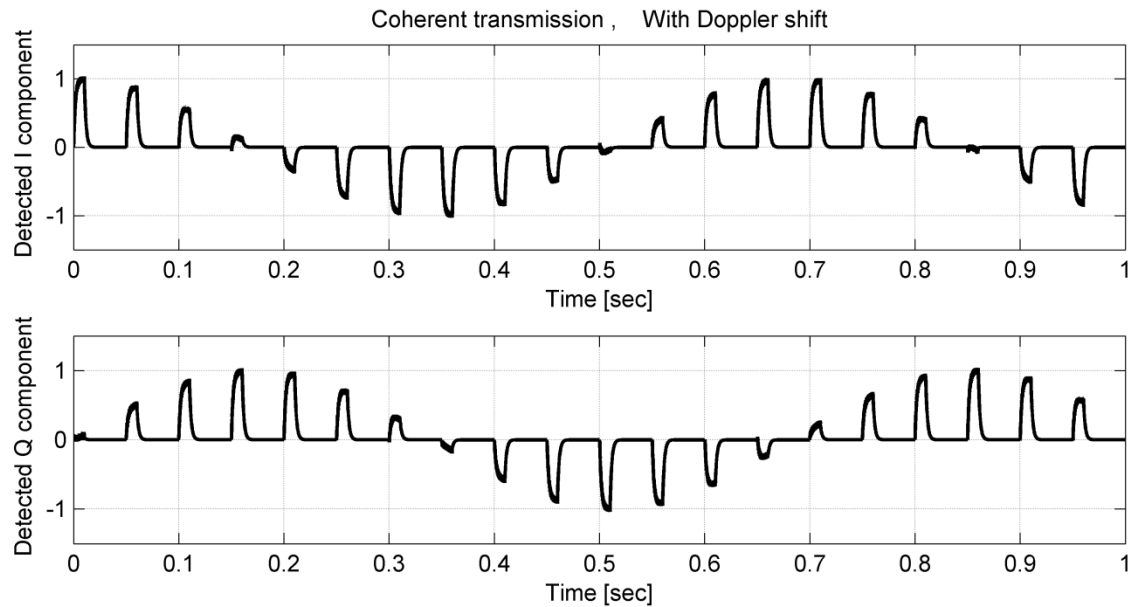


The receiver needs to perform synchronous detection.





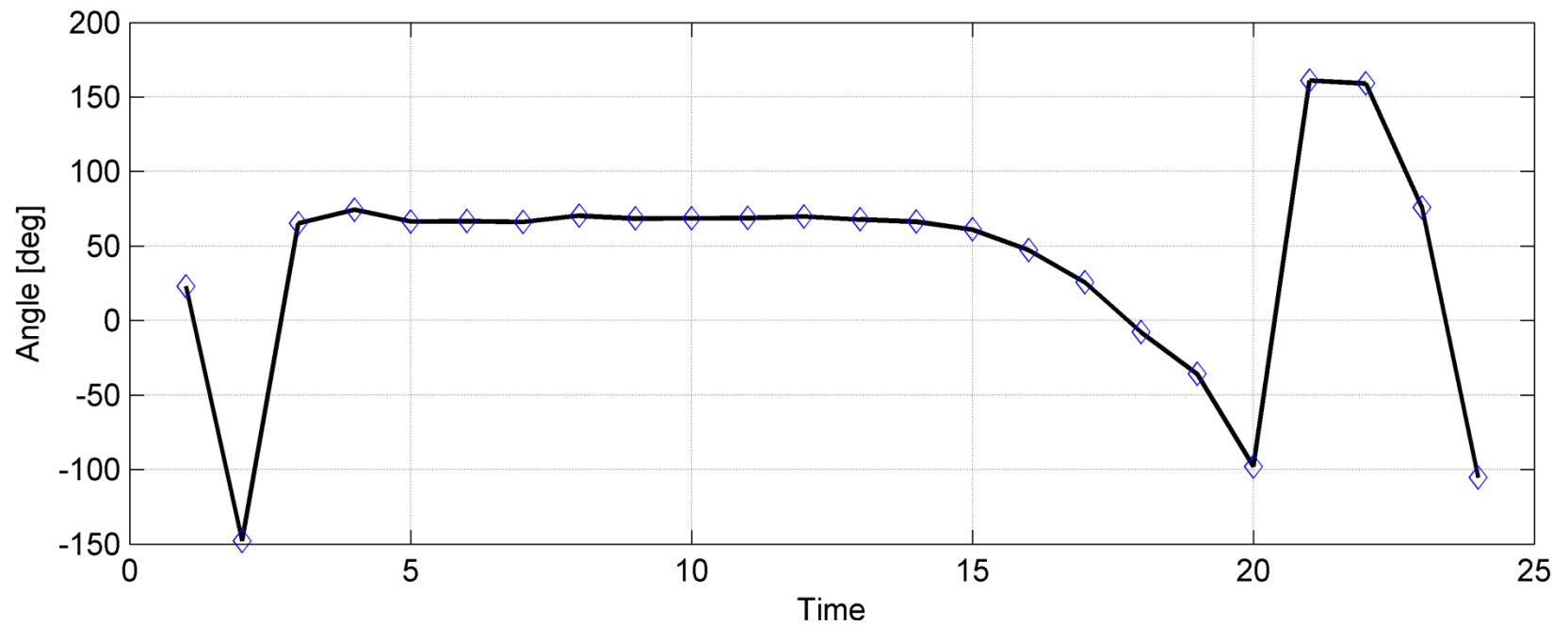
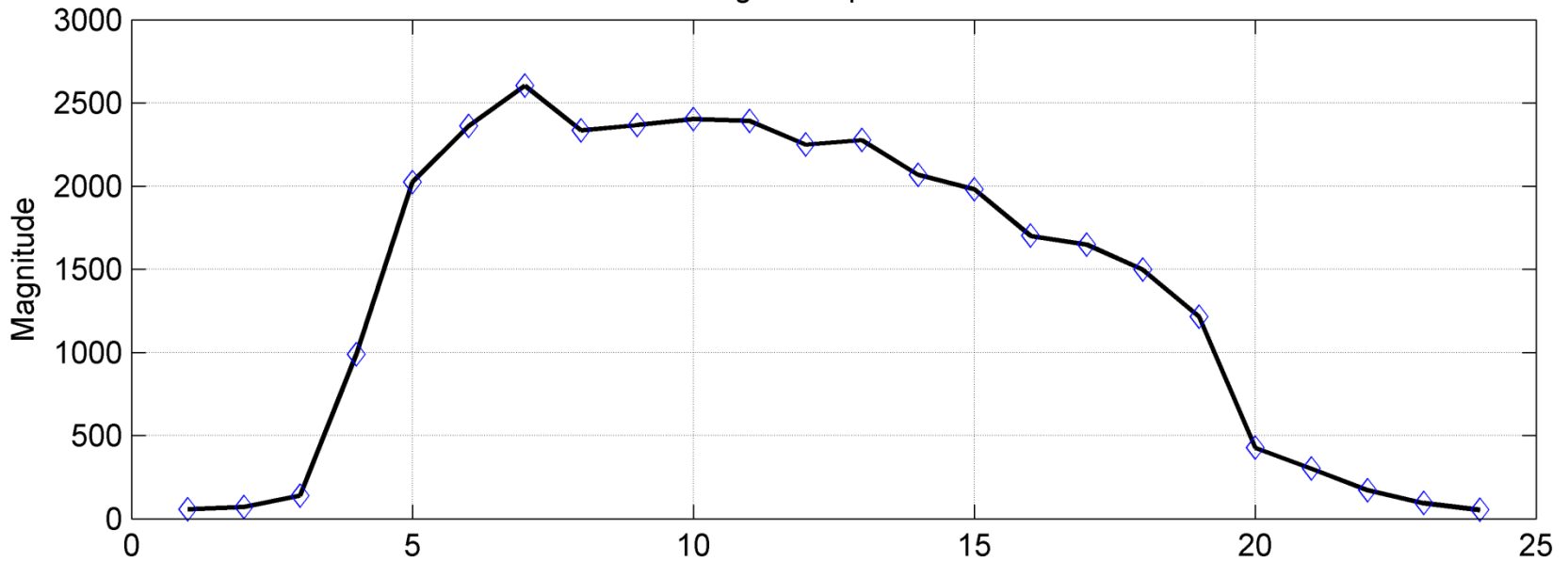
Coherent radar allows Doppler processing.



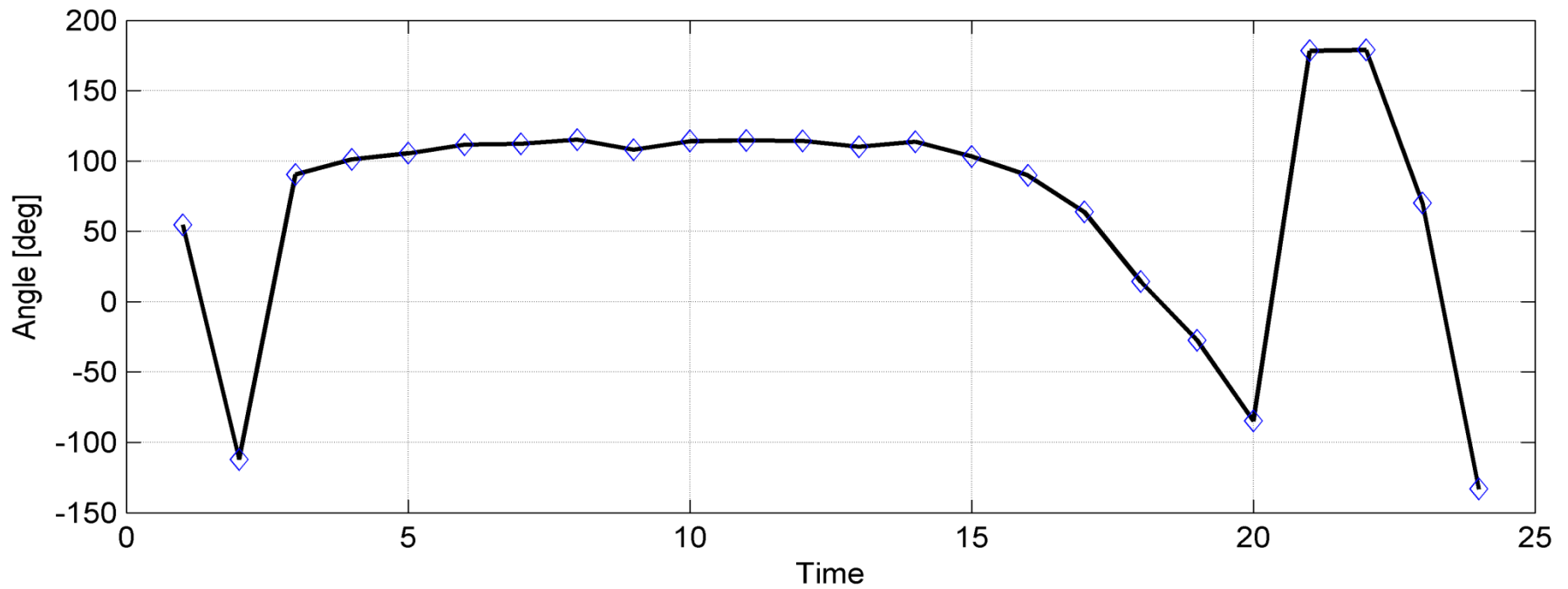
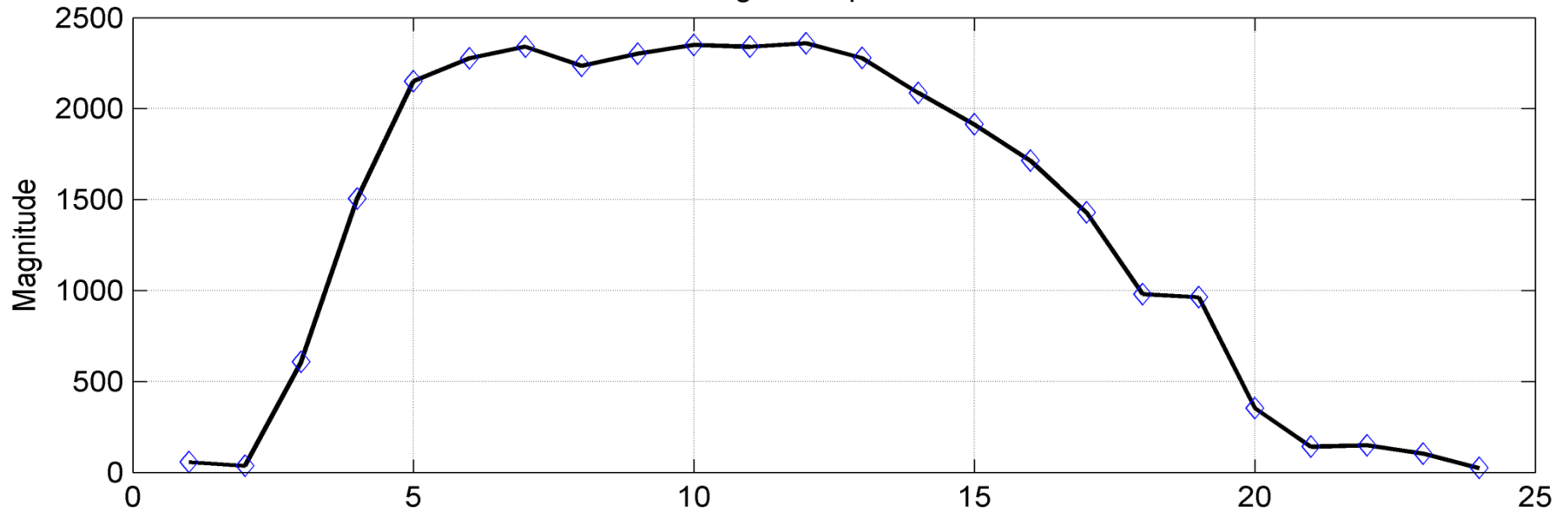
AMPLITUDE AND PHASE OF **MAGNETRON** PULSES

