# Data Management and Resource Sharing

Rigor & Reproducibility Worship 14 February 2018

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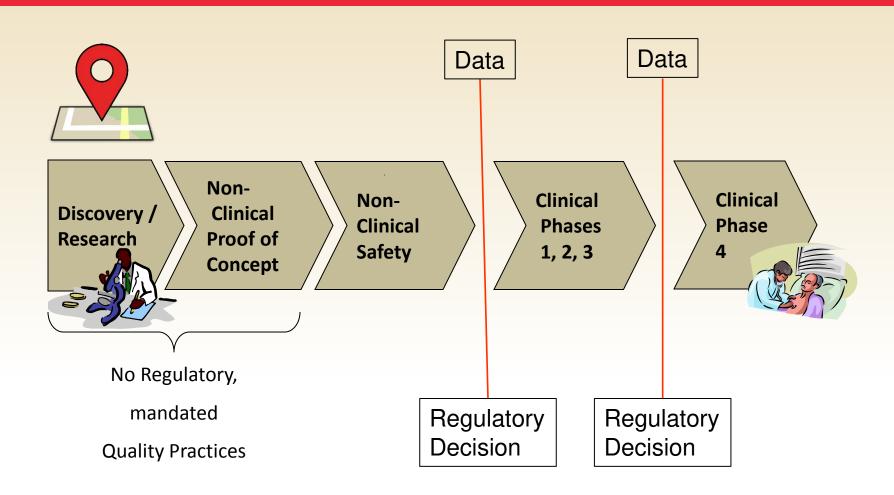




### **Topics**

- > Translational Science
- ➤ Data Lifecycle and Organization
- Good Documentation Practices

### **Product Approval Pathway**



#### Translational Science

What is the average cost to develop and gain marketing approval for a new drug?

\$2.558 billion

\$2.870 billion, including post-approval costs

Journal of Health Economics 47 (2016) 20–33

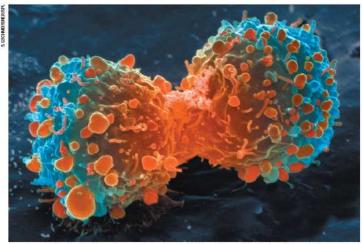
# **Economics of Reproducibility**

- > 53 landmark studies
- ➤ 6 confirmed (11%)
  - **≻**Controls
  - ➤ Reagents
  - ➤ Investigator bias
  - ➤ Described complete data set

#### COMMENT

AMAN INFLUENZA Shift expertise to track mutations where they emerge p.534 BRINSYSTEMS Past climates give valuable clues to future warming p.537 Instruction of science Descartes'
lost letter tracked using
Google p.540

onto an elusive stress hormone p.542



Many landmark findings in preclinical oncology research are not reproducible, in part because of inadequate cell lines and animal models.

# Raise standards for preclinical cancer research

C. Glenn Begley and Lee M. Ellis propose how methods, publications and incentives must change if patients are to benefit.

If forts over the past decade to characterize the genetic alterations /in human cancers have led to a better understanding of molecular drivers of this complex set of diseases. Although we in the cancer field hoped that this would lead to more effective drugs, historically, our ability to translate cancer research to clinical successhas been remarkably low?. Sadly clinical

trials in oncology have the highest failure rate compared with other therapeutic areas. Given the high unmet need in oncology, it is understandable that barriers to clinical development may be lower than for other disease areas, and a larger number of drugs with suboptimal preclinical validation will enter oncology trials. However, this low success rate is not sustainable or acceptable, and

investigators must reassess their approach to translating discovery research into greater dinical success and impact.

Many factors are responsible for the high failure rate, notwithstanding the inherently difficult nature of this disease. Certainly, the limitations of preclinical tools such as inadequate cancer-cell-line and mouse models' make it difficult for even



# Focus Areas for Reproducibility

- Randomization
- ➤ Blinding
- Sample size
- Data handling

http://www.ncbi.nlm.nih.gov/pubmed/23060188





Published in final edited form as: Nature. 2012 October 11; 490(7419): 187–191. doi:10.1038/naturel1556.

