

# Evolution of Complexity and Cost for Planetary Missions Throughout the Development Lifecycle

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**Abstract**— This paper provides an update to the complexity-cost relationship presented at the Fourth IAA International Conference On Low-Cost Planetary Missions [Bearden, 2000]. [1] A significant data set exists for a thorough examination of the relationship between on-orbit performance, cost and schedule. Since 2000, the number of missions captured in the database has grown, while maintaining a stable complexity-cost relationship. This study focuses on a subset of missions with similar costs to Discovery-Class missions over the last two decades using the complexity metric to examine system development cost at various development milestones. The further from the successful mission complexity-cost trendline the study missions began their development, the greater the cost growth at launch. The magnitude of this cost growth appears to be linearly proportional to this initial “cost optimism.”

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## 1. INTRODUCTION

The Aerospace Corporation’s Complexity Based Risk Assessment (CoBRA) methodology has linked performance attributes to overall system complexity and has related complexity, independently, to cost and schedule. CoBRA was developed in 1999 as part of a doctoral thesis [Bearden, 1999] to understand the relationship of the complexity of a system with respect to its flight system development cost, and its development schedule. [2] A database of NASA, DoD and commercial missions was investigated to develop a complexity metric as a function of series of technical and programmatic parameters. Total development cost, which includes all flight system development costs except for launch vehicle, ground systems, mission operations and data analysis, is defined in FY11 fixed year dollars.

## 2. DATA USED FOR STUDY

To examine the relationship among risk, cost and schedule for NASA missions, data were assembled for a majority of missions launched over the past two decades (1989 to 2011). The basis for the relationships to be discussed is a database of technical specifications, costs, development time, mass properties and operational status for the spacecraft shown in Table 1. These data fall into three general categories: (1) NASA planetary and near-Earth spacecraft; (2) NASA Earth-orbiting satellites; and (3) Other U.S. government, non-NASA satellite missions serving as a baseline for comparison.

Missions that are nearing launch or have been launched, but have yet to complete a significant portion of their science missions are included, but it is noted that success has yet to be determined. Missions that had foreign sponsorship or relied on unknown international contributions were not considered. Also, shuttle science experiments and university-developed spacecraft were not considered. No human-rated systems or launch systems were considered. All of the data collected, except for the cost information, are from public sources. The cost information is from a proprietary database maintained by The Aerospace Corporation and is therefore presented only in aggregate form without reference to specific programs.

To isolate the relationships in development lifecycle changes in complexity and cost, only a subset of NASA spacecraft missions that meet certain criteria and constraints were considered. See Table 2.

**Table 1. CoBRA Mission Dataset**

NASA Earth Orbiting		NASA Near-Earth/Planetary		DoD/Other Earth Orbiting	
ACRIMSat	Jason-1	ACE	Mars Odyssey	AEHF	Milstar I/II
AIM	Landsat-7	ACCESS	Mars Pathfinder	ALEXIS	MSTI 1-3
Aquarius/SAC-D	LDCM	Cassini	MAVEN	APEX	MSX
Calipso	Lewis	Clementine	MCO	DARPA SAT	NFIRE
CGRO	METEOR	CONTOUR	MER	DMSP	ORBComm
Chandra (AXAF)	MICROLAB	Dawn	MESSENGER	DSCS	Orbital Express
CHIPSat	MMS	Deep Impact	MGS	DSP	PEGSAT
Clark	NuSTAR	DS1	MPL	FORTE	POGS/SSR
Cloudsat	OCO	Galileo	MRO	GEOEye-1	RADCAL
COBE	Polar	GENESIS	MSL	GEO Lite	REX I/II
CORIOLOS	QuickSCAT	GRAIL	NEAR	GFO	SBIRS-GEO
DART	QuikTOMS	IXO	New Horizons	Globalstar	SCE
Envisat	SAMPEX	JDEM Omega	Phoenix	GLOMR I	STEP 0-4
EO-1/3	SDO	Juno	RBSP	GOES-N/R	STEX
EOS-Aqua	Seastar	JWST	SIM Lite	GPS	STPSat
EOS-Aura	SMAP	Kepler	SOHO	Ikonos	TEX
EOS-Terra	SORCE	LADEE	Spitzer (SIRTF)	Iridium	TSX-5
EUVE	SPIDR	LCROSS	ST-5	MACSAT	UFO
EXIST	SWAS	LISA	Stardust	MICROSAT	WorldView-1
FAME	Swift	LRO	STEREO	MightySat I/II	XSS-11
FAST	TDRSS	Lunar Prospector	THEIA		
FUSE	THEMIS	Magellan	TRIANA		
GALEX	TIMED	MAP	Wind		
GLAST	TOMS-EP	Mars Observer			
GPM	TOPEX/Poseidon				
GRACE	TRACE				
HESSI	TRMM				
HETE-1/2	VCL				
HST	WIRE				
IBEX	WISE				
ICESAT I/II	WXFT				
IMAGE	XTE				