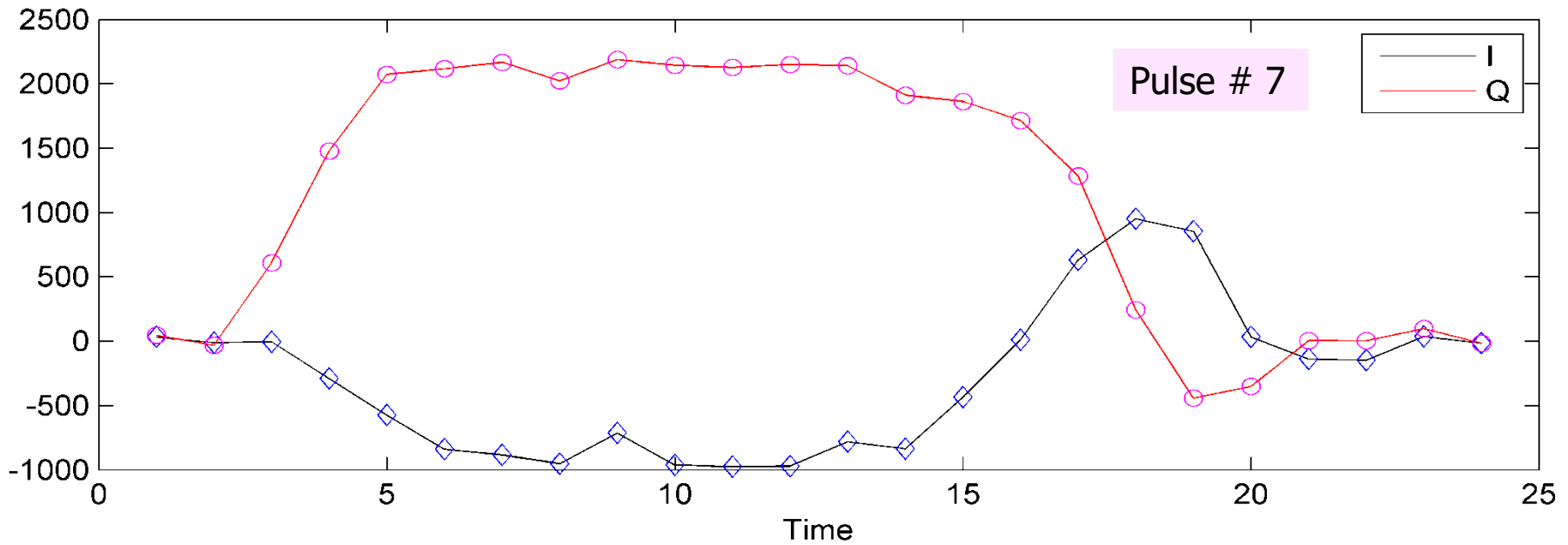
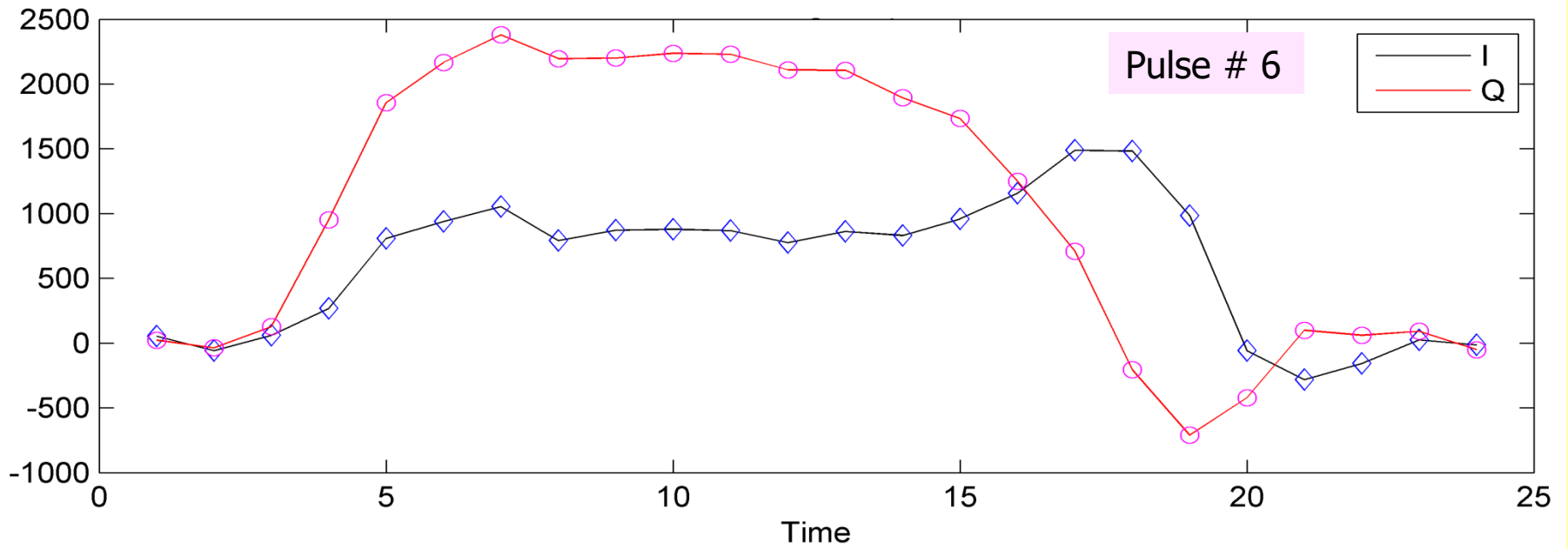


Magnetron pulse # 6

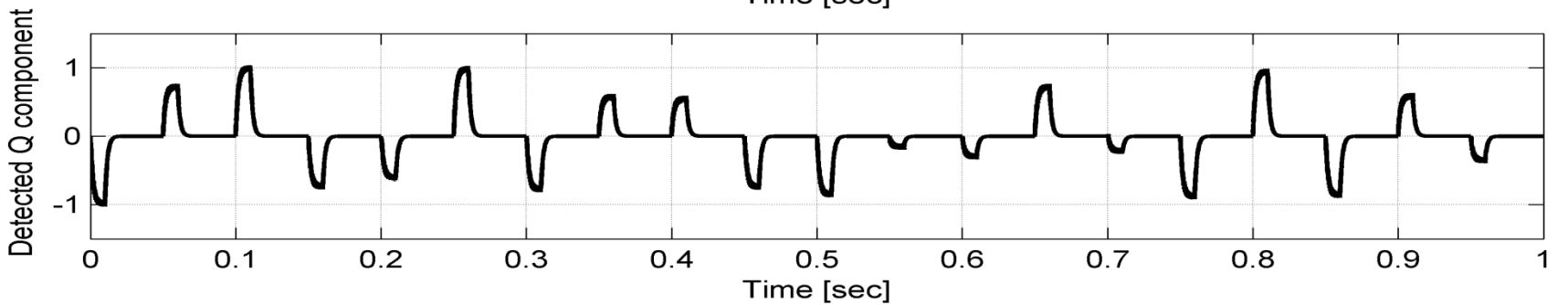
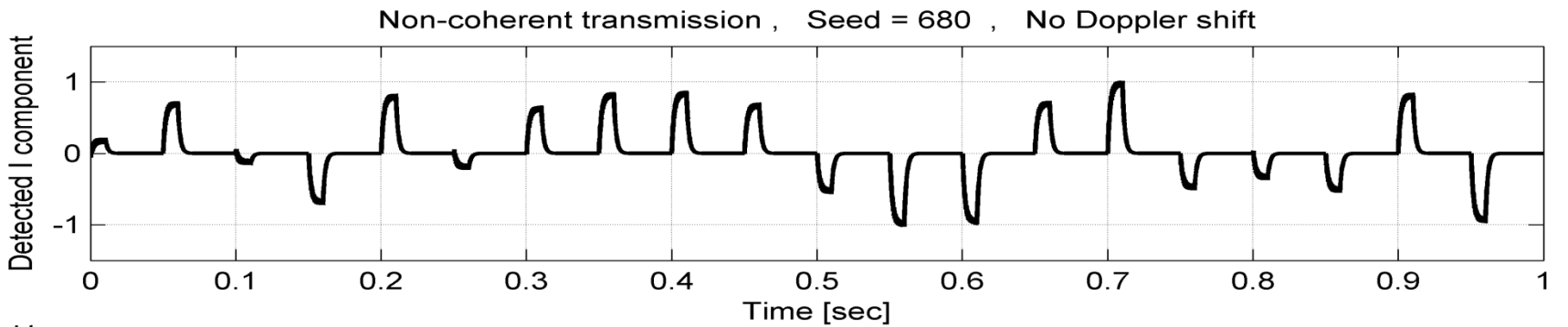
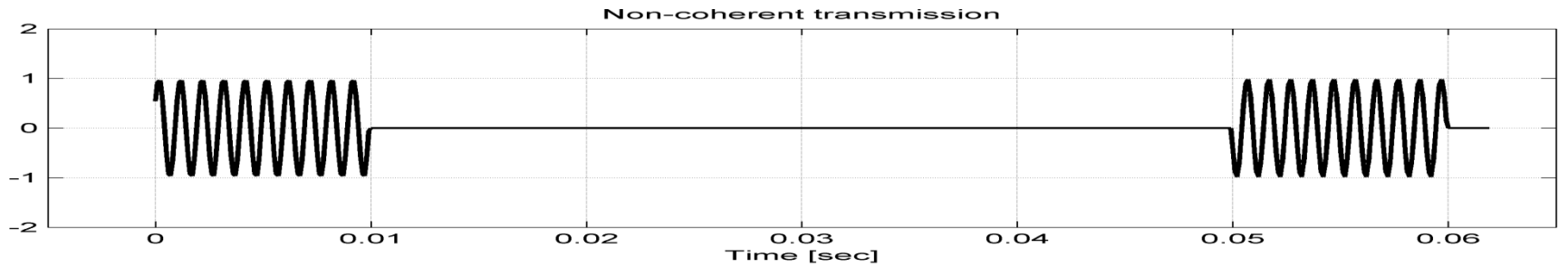
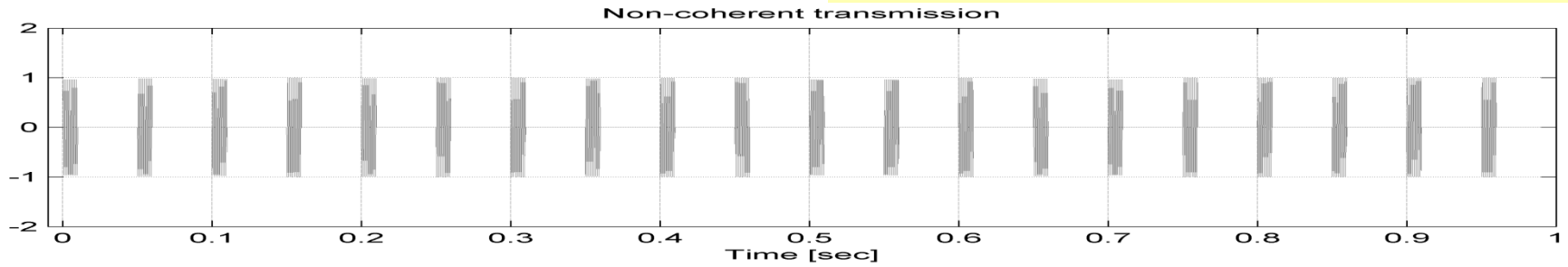


Magnetron pulse # 7

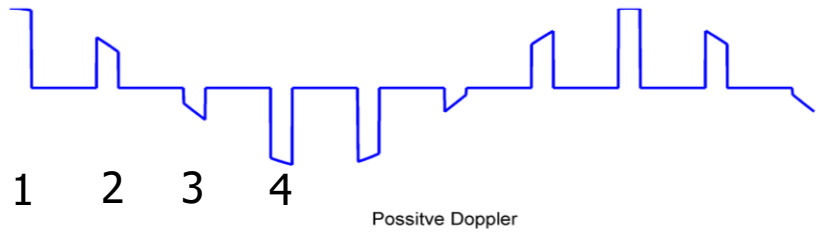
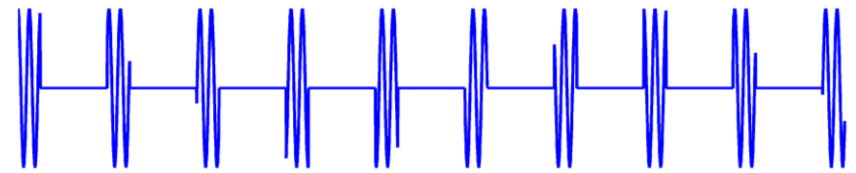
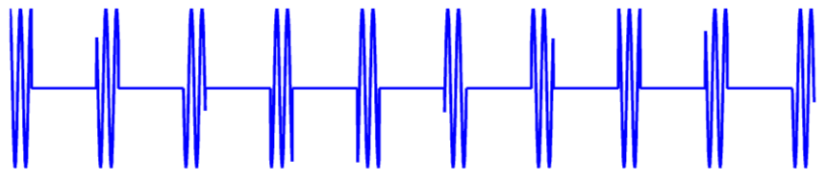




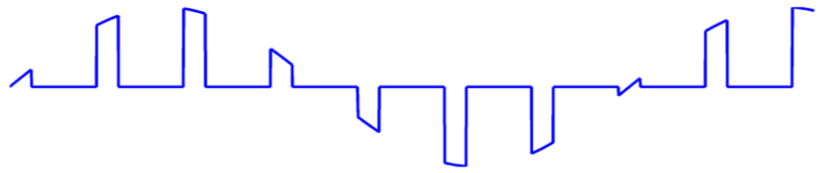
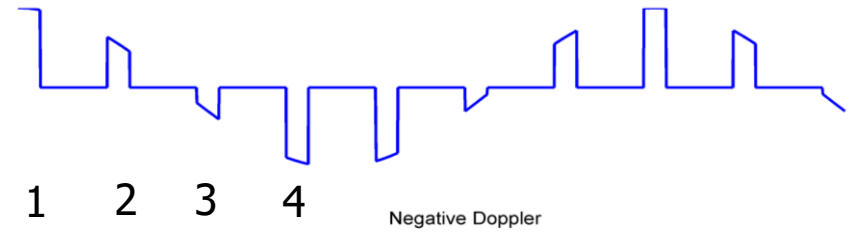
- Magnetron is a low-cost source of high-power microwave pulses
- Magnetron pulses are not coherent
- Magnetron-based radar cannot produce Doppler information and cannot separate moving targets from stationary clutter (unless the phase of each individual transmitted pulse is measured and stored)