#### A call for transparent reporting to optimize the predictive value of preclinical research

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Author Contributions R.F., A.K.G., S.C.L., J.D.P., S.D.S., U.U. and W.K. organized the workshop, R.B.D., S.E.L., S.C.L., M.R.M. and S.D.S. wrote the manuscript. All authors participated in the workshop and contributed to the editing of the manuscript.

Author Information Reprints and permissions information is available at www.nature.com/reprints. The authors declare no competing financial interests. Readers are welcome to comment on the online version of the paper.

### Resource Sharing—NIH

NIH considers the sharing of unique research resources developed through NIH-sponsored research an important means to enhance the value and further the advancement of research.

When resources have been developed with NIH funds and the associated research findings published or provided to NIH, it is important that the **results be made readily available** for research purposes to qualified individuals within the scientific community.

https://grants.nih.gov/grants/peer/guidelines general/Resource sharing plans.pdf



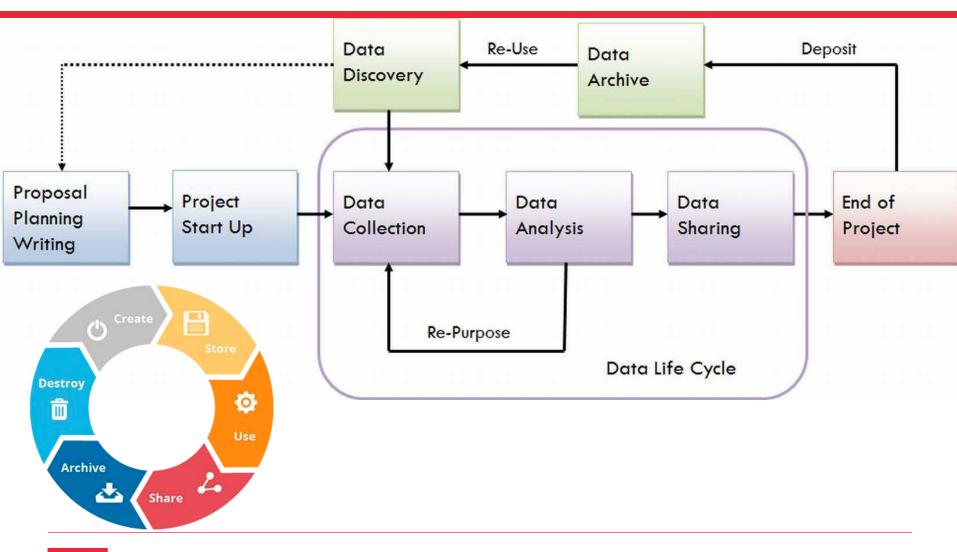
### Resource Sharing—NIH

- Samples
- Reagents
- Model organism (e.g., transgenic mouse strain)
- Data

#### Where Do We Start?

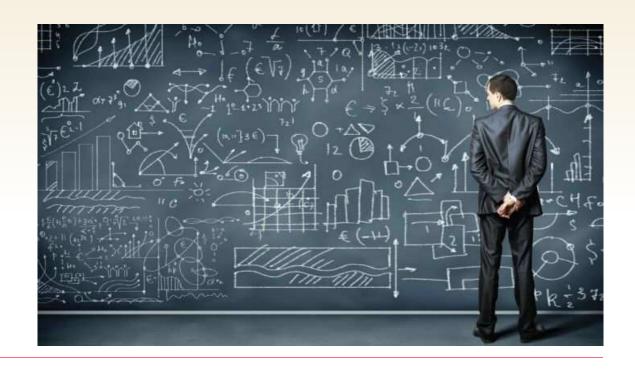


# Data Lifecycle



#### What Kind of Data?

- > Types (observational, derived, etc.)
- Format (text, numeric, modeling, images, etc.)
- Quantity
- > HIPAA
- Proprietary
- Owner