**Table 2. Study Missions – Subset of NASA Missions**

Study Missions
Phoenix
STEREO
MESSENGER
Dawn
Genesis
NEAR
MRO
LRO
Juno

### 3. MEASURING COMPLEXITY

Our goal is to understand how technical and programmatic complexity relates to cost and schedule at the system level. For this reason the complexity index is derived based on performance, mass, power, destination and technology choices, to arrive at a broad representation of the system for purposes of comparison. In examining previously built systems, cost data are generally not publicly available at the subsystem level. Furthermore, low-level schedule data can also be elusive since in many cases it's not clear which element - spacecraft, subsystem, payload, launch vehicle, or some combination - is driving the development time. For these reasons a system-level assessment is desirable.

The complexity index uses a matrix of technical factors to place in rank order a new system relative to a baseline data set. Postulated complexity drivers include both general programmatic knowledge (heritage, requirements changes, foreign involvement, amount of redundancy, contractor experience, specification level, etc.) that require some measure of subjective judgment, and demonstrable objective technical parameters (mass, power, performance, pointing accuracy, downlink data rate, technology choices, etc.). In this case we have chosen to emphasize objective technical parameters. The parameters that contribute to the calculated spacecraft complexity index, statistics of interest, and an example calculation for satellite are summarized in Table 3. The total flight system development costs (payload instruments and spacecraft bus, excluding launch and operations) and development times (period of time from contract start until ready for launch) are the independent variables against which the complexity is compared.

To estimate spacecraft complexity we:

- (1) Identify the parameters that drive the system design;
- (2) Quantify the identified parameters;
- (3) Estimate complexity for the individual parameters relative to the data set as a whole;
- (4) Combine the parameters into an aggregate complexity index.

Parameters were selected based on requirements that exert a significant influence on spacecraft design or development. The relative importance of the parameters contributing to the complexity index and correlation among them in representing and/or driving technology selections remains to be fully investigated. Correlation of variables has the potential effect of weighting the estimate toward a single subsystem if an excessive number of interdependent parameters are used (e.g., solar array area is dependent on solar cell type, orbit geometry and required end-of-life power, all of which contribute to the complexity calculation) or under-representing the contribution of a particular subsystem to overall system complexity if an insufficient number of parameters are used. A balance is achieved by

assuring that at least one parameter and at most six parameters are used to represent each of the traditional spacecraft subsystems.

All parameters are demonstrable parameters dictated by project, mission or system requirements. The strength of using a number of interrelated parameters is that peculiarities associated with the specific implementation approach will be averaged out. These descriptive parameters (number of instruments, mass, technical performance, subsystem characteristics, technological choices, etc.) are normalized based on the applicable range as designated by the programs in the database. There are two types of calculations:

- (1) Discrete choices such as propulsion type (none, cold gas, monopropellant, bipropellant, or ion engine);

Discrete choices are determined as follows:

$$f_i = (i - 1)/(m - 1) \quad (1)$$

$i = 1 \dots m$ , where  $m$  is the number of discrete choices

- (2) Continuous parameters (e.g. mass, power, pointing accuracy) that represent a range of potential values bounded by a minimum and maximum.