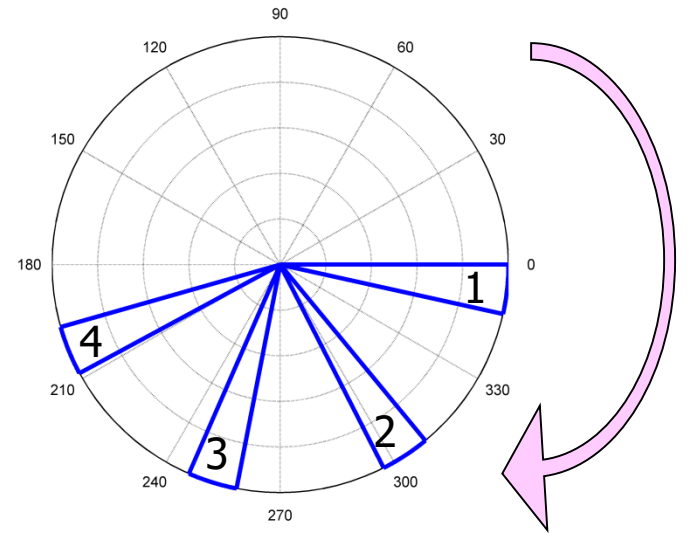
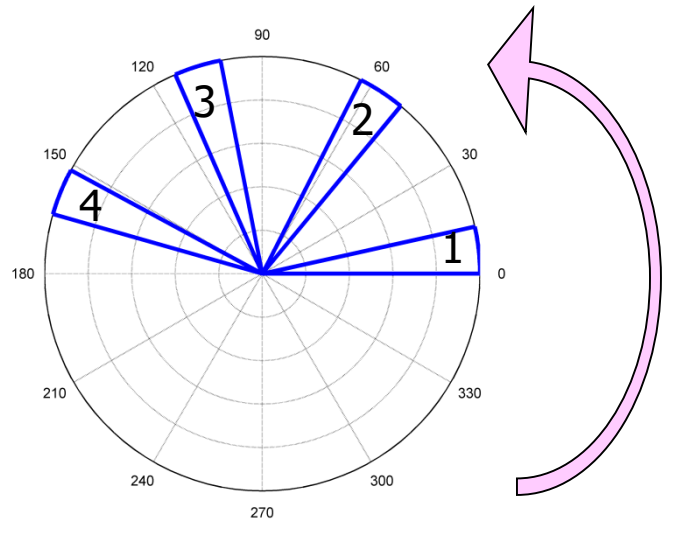
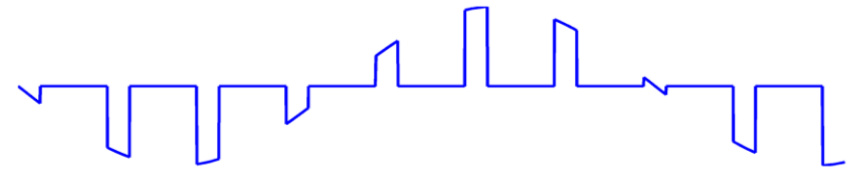
Why do we need both I and Q ?



I



Q



Checking the coherency of the reflection from the same 5 identical balls,
each with a σ of 0.1m^2 ,
when observed from an approaching aircraft

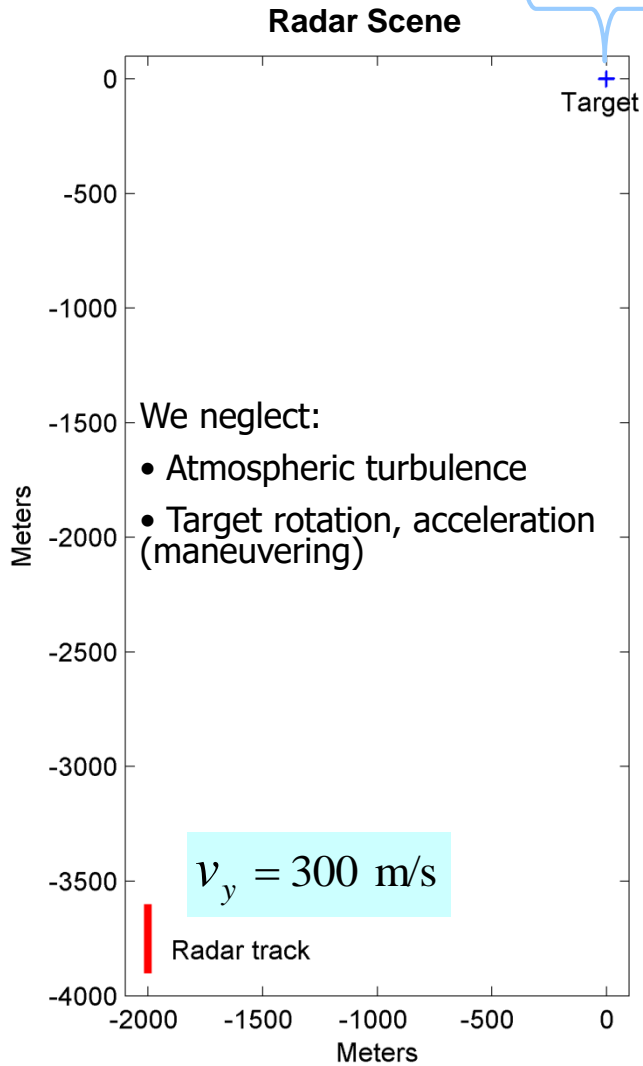
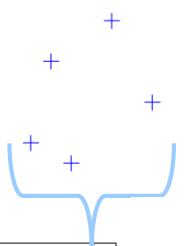
The change in phase and amplitude, of the reflection,
after removing the common Doppler effect

$$\sigma_k = \sigma_1 = 0.1 \text{ m}^2$$

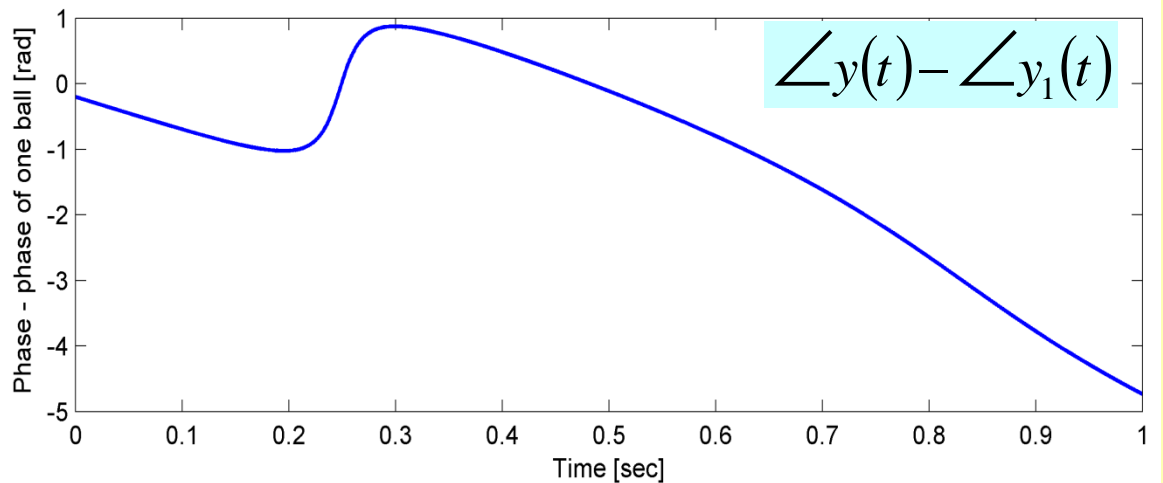
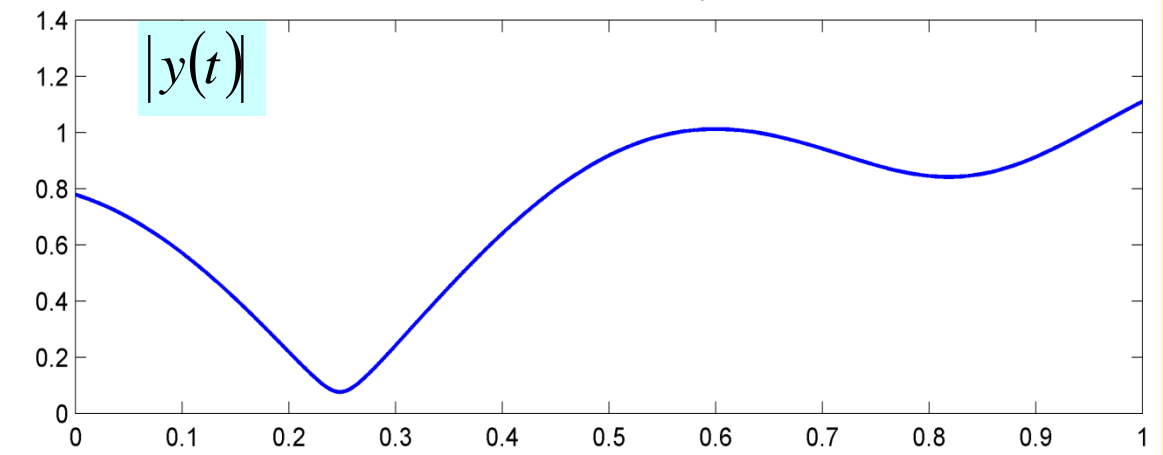
$$\lambda = 0.2 \text{ m}$$

$$y(t) = \sum_{k=1}^5 \sqrt{\sigma_k} \exp[-j4\pi R_k(t)/\lambda]$$

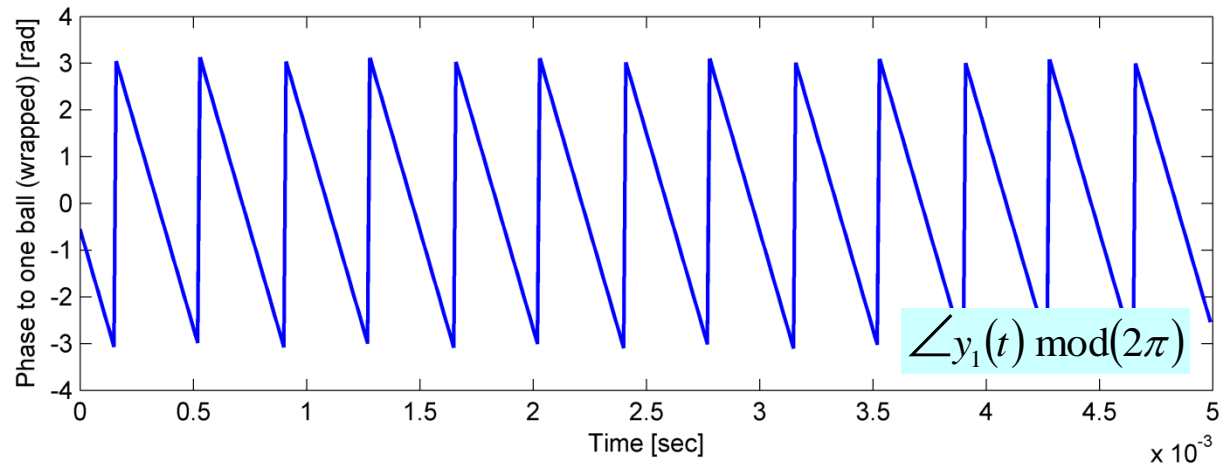
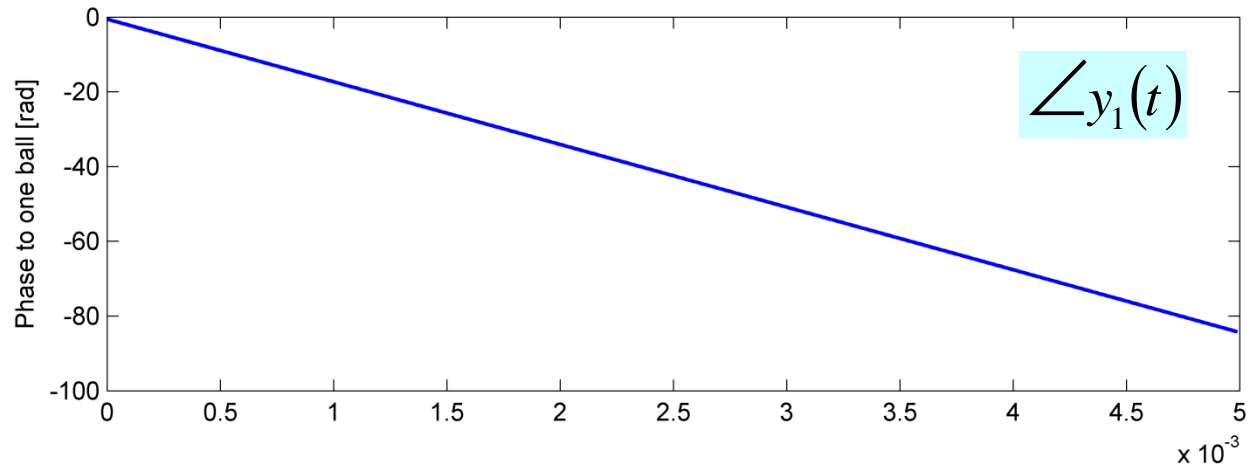
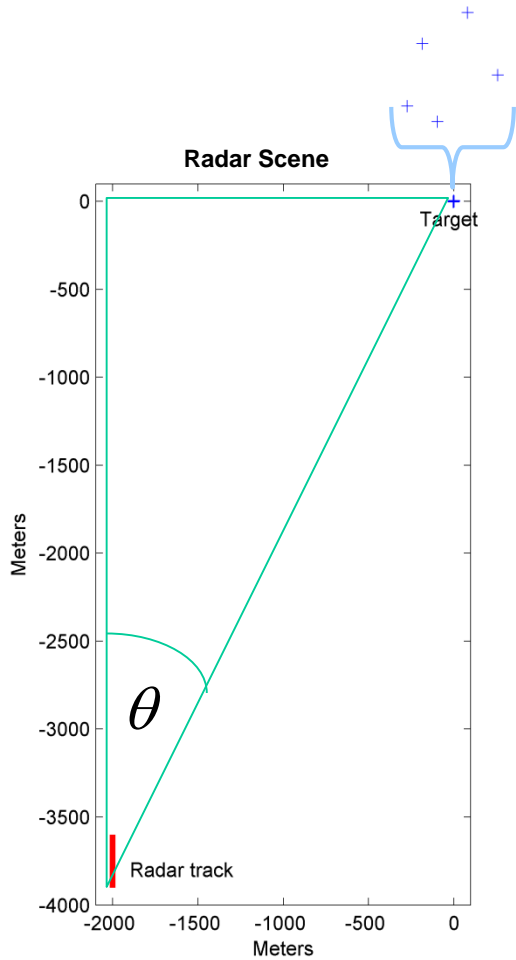
$$y_1(t) = \sqrt{\sigma_1} \exp[-j4\pi R_1(t)/\lambda]$$