- Location
- Accessibility / Security (including encryption)
- Back-up
- Retention / Archival
- Data Migration / Operating System Sustainability

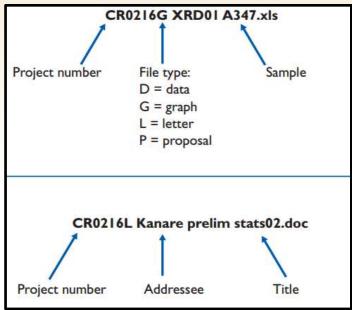
#### Electronic Data

- Audit Trails / Meta Data
- Transposition Errors
- Software Compatibility
- Program Updates
  - Automatic
  - Impact to significant digits

#### **Electronic Documents**

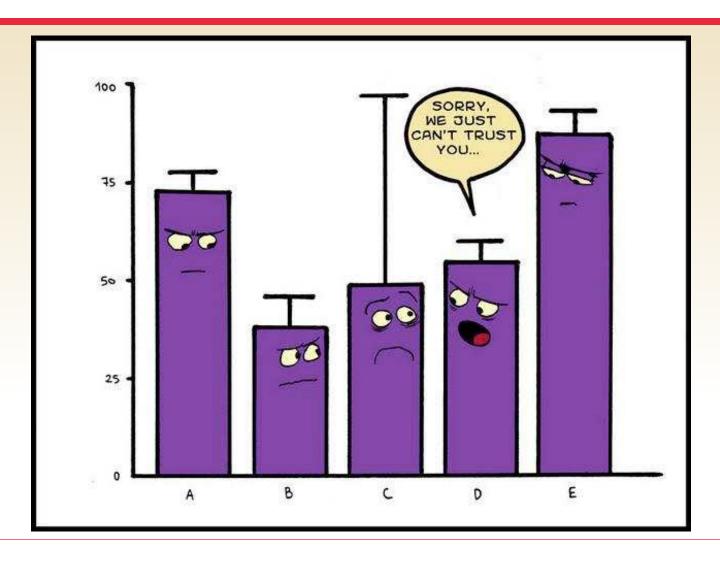
- Standard File Naming System
  - Brief, descriptive, consistent, dated
  - Plans for edits and changes





- > Fixed (No changes)
- Updated with no changes
- Updated with changes—Change Control

### Data Manipulation



# Mechanisms for Data Sharing

- > Email
- Online repositories
- Supplemental to publication
- Sharing agreements
- Mixed
- Conditions / Exclusions

# Promoting an Open Research Culture Data Sharing

Promoting an open research culture

Author guidelines for journals could help to promote transparency, openness, and reproducibility

#### Summary of the eight standards and three levels of the TOP guidelines

Levels 1 to 3 are increasingly stringent for each standard. Level 0 offers a comparison that does not meet the standard.

LEVEL 0 LEVEL 1 LEVEL 2 LEVEL

Data transparency

Journal encourages data sharing—or says nothing. Article states whether data are available and, if so, where to access them. Data must be posted to a trusted repository. Exceptions must be identified at article submission. Data must be posted to a trusted repository, and reported analyses will be reproduced independently before publication.

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#### Data Management Plan—NIH



National Institutes of Health Plan for Increasing Access to Scientific Publications and Digital Scientific Data from NIH Funded Scientific Research

February 2015

#### Data Management Plans

Ensure that all extramural researchers receiving Federal grants and contracts for scientific research and intramural researchers develop data management plans, as appropriate, describing how they will provide for long-term preservation of, and access to, scientific data in digital formats resulting from federally funded research, or explaining why long-term preservation and access cannot be justified. (OSTP memo, element 4b)

# Data Sharing Plan—Template

- > What
- > Who
- Where
- > When
- > How

https://www.niaid.nih.gov/research/sample-data-sharing-plan

Example Plan addressing Key Elements for a Data Sharing Plan under NIH Extramural Support (For questions, contact the NIH Office of Extramural Research (OER), Email Sharing@nih.gov)