Continuous (non-discrete) parameters are analyzed using the Microsoft Excel<sup>®</sup> function PERCENTRANK that returns the rank of a value in a data set as a percentage of the data set. This function is used to evaluate the relative standing of a value within the data set. For example, PERCENTRANK can be used to evaluate the standing of launch mass among all launch masses in the database. If the value does not match one of the values in the array, PERCENTRANK interpolates to return the correct percentage rank. The syntax is as follows:

$$f_j = \text{PERCENTRANK}(\text{Array}, X) \quad j = 1 \dots n \quad (2)$$

where, *Array* is the range of data with numeric values that defines relative standing<sup>1</sup>, *X* is the value for which we want to know the rank, and *n* is the number of continuous parameters.

With  $f_i$  and  $f_j$  representing discrete and continuous complexity parameters, the overall complexity factor,  $F_c$ , is defined as:

$$F_c = (\sum f_i + \sum f_j)/(m + n) \quad i = 1 \dots m, j = 1 \dots n \quad (3)$$

<sup>1</sup> Note that for parameters where a smaller number implies a more strenuous requirement such as pointing accuracy and knowledge that  $f_j$  is calculated as  $1 - \text{PERCENTRANK}(\text{Array}, X)$ .

**Table 3. Factors Contributing to Complexity Index Calculation**

**Complexity-based Risk Analysis (CoBRA)**

Factor	Unit	Min	Median	Mean	Max	Example Satellite	
Development Cost (Project)	(FY11\$M)	2.2	230	490	4463	857	
Development Time (Project)	(mos)	12	48	53	126	62	
Payload Mass	(kg)	0	120	317	6065	160	63%
Payload Average Power	(W)	0	107	387	6000	175	67%
Payload Peak Power	(W)	0	199	569	13025	175	46%
Payload Data Rate (average)	(kbps)	0	196	29871	330000	10	18%
Number of Payloads		1	3	4	27	10	93%
Payload Type		0	2	2	4	1	12%
Data Volume	(MB/day)	0	458	510675	21168000	107	32%
Foreign Partnership		{ None, GS, LV, SC Bus, PL, mult PL }				0	0%
Design Life	(mos)	0	36	52	330	30	42%
Launch Mass (Wet)	(kg)	17	658	1339	18189	371	30%
Satellite Mass (Dry)	(kg)	17	594	1102	16329	325	30%
Bus Dry Mass	(kg)	15	421	724	10264	165	21%
Level of Redundancy	(%)	0%	20%	40%	100%	0%	0%
Orbit Regime		{ STS/ISS, GEO, LEO/MEO, H-LEO/Dip, NE, Interplan (au) }				2.0	76%
BOL Power	(W)	12	688	1663	18000	296	24%
Orbit Average Power	(W)	3	327	732	6687	260	39%
EOL Power	(W)	3	577	1301	11264	271	29%
Solar Array Area	(m <sup>2</sup> )	0	6	14	175	1.0	8%
Solar Cell Type		{ None/Si, GaAs, GaAs-mult, GaAs-conc, RTG/R }				GaAs	50%
Battery Type		{ Lead-acid, NiCd, SNIcd, NiH2, Li-Ion }				li-ion	100%
Battery Capacity	(A-hr)	1	37	61	1222		
# Articulated Structures		0	0	1	13	0	0%
# Deployed Structures		0	1	2	20	0	0%
Solar Array Configuration		{ body-fixed, deployed, single-axis, articulated }				D	33%
Structures Material		{ Aluminum, Al w/Comp-face, Exotic, Composite }				Al	0%
ADCS Type		{ None/Magnetic, GG, Spin, 3-axis, Hi-spin, 3-axis (ST), Dual }				Spin	40%
Pointing Accuracy	(deg)	1.70E-06	0.1000	1.0782	20	1.00000	22%
Pointing Knowledge	(deg)	1.11E-06	0.0250	0.6459	20	0.01000	68%
#Thrusters+Tanks	(#)	0	9	10	52	11	60%
Propulsion Type		{ None, Cold-Gas, Mono, Biprop-(blow,pres), OB+US, Ion }				Mono	40%
Total Impulse (delta-V)	(m/sec)	0	117	1203	80130	320	70%
Downlink Comm Band		{ UHF/VHF/SHF, S, L, X, Ka/Ku, EHF, Opt }				S	20%
Max Downlink Data Rate	(kbps)	1	1280	37587	1460926	78	20%
Max Uplink Command Data Rate	(kbps)	0	2	276	40000	1.0	8%
Data Storage Capacity	(Mbytes)	0	580	36704	3000000	1024	57%
Thermal Type		{ passive, heaters, active, cryo }				heaters	33%
Multi-Element System?		{ single-sc, tdrs, mult (cl, aerobr, rend), entry/landed }				Single	0%
Complexity Index		2%	47%	45%	78%		35%
Normalized Complexity Index		0%	59%	56%	100%		43%