Amp return from 5 balls, $\sigma_1=0.1[\text{m}^2]$

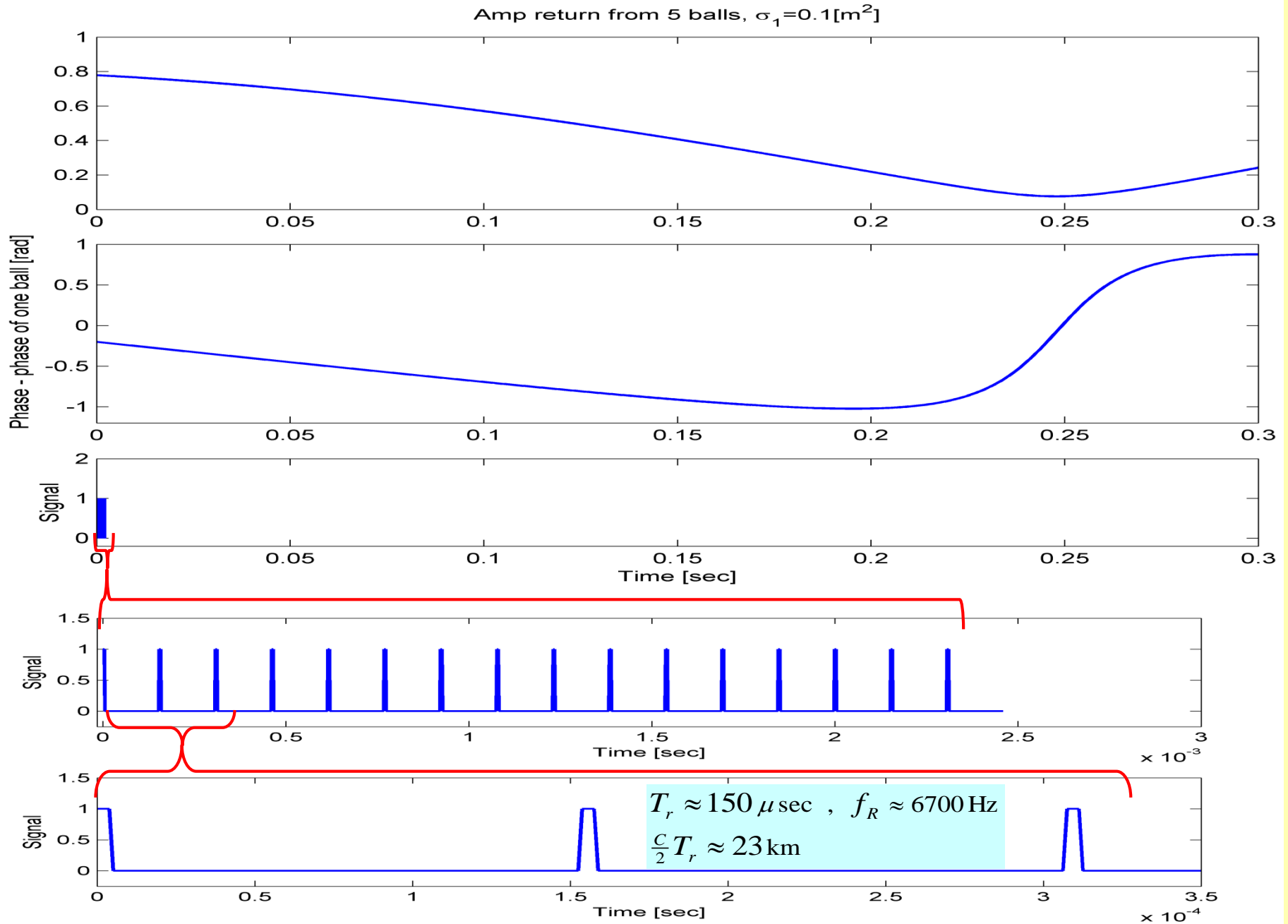


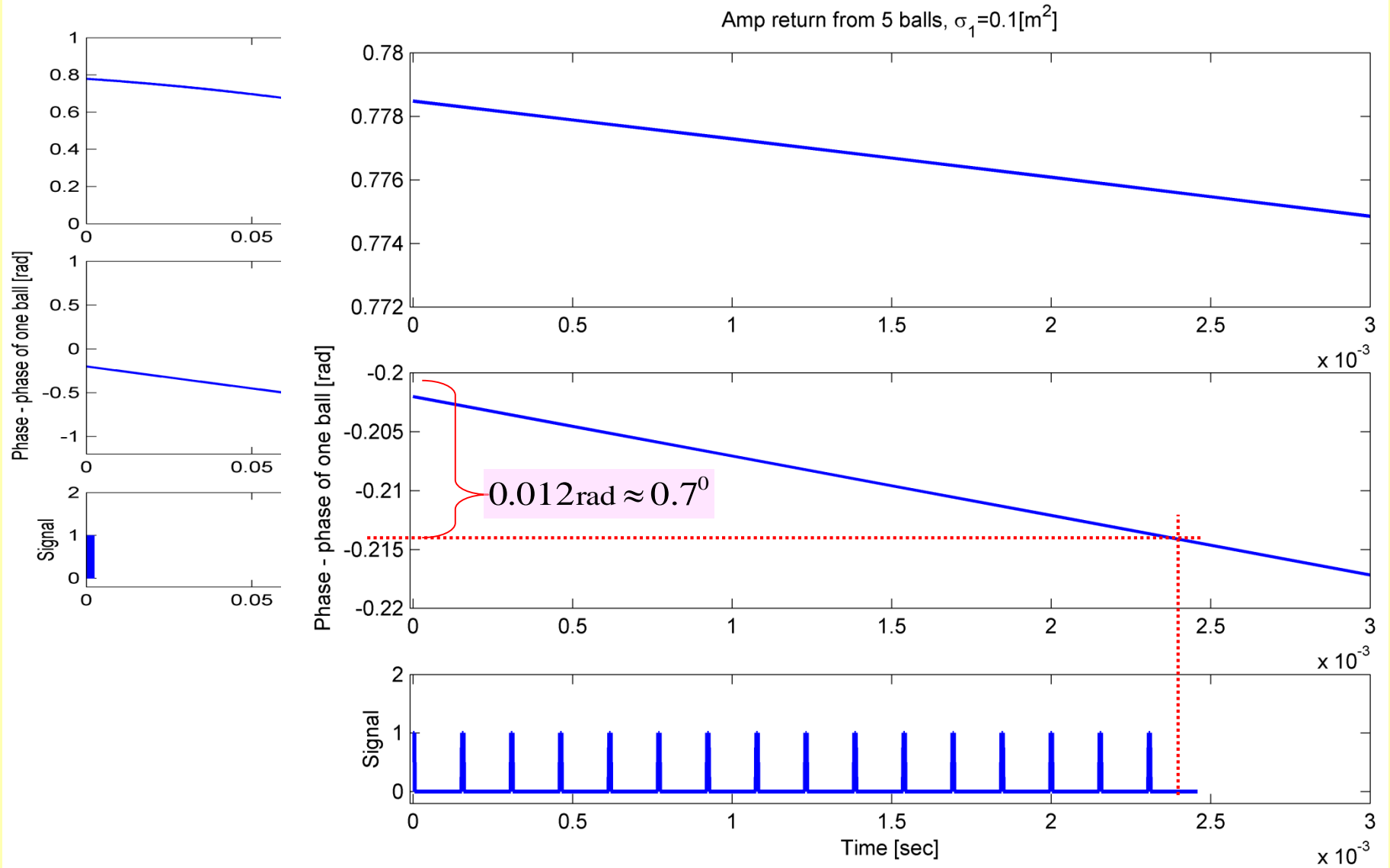
$$y_1(t) = \sqrt{\sigma_1} \exp[-j4\pi R_1(t)/\lambda]$$



$$f_D = \frac{2v}{\lambda} \cos \theta = \frac{2 \cdot 300}{0.2} \frac{3900}{\sqrt{2000^2 + 3900^2}} = 2670 \text{ Hz}$$

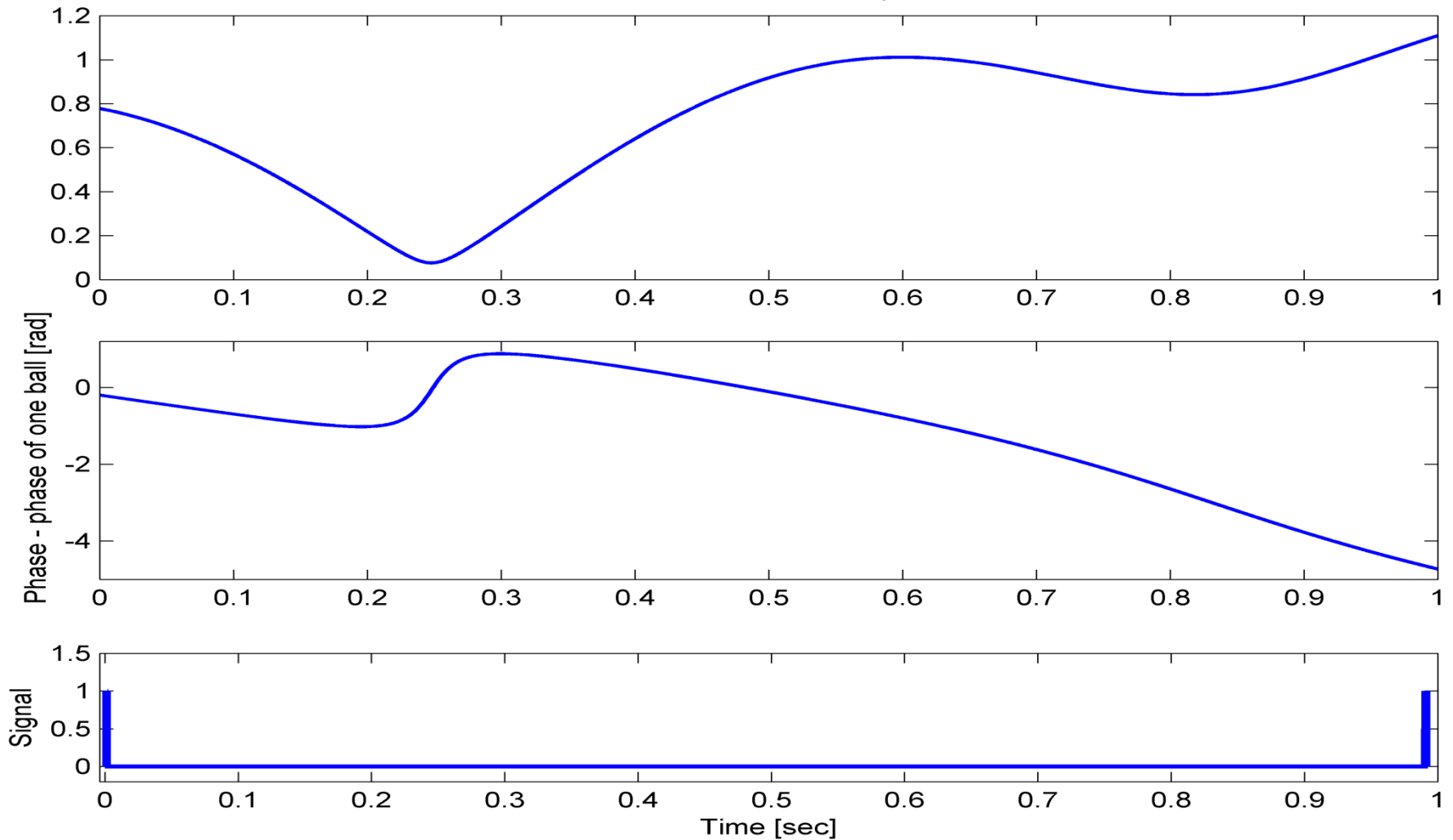
$$f_{D \text{ measured}} = \frac{13.3 \text{ cycles}}{0.005 \text{ sec}} = 2660 \text{ Hz}$$





Within the pulse-train the amplitude and phase do not change from pulse-to-pulse (except for a ramp phase change due to Doppler) implying coherence. The common amplitude is drawn from a Rayleigh PDF. The common phase is drawn from a uniform PDF.

Amp return from 5 balls, $\sigma_1=0.1[m^2]$



From pulse-train to pulse-train (scan-to-scan) the amplitude and phase change and are practically independent.
 The amplitudes are drawn from a Rayleigh PDF. The phases are drawn from a uniform PDF.

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COHERENCE ESTIMATION FOR A RADAR SIGNAL REFLECTED FROM AN AIR TARGET

M. M. Chernykh and O. V. Vasil'ev

Coherence estimation is considered for a centimeter-range radar signal as reflected from actual air targets; the coherence interval is determined, and numerical values are obtained for various types of air target.

Experimental results. The general method was to record signals from actual air targets at the intermediate frequency for subsequent spectral processing by a ground-level recording system based on a highly coherent Doppler-pulse radar working in the centimeter range. The single aircraft flew in straight stationary lines over the radar at distances of 50-100 km at heights of 5000-6000 m with speeds of 800-1000 km/h under simple weather conditions.

CONCLUSIONS

An experimental analysis has been performed for centimeter-range signals reflected from actual aircraft, which indicates the following regularities. When one observes a target of fighter class from the forward hemisphere, the reflected signal has a strong dominant component. The coherence in the reflected signal is largely determined by the stability of the high-frequency structure in that component. The length of the coherence interval governs the maximum coherent processing time and is 300-400 msec, while the correlation interval for the amplitude of the dominant component lies in the range 450-600 msec.

Other RCS PDFs

- Many radar targets are *not* well-modeled as “many equal-strength scatterers”
- Many distributions have been proposed for modeling radar targets based on empirical measurements

<i>Model Name</i>	<i>pdf for RCS σ</i>	<i>Comment</i>
Nonfluctuating, Marcum, Swerling 0, or Swerling 5	$p(\sigma) = \delta_D(\sigma - \bar{\sigma})$ $\text{var}(\sigma) = 0$	Constant echo power, e.g. calibration sphere or perfectly stationary reflector with no radar or target motion.
Rayleigh / Exponential, Chi-square of degree 2	$p(\sigma) = \frac{1}{\bar{\sigma}} \exp\left[-\frac{\sigma}{\bar{\sigma}}\right]$ $\text{var}(\sigma) = \bar{\sigma}^2$	Many scatterers, randomly distributed, none dominant. Used in Swerling case 1 and 2 models.
Chi-square of degree 4	$p(\sigma) = \frac{4\sigma}{\bar{\sigma}^2} \exp\left[-\frac{2\sigma}{\bar{\sigma}}\right]$ $\text{var}(\sigma) = \bar{\sigma}^2/2$	Approximation to case of many small scatterers + one dominant, with RCS of dominant equal $\approx 3x$ sum of RCS of others. Used in Swerling case 3 and 4 models.

$$\sigma = \frac{A^2}{2}, \quad \bar{\sigma} = A_0^2$$

$$p(A) = \frac{A}{A_0^2} \exp\left(-\frac{A^2}{2A_0^2}\right), \quad 0 \leq A$$

$$p(A) = \frac{2A^3}{A_0^4} \exp\left(-\frac{A^2}{A_0^2}\right), \quad 0 \leq A$$

RCS σ is related to the received power. The amplitude A is related to $\sqrt{\sigma}$. But if we use a square-law detector that yields A^2 we are back to the PDF of σ .

<i>Model Name</i>	<i>pdf for RCS σ</i>	<i>Comment</i>
Chi-square of degree $2m$, Weinstock	$p(\sigma) = \frac{m}{\Gamma(m)\bar{\sigma}} \left[\frac{m\sigma}{\bar{\sigma}} \right]^{m-1} \exp\left[-\frac{m\sigma}{\bar{\sigma}}\right]$ $\text{var}(\sigma) = \bar{\sigma}^2/m$	Generalization of the two preceding cases. Weinstock cases correspond to $0.6 \leq 2m \leq 4$. Higher degrees correspond to presence of a more dominant single scatterer.
Rice or Rician, non-central chi-square of degree 2	$p(\sigma) = \frac{1}{\bar{\sigma}} (1+a^2) \exp\left[-a^2 - \frac{\sigma}{\bar{\sigma}}(1+a^2)\right] \cdot I_0\left[2a\sqrt{(1+a^2)(\sigma/\bar{\sigma})}\right]$ $\text{var}(\sigma) = \frac{(1+2a^2)}{(1+a^2)^2} \bar{\sigma}^2$	Exact solution for one dominant scatterer plus many small ones. Ratio of dominant RCS to sum of small RCS is a^2 .

<i>Model Name</i>	<i>pdf for RCS σ</i>	<i>Comment</i>
Weibull	$p(\sigma) = CB\sigma^{C-1} \exp[-B\sigma^C]$ $\bar{\sigma} = \Gamma(1+1/C)B^{-1/C}$ $\text{var}(\sigma) = B^{-2/C} [\Gamma(1+2/C) - \Gamma^2(1+1/C)]$	Empirical fit to many measured target and clutter distributions. Can have longer "tail" than above cases.
Log-normal	$p(\sigma) = \frac{1}{\sqrt{2\pi} s\sigma} \exp\left[-\ln^2(\sigma/\sigma_m) / 2s^2\right]$ $\bar{\sigma} = \sigma_m \exp(s^2/2)$ $\text{var}(\sigma) = \sigma_m^2 \exp(s^2) [\exp(s^2) - 1]$	Empirical fit to many measured target and clutter distributions. "Tail" is longest of above cases. σ_m is the median value of σ .

$$p(A) = \frac{2}{\Gamma(C)} \left(\frac{C}{B}\right)^C A^{2C-1} \exp\left(\frac{-C}{B} A^2\right), 0 < A$$

$$a = A^2$$

$$p(a) = \frac{a^{C-1} \exp(-\frac{C}{B} a)}{(\frac{B}{C})^C \Gamma(C)}, 0 < A$$

Generalized Chi

Gamma pdf

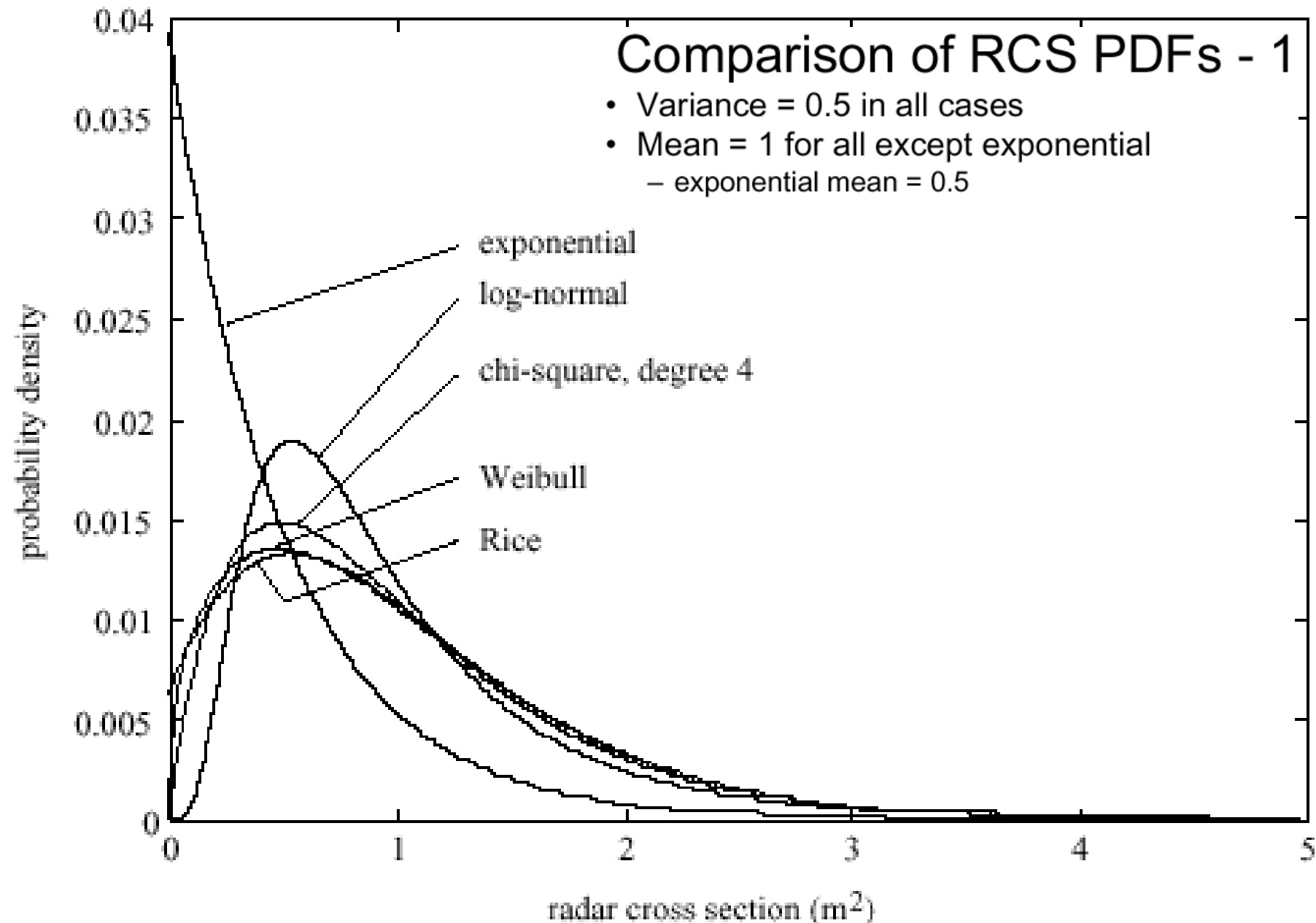
TABLE 2.3 Common Statistical Models for Radar Cross Section

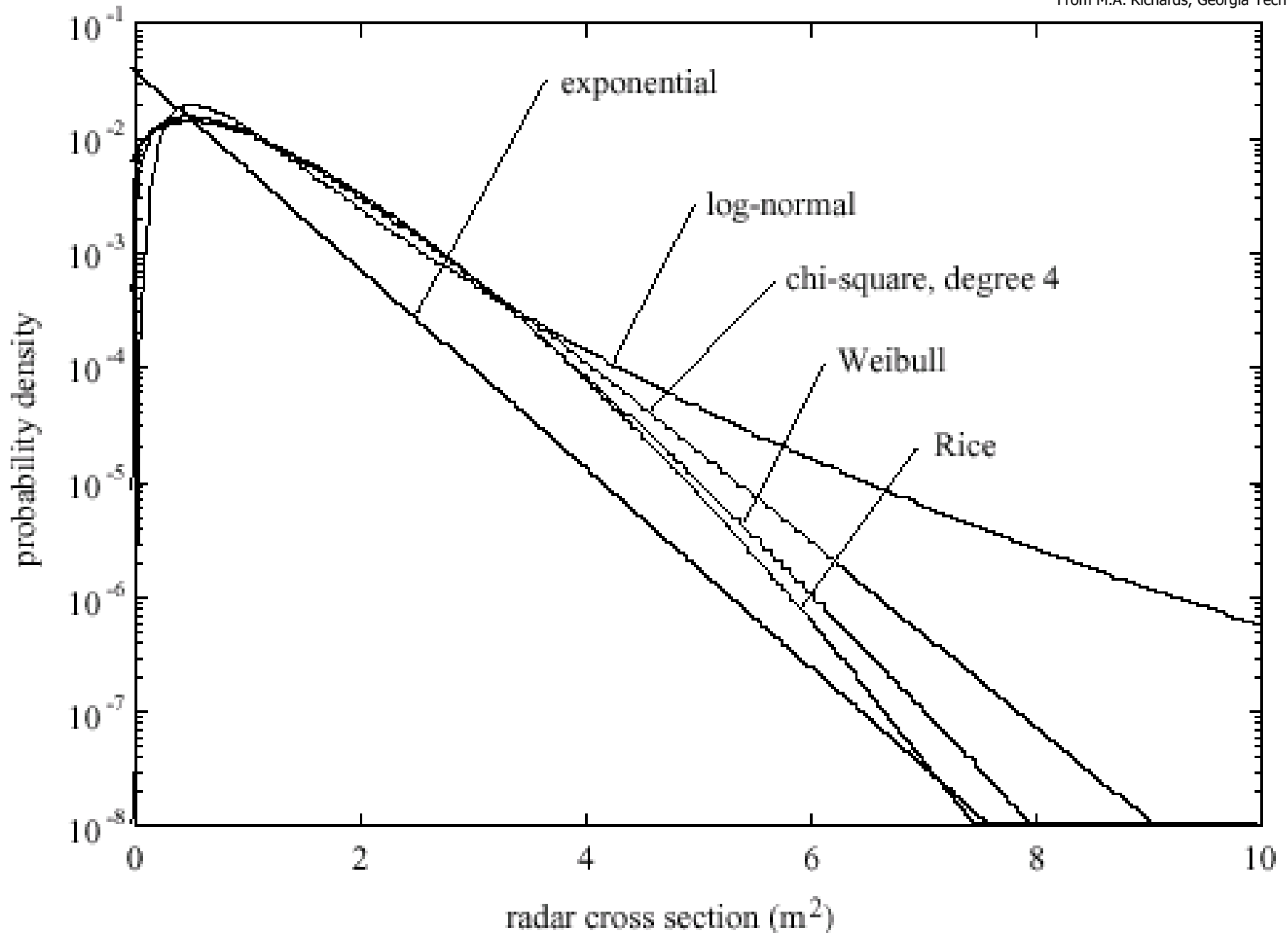
Model name	pdf for RCS σ	Comment
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Chi-square of degree 4	$p_{\sigma}(\sigma) = \frac{4\sigma}{\bar{\sigma}^2} \exp\left[-\frac{2\sigma}{\bar{\sigma}}\right]$ $\text{var}(\sigma) = \bar{\sigma}^2/2$	Approximation to case of many small scatterers + one dominant, with RCS of dominant equal to $1 + \sqrt{2}$ times the sum of RCS of others. Used in Swerling case 3 and 4 models.
Chi-square of degree $2m$, Weinstock	$p_{\sigma}(\sigma) = \frac{m}{\Gamma(m)\bar{\sigma}} \left[\frac{m\sigma}{\bar{\sigma}}\right]^{m-1} \exp\left[-\frac{m\sigma}{\bar{\sigma}}\right]$ $\text{var}(\sigma) = \bar{\sigma}^2/m$	Generalization of the two preceding cases. Weinstock cases correspond to $0.6 \leq 2m \leq 4$. Higher degrees correspond to presence of a more dominant single scatterer.
Rice or Rician, noncentral chi-square of degree 2	$p_{\sigma}(\sigma) = \frac{1}{\bar{\sigma}} (1 + a^2) \exp\left[-a^2 - \frac{\sigma}{\bar{\sigma}}(1 + a^2)\right]$ $\times I_0\left[2a\sqrt{(1 + a^2)(\sigma/\bar{\sigma})}\right]$ $\text{var}(\sigma) = \frac{(1 + 2a^2)}{(1 + a^2)^2} \bar{\sigma}^2$	Exact solution for one dominant scatterer plus many small ones. Ratio of dominant RCS to sum of small RCS is a^2 .
Weibull	$p_{\sigma}(\sigma) = CB\sigma^{C-1} \exp[-B\sigma^C]$ $\bar{\sigma} = \Gamma(1 + 1/C)B^{-1/C}$ $\text{var}(\sigma) = B^{-2/C}[\Gamma(1 + 2/C) - \Gamma^2(1 + 1/C)]$	Empirical fit to many measured target and clutter distributions. Can have longer "tail" than previous cases.
Log-normal	$p_{\sigma}(\sigma) = \frac{1}{\sqrt{2\pi} s\sigma} \exp\left[-\ln^2(\sigma/\sigma_m)/2s^2\right]$ $\bar{\sigma} = \sigma_m \exp(s^2/2)$ $\text{var}(\sigma) = \sigma_m^2 \exp(s^2)[\exp(s^2) - 1]$	Empirical fit to many measured target and clutter distributions. "Tail" is longest of previous cases. σ_m is the median value of σ .