The mean complexity index is determined by averaging the individual factors. Equal weighting is used so that the analysis does not differentiate among parameters or allow any particular parameter to dominate.  $F_c$  is then normalized relative to the maximum and minimum to arrive at a percentage between 0 and 100%.

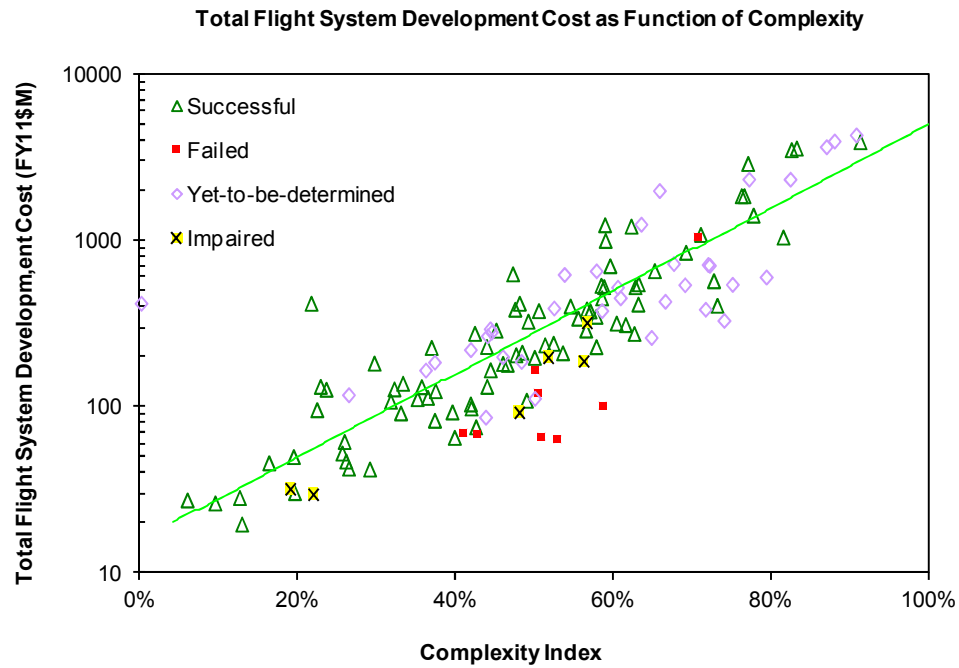
#### 4. EXAMINATION OF TRENDS

Using the process described above, a complexity index was calculated for the programs in Table 1, using the parameters shown in Table 3. The updated plot, Figure 1, displays a plot of complexity, as calculated by the CoBRA process, relative to the cost of satellite systems that were successful, impaired, suffered a failure, or status is yet-to-be determined. There are several conclusions that can be drawn from this figure. First, a high correlation between complexity and cost is apparent. Second, the average complexity of impaired and failed missions is higher than that of successful missions for the same cost. While the complexity index does not predict the manner in which a failure will occur, it does define a framework within which a new mission being considered may be compared with past missions. Bitten, Bearden and Emmons [2005] took the CoBRA methodology a step further with multi-variable regression analysis performed for a set of NASA science missions to determine quantitative relationships between complexity as a function of cost and schedule, schedule as a function of cost and complexity, and cost as a function of schedule and complexity. [3]