Mark Richards: *Fundamentals of radar signal processing.*

One- vs. Two-Parameter PDFs

- One-parameter (mean) vs. two-parameter (mean, shape (variance))
 - two-parameter allows separate specification of mean and variance
 - fits wider range of observations
- One-parameter: nonfluctuating, exponential, chi-square of fixed degree
- Two-parameter: Rician, Weibull, log-normal





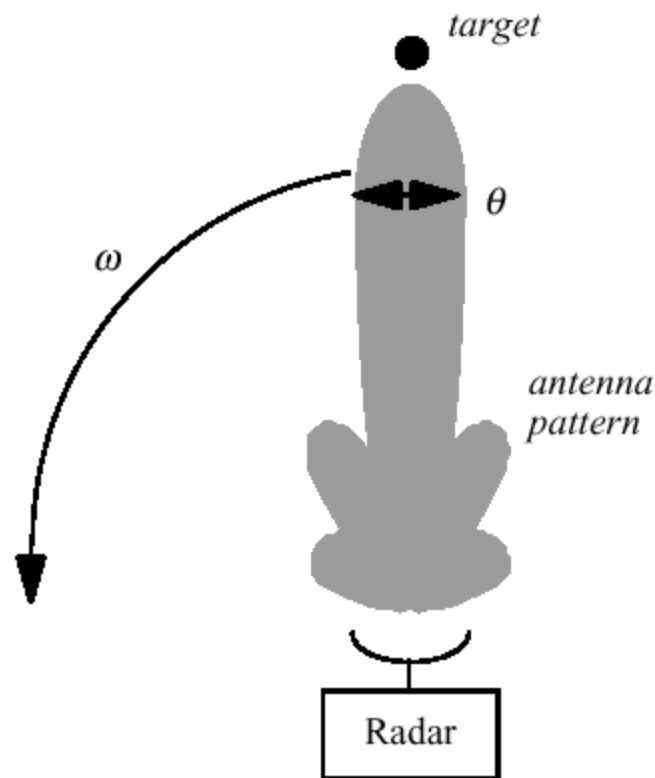
- In addition to pdf, we also need to know the correlation properties of RCS or voltage in angle, time, and frequency
- For a rigid target (building, vehicle):
 - if we illuminate from the same aspect and at same frequency, we expect the same measured response
 - if we change frequency or aspect, the relative phase of the individual echoes changes \Rightarrow measured response changes
- For natural terrain, response typically changes over time even if radar does not move or change frequency
 - in trees, leaves move due to wind
 - waves on ocean surface move
- This “internal motion” causes clutter to decorrelate in time
 - function of type of terrain, wind speed, frequency

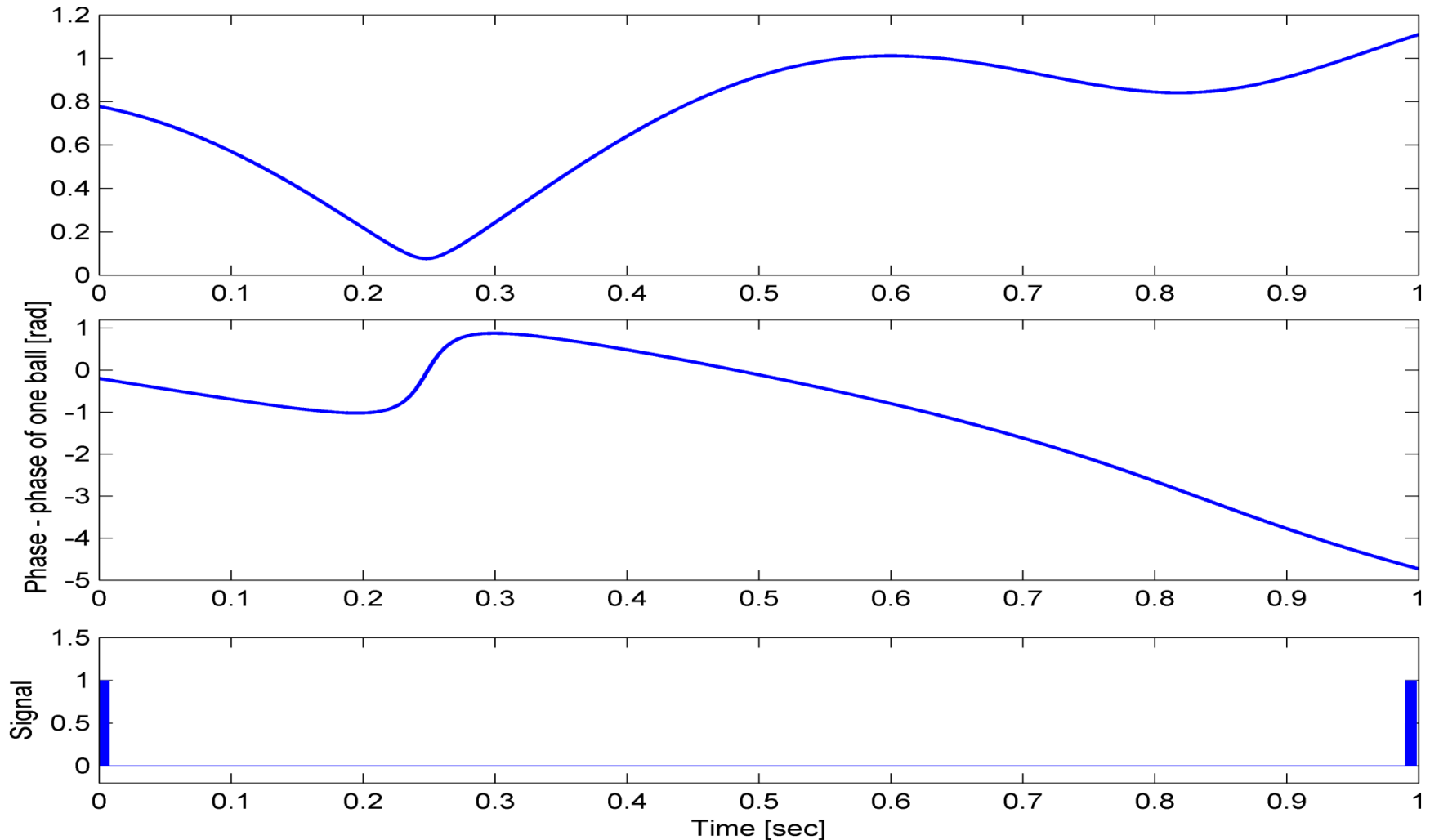
Pulse-to-Pulse Decorrelation

- Many detection algorithms operate on a group of N repeated measurements from N different pulses
 - pulses will be integrated to improve SNR
- An important question is the pulse-to-pulse decorrelation
 - noise aside, do we get the same measurement on successive pulses?
 - or does it change with time?

Rotating Antenna Model

- On each rotation (“scan”), N pulses are received from the target
- How are the N measured echoes at a given range related?
- And how do they relate to the N new measurements from the next scan?



Amp return from 5 balls, $\sigma_1=0.1[\text{m}^2]$ 

From pulse-train to pulse-train (scan-to-scan) the amplitude and phase change and are practically independent.

The amplitudes are drawn from a Rayleigh PDF. The phases are drawn from a uniform PDF.

Decorrelation Models for N -Pulse Bursts

- Scan-to-scan: assume the measured pulse amplitudes on one scan are independent of those on the first scan
 - but drawn from same pdf
- Intra-scan: Two limiting assumptions are commonly used:
 - scan-to-scan: all N measurements are identical
 - pulse-to-pulse: all N measurements are independent of one another, but from same pdf

Within
a scan

Swerling Models

- A significant body of detection theory has been developed using the four Swerling models of target RCS and Gaussian noise
 - *relatively* tractable
- Applicable to envelope detection based on N pulses
- Swerling proposed four different target models formed by choosing one each of
 - two different target RCS pdfs
 - two different intra-pulse decorrelation models

Swerling PDFs

- Rayleigh (exponential power)
 - good for complex targets with no dominant scatterer
- Chi-square of degree 4
 - approximation to case of one dominant scatterer and many smaller scatterers
 - total RCS of small scatterers = RCS of dominant scatterer / 3 ←
 - Rician pdf is exact for this case
 - but harder to work with

Swerling Decorrelation Models

- Scan-to-scan and pulse-to-pulse models discussed earlier

Swerling Models

<i>Probability Density Function of RCS</i>	<i>Decorrelation</i>	
	<i>scan-to-scan</i>	<i>pulse-to-pulse</i>
Rayleigh/exponential	Case 1	Case 2
chi-square, degree 4	Case 3	Case 4

- Nonfluctuating target sometimes called Swerling case 0 or case 5



Peter Swerling
1929 - 2000

P. Swerling “Probability of detection for fluctuating targets”, RAND RM-1217, March 17, 1954. Reprinted *IRE Trans. Information Theory*, vol. 6, no. 5, 1960, pp. 269-308.

P. Swerling “Radar probability of detection for some additional fluctuating target cases”, *IEEE Trans. Aerospace and Electronic Systems*, vol. 33, no. 2, 1997, pp. 698-709.

*Personal note: **Peter Swerling** was a research assistant to **Jess Ira Marcum** at the Rand Corp. in the 50's. I had the privilege to meet him at a dinner at the home of Solomon Golomb in Pasadena, CA. That evening I learned that the assignment to extend Marcum's work to fluctuating targets took Swerling a whole year to complete, The following related MATLAB example can be completed in few hours.*



בדיקת המודלים של סוורלינג

המטרה תתואר ע"י 9 מחזירים בעלי שח"מ זהה $\sigma_n = 0.1m^2, n = 1, 2, \dots, 9$ ועוד מחזיר עשירי שלו שח"מ הנתון ע"י:

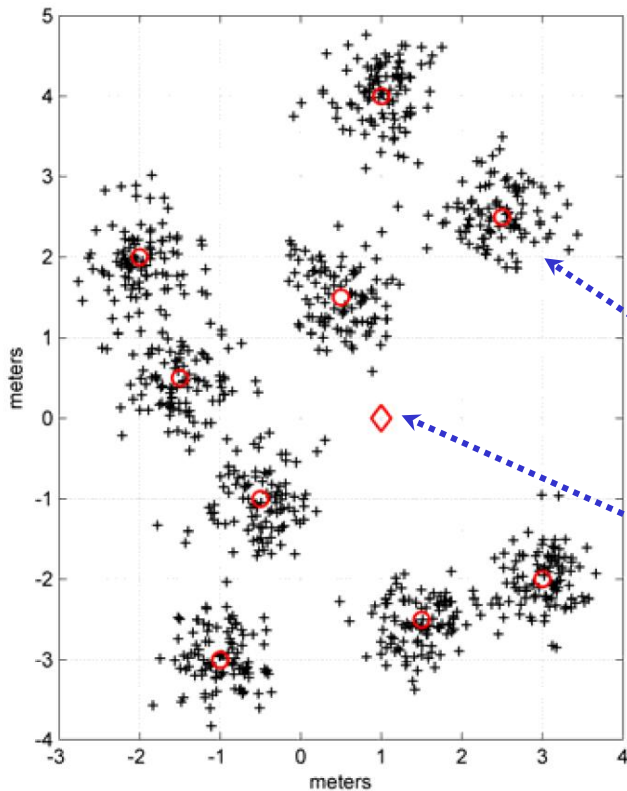
$$\sigma_{10} = \begin{cases} \sigma_1, & \text{SW1, SW2} \\ K \sum_{n=1}^9 \sigma_n, & \text{SW3, SW4} \end{cases} \quad (4)$$

המיקום הממוצע (במטרים) של תשעת המחזירים הקטנים נתון ע"י הקואורדינטות הקרטזיות שבטבלה: כדי להשיג אקראיות, המיקום האמיתי בהרצה מס. q נתון ע"י:

$$\begin{aligned} x_{n,q} &= \bar{x}_n + \Delta x_{n,q}, \quad n = 1, 2, \dots, 9 \\ y_{n,q} &= \bar{y}_n + \Delta y_{n,q}, \quad n = 1, 2, \dots, 9 \end{aligned} \quad , \quad q = 1, 2, \dots, Q \quad (5)$$

כאשר $\Delta x_{n,q}$ ו $\Delta y_{n,q}$ לקוחים מפילוג גאומי עם ממוצא אפס וסטטיית תקן 0.35 מטר. המיקום של מחזיר מס. 10 לא ישתנה מהרצה להרצה. דוגמא למיקומים של 10 המחזירים, ב $Q = 100$ הרצות, מופיעה בציור 1.

$\lambda = 0.25m$



n	\bar{x}_n	\bar{y}_n
1	3.0	-2.0
2	1.0	4.0
3	-2.0	2.0
4	-1.0	-3.0
5	0.5	1.5
6	-0.5	-1.0
7	2.5	2.5
8	-1.5	0.5
9	1.5	-2.5
10	1.0	0.0

הרצות

יש לבצע $Q = 100$ הרצות. בכל הרצה מקיף המכ"מ את המטרה בת 10 המחזירים במעגל שרדיוסו 10 ק"מ ומרכזו בנקודה $(x = 0, y = 0)$. המשרעת המתקבלת מהשח"מ השקול (משוואה 1) נמדדת כל 1/4 מעלה. כלומר בכל הרצה מתקבלות 1440 מדידות, ובמאה הרצות מתקבלות $M = 144000$ מדידות.

מ M המדידות של A יש לחשב PDF. ממומלץ להשתמש בפקודות MATLAB מסוג:

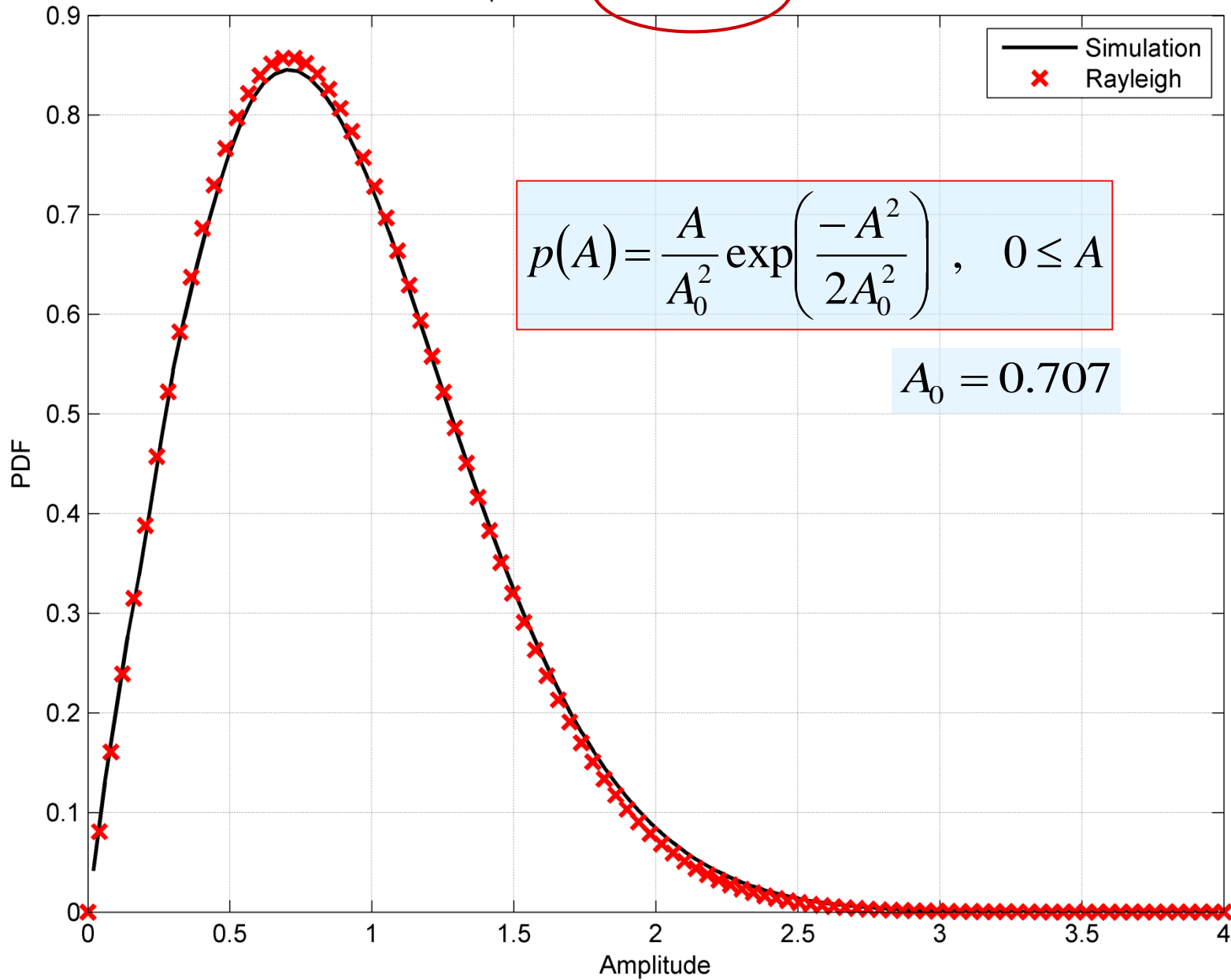
```
vv=linspace(0.02, 3.98);
[qdf(q, :)]=ksdensity(A(q, :), vv, 'kernel', 'epanechnikov', 'support', 'positive');
pdf=1/Q*sum(qdf);
```

את ה PDF שהתקבל מהסימולציה יש לצייר יחד עם ה PDF התאורטי לפי נוסחאות (2) או (3) בהתאמה. יש כמובן למצוא את הפרמטר A_0 של ה PDF התאורטי לקבלת ההתאמה הטובה ביותר.

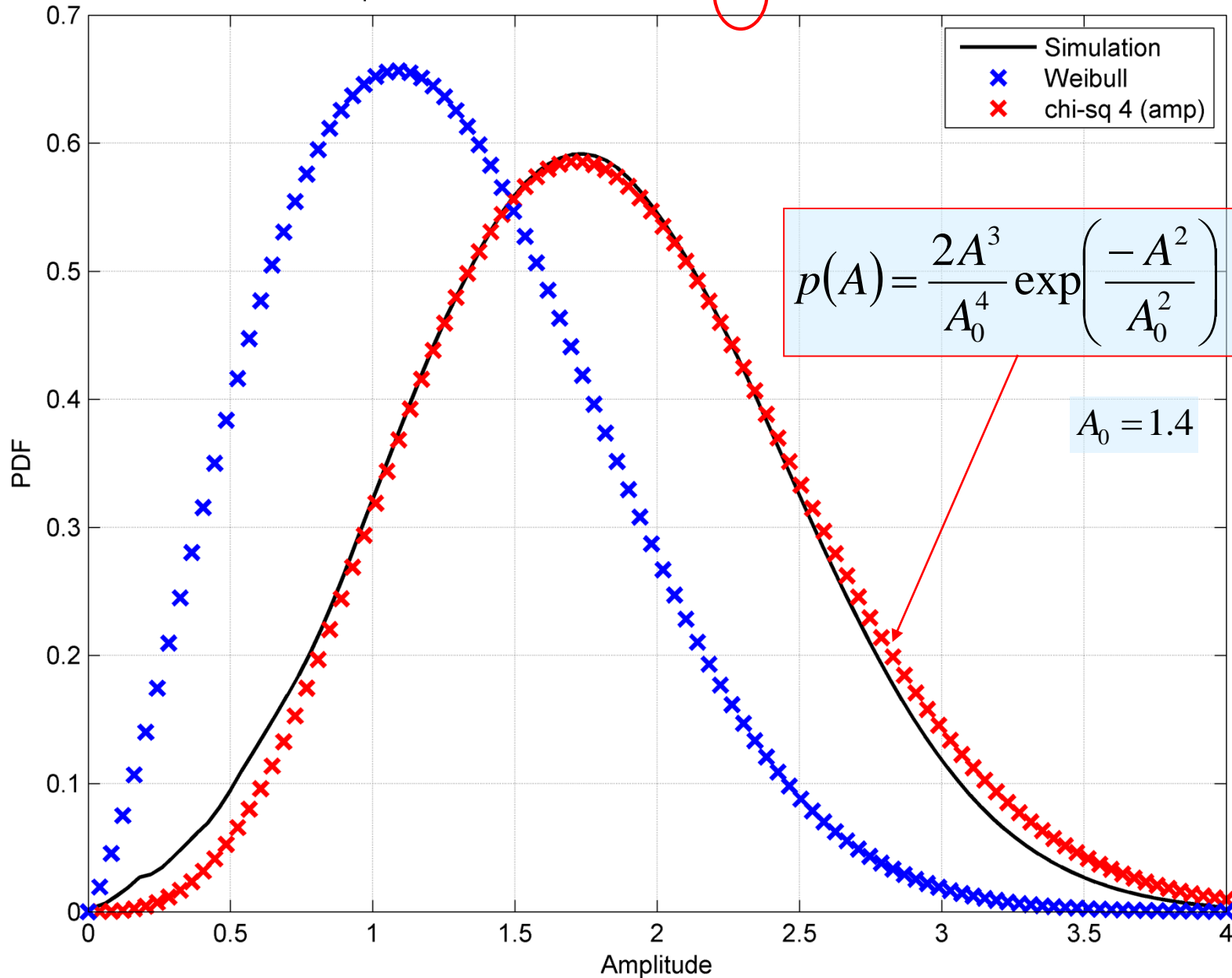
יש לבצע את הסימולציה פעם אחת עבור $SW1, SW2$, כאשר מתקיימת השורה העליונה של משוואה (4).

ופעם שנייה עבור $SW3, SW4$, להם מתקיימת השורה התחתונה של משוואה (4).

PDF of amplitude of 10 identical balls, seed=456



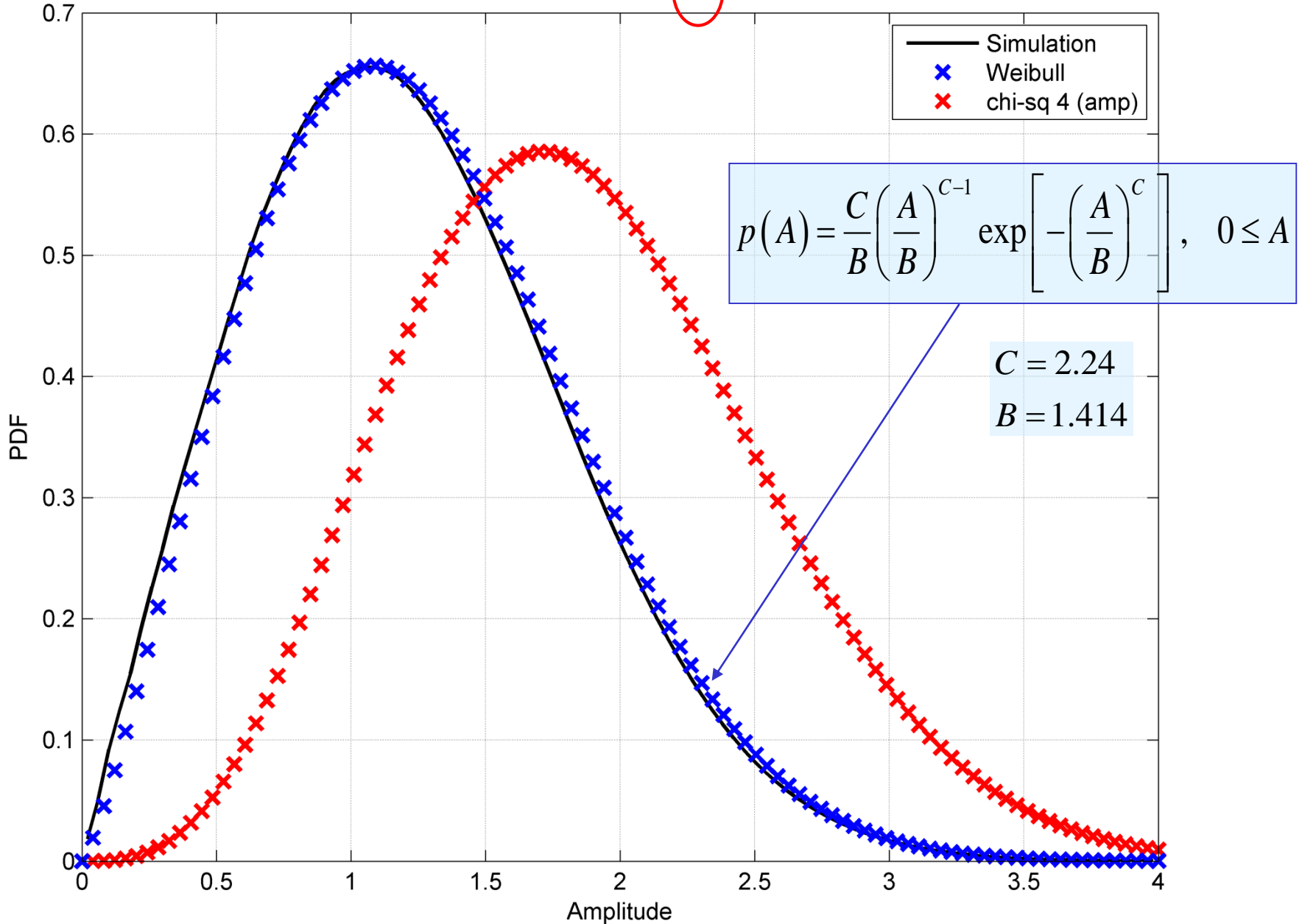
PDF of amplitude of 9+1 balls, Dominant=3 X sum of others seed=456



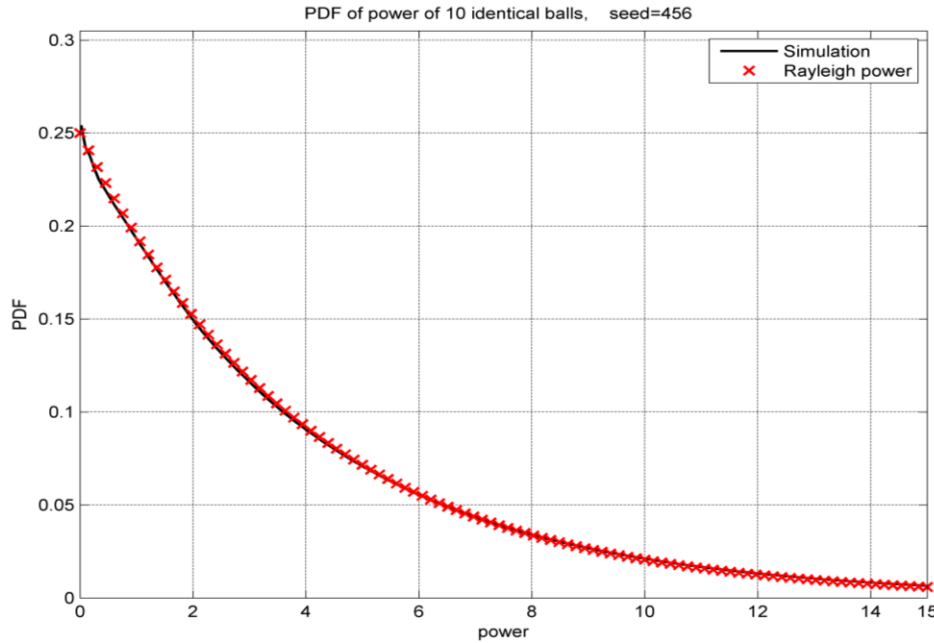
$$p(A) = \frac{2A^3}{A_0^4} \exp\left(-\frac{A^2}{A_0^2}\right), \quad 0 \leq A$$