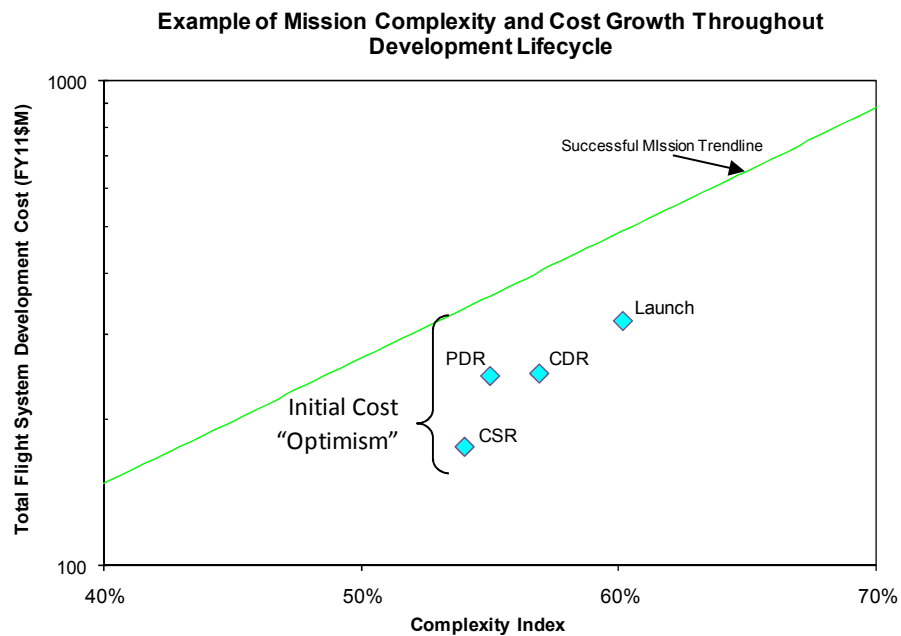
Figure 2 is an example of the typical complexity and cost growth of missions during their developmental lifecycle, with initial “cost optimism.”

Also captured is the development cost estimates (phases B-D) at milestones CSR, PDR, CDR, and launch. The changes in cost growth between CSR and launch as a function of distance from the overall successful mission trendline is shown in Figure 3.

Several observations can be drawn from this data. First, as projects progress in their development lifecycle, costs increase for all missions except NEAR, which had a large cost reduction early in development. Also, this relationship appears to be linearly proportional to the extent of initial cost estimation optimism at CSR. If this relationship holds over time and for a large data set, it could be of tremendous utility in assessing new mission concepts’ potential for cost growth.