$A_0 = 1.4$

PDF of amplitude of 9+1 balls, Dominant=1 X sum of others seed=456

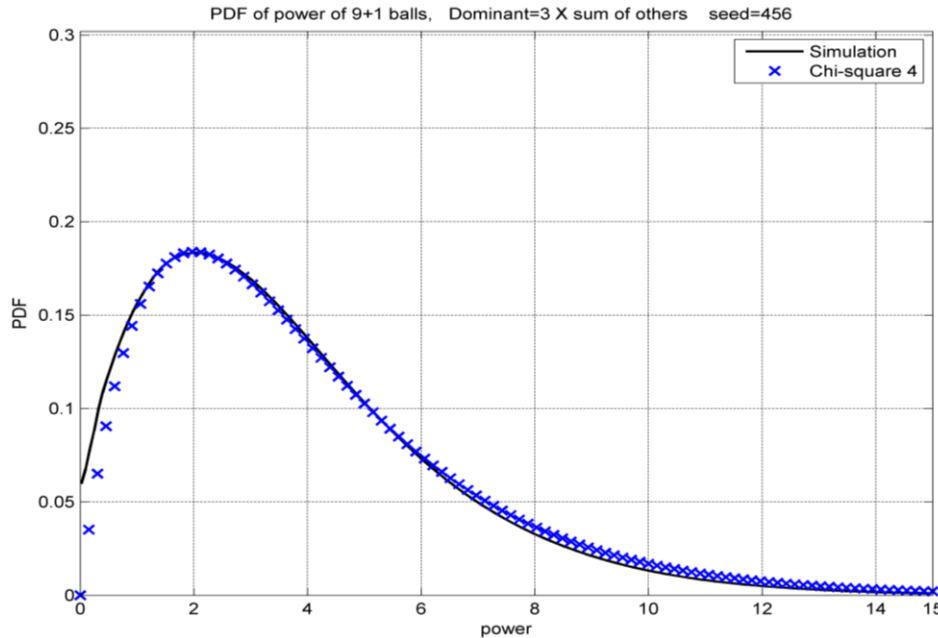


<i>Model Name</i>	<i>pdf for RCS σ</i>	<i>Comment</i>
Nonfluctuating, Marcum, Swerling 0, or Swerling 5	$p(\sigma) = \delta_D(\sigma - \bar{\sigma})$ $\text{var}(\sigma) = 0$	Constant echo power, e.g. calibration sphere or perfectly stationary reflector with no radar or target motion.
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Chi-square of degree 4	$p(\sigma) = \frac{4\sigma}{\bar{\sigma}^2} \exp\left[-\frac{2\sigma}{\bar{\sigma}}\right]$ $\text{var}(\sigma) = \bar{\sigma}^2/2$	Approximation to case of many small scatterers + one dominant, with RCS of dominant equal 3x sum of RCS of others. Used in Swerling case 3 and 4 models.
Chi-square of degree M	$p_\sigma(\sigma) = \chi^2\left\{\sigma; M, \bar{\sigma}\right\} = \frac{\sigma^{\frac{M}{2}-1}}{\left(\frac{\bar{\sigma}}{2}\right)^{\frac{M}{2}} \Gamma\left(\frac{M}{2}\right)} \exp\left(-\frac{\sigma}{\bar{\sigma}}\right)$	Moment generating function $C_{\chi^2}(s) = \left(1 - s \frac{\bar{\sigma}}{2}\right)^{-\frac{M}{2}}$



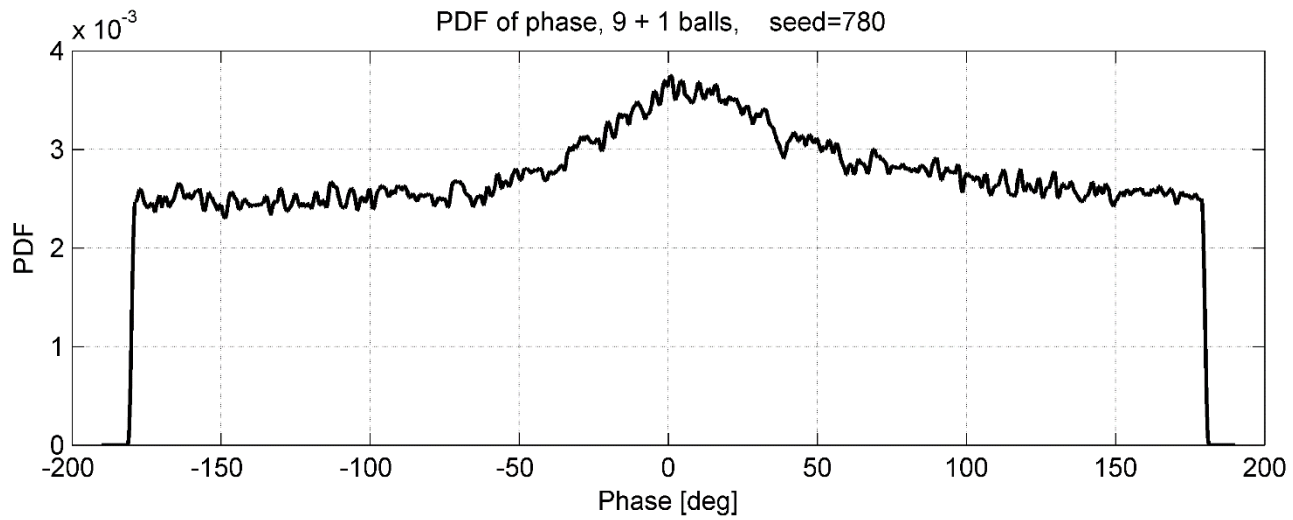
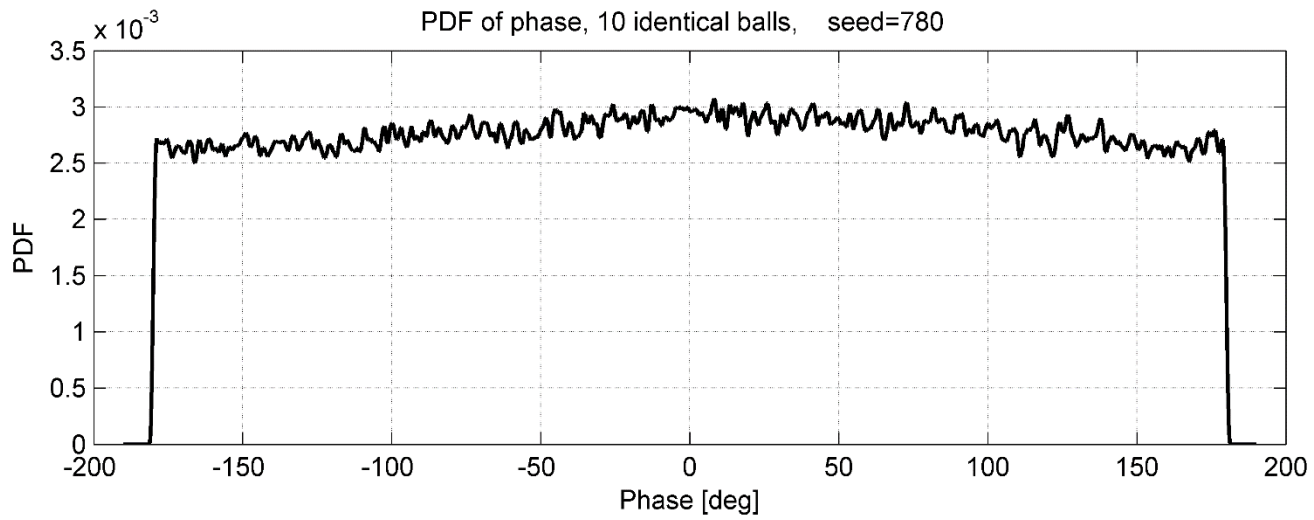
PDF of RCS or Power (SW1, SW2)

$$p_{\sigma}(\sigma) = \frac{1}{\bar{\sigma}} \exp\left(-\frac{\sigma}{\bar{\sigma}}\right), \bar{\sigma} = 4$$



PDF of RCS or Power (SW3, SW4)

$$p_{\sigma}(\sigma) = \frac{4\sigma}{\bar{\sigma}^2} \exp\left(-\frac{2\sigma}{\bar{\sigma}}\right), \bar{\sigma} = 4$$



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RCS Measurements and ISAR Images of Small UAVs

Massimiliano Pieraccini, Lapo Miccinesi, Neda Rojhani, University of Florence, Firenze, Italy

Pieraccini, M., Miccinesi, L. and Rojhani, N.
"RCS measurements and ISAR images of
small UAVs", *IEEE Aerospace and Electronic
Systems Magazine*, vol. 32 (9), pp. 28-32, 2017.

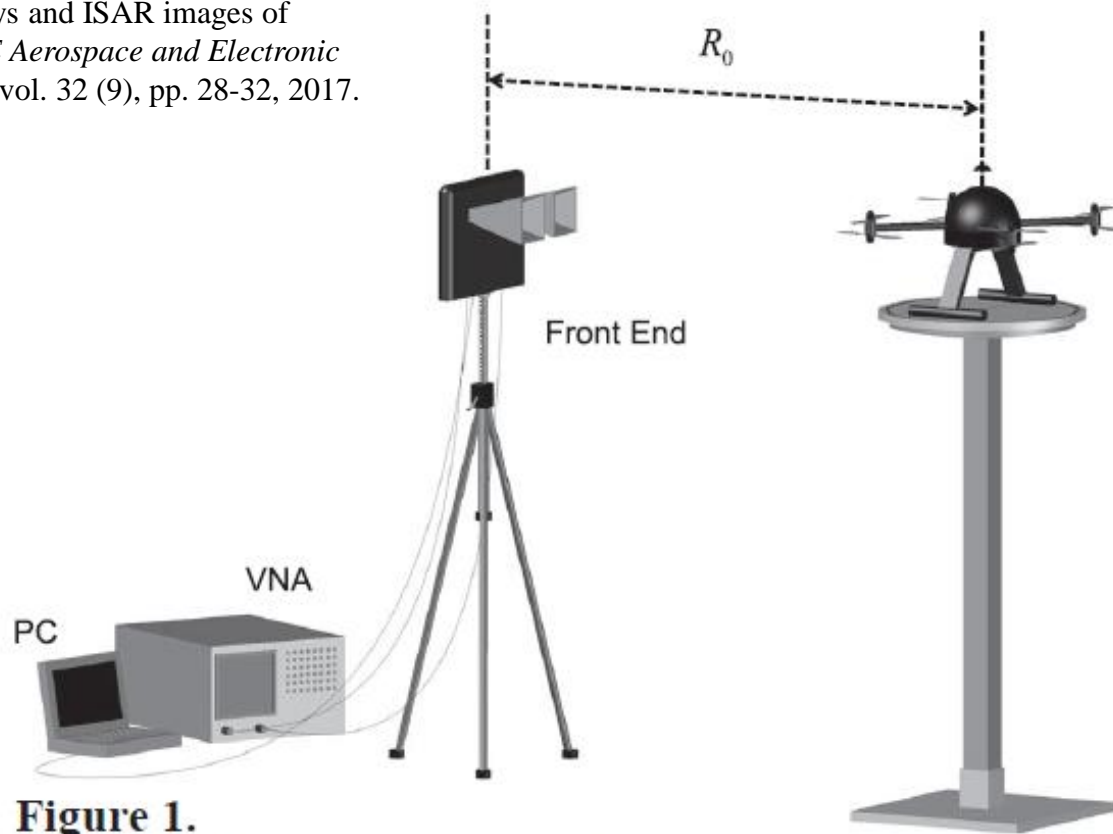


Figure 1.

Sketch of the measurement equipment.



Figure 2.
SYMA quadcopter.



Figure 3.
AirVision quadcopter.

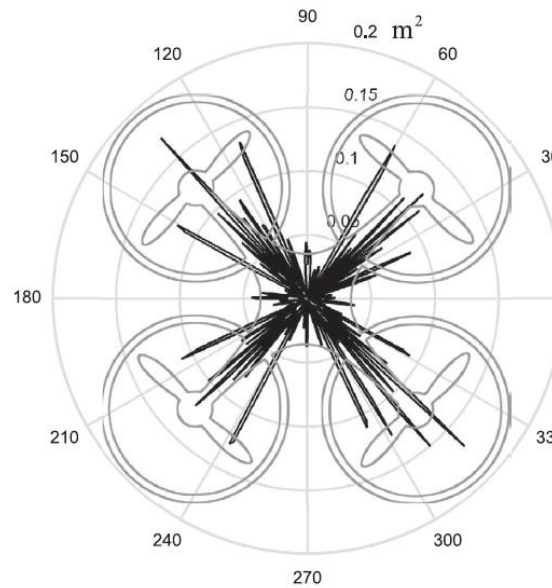


Figure 4.
RCS angular pattern of SYMA quadcopter.

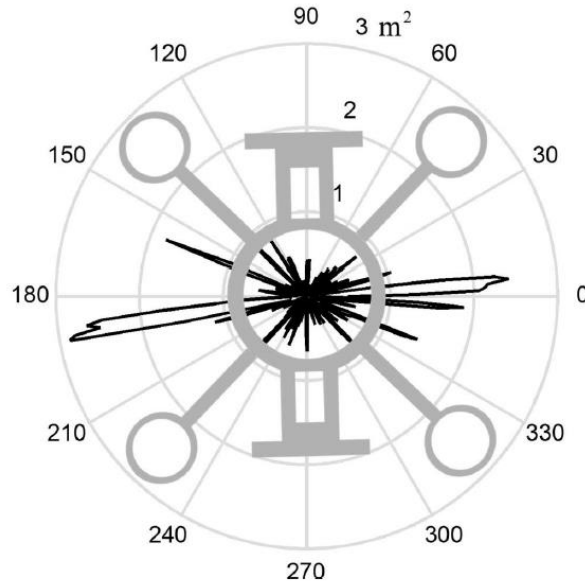


Figure 5.
RCS distribution of AirVision quadcopter.

Finally, by using the whole measurement set (9,270 single acquisitions), we calculated the RCS statistical distribution, as shown in Figure 10, that resulted in very good agreement with the well-known Swerling Case I distribution [12]:

$$p(\sigma) = \frac{1}{\bar{\sigma}} e^{-\frac{\sigma}{\bar{\sigma}}} \quad (4)$$

where $\bar{\sigma}$ is the mean value of RCS. Indeed the χ^2 goodness-of-fit test rejected the null hypothesis at the 1% significance level.

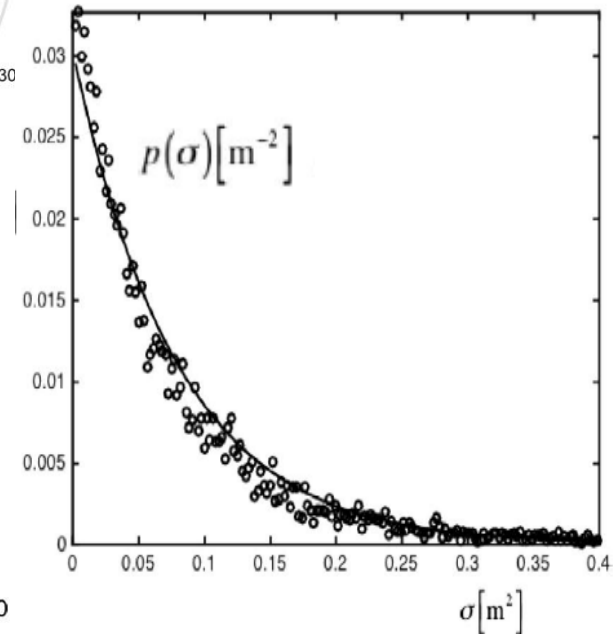


Figure 10.
RCS distribution of AirVision quadcopter using the whole measurement set for 3D ISAR. The full line is the Swerling distribution.