**Figure 1. Updated Complexity and Cost Relationship**



**Figure 2. Complexity and Cost Growth throughout Development Lifecycle**

## Cost Growth vs Optimism @ CSR

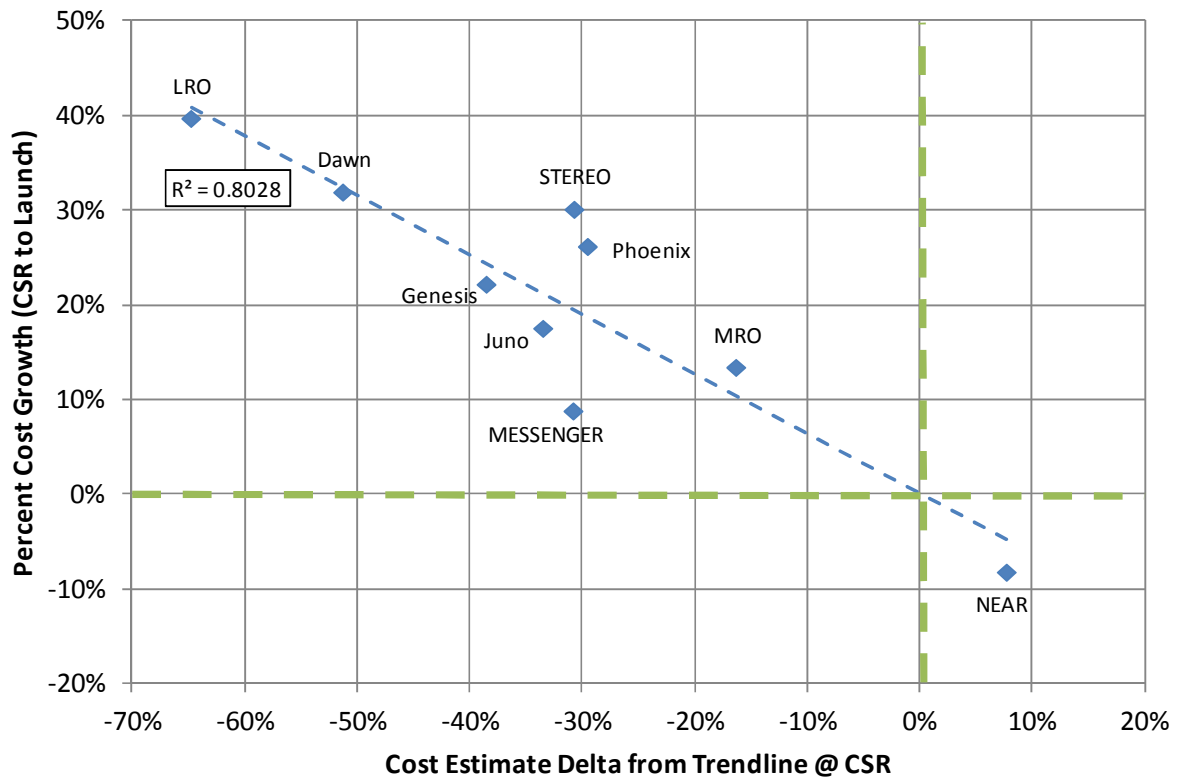


Figure 3. Cost Growth as a Function of Early Milestone Optimism

## 5. SUMMARY

As the CoBRA database mission-set has increased over the last decade, the strong complexity-cost relationship has continued. For the study mission subset selected, the relationship between early lifecycle cost optimism and development lifecycle cost growth appear to follow a linear trend. This relationship demands further study. As new missions are added to the larger dataset, more of the intermediate milestone data is becoming available. Intermediate milestone data is being made available via retrospective studies such as NASA's Cost Analysis Data Requirement (CADRe). Similar follow-on studies of larger datasets are needed to strengthen this development lifecycle cost growth relationship.

## 6. REFERENCES

- [1] Bearden, David A., A Complexity-Based Risk Assessment Of Low-Cost Planetary Missions: When Is A Mission Too Fast And Too Cheap?, Fourth IAA International Conference On Low-Cost Planetary Missions, May 2-5, 2000.
- [2] Bearden, D.A.: A Methodology for Technology Insertion Assessment Balancing Performance, Cost and Risk, Ph.D. Dissertation, University of Southern California, May, 1999.
- [3] Bitten, R.E., Bearden, D.A. and Emmons, D.E.: "A Quantitative Assessment Of Complexity, Cost, And Schedule: Achieving a Balanced Approach For Program Success", Sixth IAA International Conference On Low-Cost Planetary Missions, May 2005.

## BIOGRAPHIES



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## ACRONYMS

ACE Advanced Composition Experiment  
ACRIMSat Active Cavity Radiometer Irradiance Monitor Satellite  
AEHF Advanced Extremely High Frequency  
AIM Aeronomy of Ice in the Mesosphere  
Al Aluminum  
ALEXIS Array of Low-Energy X-ray Imaging Spectrometers  
APEX Advanced Photovoltaic and Electronics Experiment  
APL Applied Physics Laboratory  
Aquarius/SAC-D Satellite de Aplicaciones Cientificas-D  
ATP authority to proceed  
au astronomical units  
A-hr Ampere hours  
BATC Ball Aerospace Technology Corporation  
Biprop bipropellant  
BMDO Ballistic Missile Defense Organization  
BOL Beginning of Life  
CGRO Compton Gamma Ray Observatory  
Chandra (AXAF) Advanced X-ray Astrophysics Facility  
CHIPSat Cosmic Hot Interstellar Plasma Spectrometer  
CNES French space agency  
COBE Cosmic Background Explorer  
CONAE Argentine Space Commission  
CONTOUR Comet Nucleus Tour  
CTA Computer Technology Associates  
DARPA Defense Advanced Research Project Administration  
DART Demonstration for Autonomous Rendezvous Technology  
deg degrees  
DMSP Defense Meteorological Satellite Program  
DOD Department of Defense  
DOE Department of Energy  
DSI Defense Systems, Inc.  
DS-1 Deep Space 1 Mission  
DSCS Defense Satellite Communications System  
DSP Defense Support Program  
EO-1 Earth Observing 1 Mission  
EOL End of Life  
EOS Earth Observing System  
ESSP Earth System Science Pathfinder  
EUVE Extreme Ultra-Violet Explorer  
EXIST Energetic X-ray Imaging Survey Telescope  
FAME Full-sky Astrometric Mapping Explorer  
FAST Fast Auroral Snapshot Explorer  
FUSE Far Ultraviolet Spectroscopic Explorer  
FORTE Fast On-orbit Recording of Transient Events  
FY fiscal year  
GaAs gallium arsenide  
GALEX GALaxy Explorer  
GEO Geosynchronous Earth Orbit  
GFO Geosat Follow-on  
GLAST Gamma-ray Space Telescope

GLOMR Global Low Orbiting Message Relay  
 GOES Geostationary Operational Environmental Satellite  
 GPM Global Precipitation Measurement  
 GPS Global Positioning System  
 GRACE Gravity Recovery and Climate Experiment  
 GRAIL Gravity Recovery and Interior Laboratory  
 GSFC Goddard Space Flight Center  
 HESSI High Energy Solar Spectroscopic Imager  
 HETE High Energy Transient Experiment  
 HST Hubble Space Telescope  
 IBEX Interstellar Boundary Explorer  
 ICESAT Ice, Cloud, and land Elevation Satellite  
 IMAGE Imager for Magnetopause-to-Aurora Global Exploration  
 IXO International X-ray Observatory  
 JDEM Joint Dark Energy Mission  
 JHU Johns Hopkins University  
 JPL Jet Propulsion Laboratory  
 JWST James Webb Space Telescope  
 kbps kilobits per second  
 kg kilogram  
 KSLOC thousand software lines of code  
 LADEE Lunar Atmosphere and Dust Experiment Explorer  
 LANL Los Alamos National Laboratory  
 LaRC Langley Research Center  
 LASP Laboratory for Atmospheric and Space Physics  
 LCROSS Lunar Crater Observation and Sensing Satellite  
 LDCM Landsat Data Continuity Mission  
 LEO low Earth orbit  
 Lib Libration Point  
 Li-ion Lithium-ion  
 LISA Laser Interferometer Space Antenna  
 LLC Limited Liability Corporation  
 LMA Lockheed Martin Astronautics  
 LRO Lunar Reconnaissance Orbiter  
 LV Launch Vehicle  
 MACSAT Multiple-Access Communications Satellite  
 MAP Microwave Anisotropy Probe  
 MAVEN Mars Atmosphere and Volatile Evolution Mission  
 Mbytes megabytes  
 MCO Mars Climate Orbiter  
 MEO Medium Earth Orbit  
 MER Mars Explorer Rover  
 MESSENGER MErcury Surface, Space ENvironment, GEochemistry, and Ranging  
 MGS Meteosat Second Generation  
 MIT Massachusetts Institute of Technology  
 MGS Mars Global Surveyor  
 MIDEX Medium-Class Explorers  
 MIPS Millions Instructions Per Second  
 MIT Massachusetts Institute of Technology  
 m meter  
 MMS Magnetospheric Multiscale  
 MPL Mars Polar Lander

MRO Mars Reconnaissance Orbiter  
 MSFC Marshall Space Flight Center  
 MSL Mars Science Lander  
 MSTI Miniature Sensor Technology Integration  
 MSX Midcourse Space Experiment  
 NASA National Aeronautics and Space Administration  
 NEAR Near-Earth Asteroid Rendezvous Mission  
 NFIRE Near Field Infrared Experiment  
 NiCd Nickel Cadmium  
 NiH<sub>2</sub> Nickel Hydrogen  
 NOAA National Oceanic and Atmospheric Agency  
 NRL Naval Research Laboratory  
 NRO National Reconnaissance Office  
 NuSTAR Nuclear Spectroscopic Telescope Array  
 OCO Orbiting Carbon Observatory  
 ONR Office of Naval Research  
 PEGSAT Pegasus Satellite  
 POGS/SSR Polar-Orbit Geomagnetic Survey Solid-State Recorder  
 PL payload  
 QuickSCAT Quick Scatterometer  
 QuikTOMS Quick Total Ozone Mapping Spectrometer  
 R Correlation Coefficient  
 RADCAL radiation calibration  
 RBSP Radiation Belt Storm Probes  
 REX Radiation Experiment  
 SAC-B Satelite de Aplicaciones Cientificas B mission  
 SAMPEX Solar Anomalous and Magnetospheric Particle Explorer  
 SBIRS Space Based Infrared System  
 SC Spacecraft  
 SCE Selective Communications Experiment  
 SDIO Strategic Defense Initiative Organization  
 SDO Solar Dynamics Observatory  
 sec second  
 SHF Super High Frequency  
 SIM Space Interferometry Mission  
 SIRTf Space InfraRed Telescope Facility  
 SMAP Soil Moisture Active Passive  
 SMEX Small Explorer Program  
 SNiCd Super Nickel Cadmium  
 SNL Sandia National Laboratory  
 SOHO Solar & Heliospheric Observatory  
 SORCE Solar Radiation and Climate Experiment  
 SPIDR Spectroscopy and Photometry of the IGMs Diffuse Radiation  
 SSTI Small Spacecraft Technology Initiative  
 ST Star Tracker  
 ST Space Technology  
 STEP Space Test Experiment Platform  
 STEREO Solar TEREstrial Observatory  
 STEX Space Technology Experiment  
 STP Space Test Program  
 SWAS Submillimeter Wave Astronomy Satellite

TDRSS Tracking and Data Relay Satellite System  
TEX Transceiver Experiment  
THEMIS Time History of Events and Macroscale Interactions during Substorms  
TIMED Thermosphere Ionosphere Mesosphere Energetics and Dynamics  
TOMS-EP Total Ozone Mapping Spectrometer - Earth Probe  
TOPEX Topography Experiment  
TRACE Transition Region and Coronal Explorer  
TRMM Tropical Rainfall Measuring Mission  
TSX Tri-Service Experiment  
UFO UHF Follow-On  
UHF Ultra-High Frequency  
VCL Vegetative Canopy Lidar  
VHF Very-High Frequency  
W Watt  
WIRE Wide-field Infrared Experiment  
WISE Wide-Field Infrared Survey Explorer  
WXFT Wide Field X-Ray Telescope Mission  
XSS eXperimental Satellite System  
XTE X-ray Timing Explorer