Example Data Sharing Plan for FOA-XX-XXXX
What data that will be shared:  I will share phenotypic data associated with the collected samples by depositing these data at which is an NIH-funded repository. Genotype data will be shared by depositing these data at Additional data documentation and de-identified data will be deposited for sharing along with phenotypic data, which includes demographics, family history of XXXXXX disease, and diagnosis, consistent with applicable laws and regulations. I will comply with the NIH GWAS Policy and the funding IC's existing policies on sharing data on XXXXXXX disease genetics to include secondary analysis of data resulting from a genome wide association study through the repository. Meta-analysis data and associated phenotypic data, along with data content, format, and organization, will be available at Submitted data will confirm with relevant data and terminology standards
Who will have access to the data:  I agree that data will be deposited and made available through which is an NIH-funded repository, and that these data will be shared with investigators working under an institution with a Federal Wide Assurance (FWA) and could be used for secondary study purposes such as finding genes that contribute to process of XXXXXX. I agree that the names and Institutions of persons either given or denied access to the data, and the bases for such decisions, will be summarized in the annual progress report. Meta-analysis data and associated phenotypic data, along with data content, format, and organization, will be made available to investigators through
Where will the data be available:  I agree to deposit and maintain the phenotypic data, and secondary analysis of data (if any) at
When will the data be shared:  I agree to deposit genetic outcome data into repository as soon as possible but no later than within one year of the completion of the funded project period for the parent award or upon acceptance of the data for publication, or public disclosure of a submitted patent application, whichever is earlier.
How will researchers locate and access the data:

and presentations that I author or co-author about these data, as well as acknowledge the repository and

NIH-funded repository, this repository has policies and procedures in place that will provide data access to qualified researchers, fully consistent with NIH data sharing policies and applicable laws and

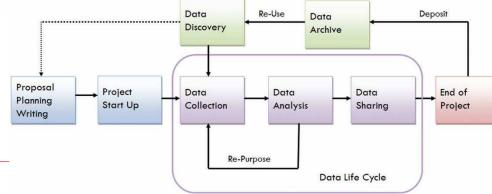
funding source in any publications and presentations. As I will be using

Rev 20100831

regulations.

#### Data Management and Sharing Plan

- ☐ Data Description / Types
- ☐ Data Standards for Format and Content
- Mechanisms for Access and Sharing (Provisions, Privacy Protection, confidentiality, Security, Intellectual Property, etc.)
- ☐ Provisions for Data Reuse and Redistribution
- ☐ Archiving / Long-term Preservation and access
- l Other...

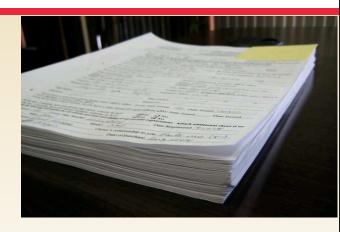


# Data Quality: GDP Principles

- > Attributable
- **L**egible
- > Contemporaneous
- Original
- > Accurate

#### **ALCOA+**

Complete, Consistent, Enduring, Readily Available





### Data Quality: Exercise

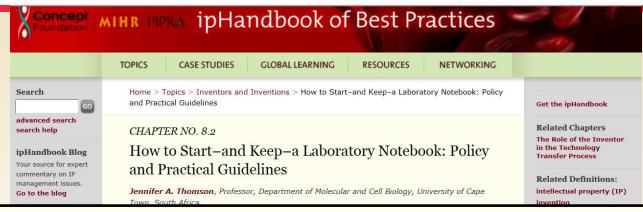


- ☐ Month / Day / Year
- □ Day / Month / Year





# Laboratory Notebooks



#### 1. What is a Laboratory Notebook?

Although you may think you will remember what you did and why you did a certain experiment in a week's time, YOU WILL NOT! And nor will anyone else in your laboratory. Hence the need for laboratory notebooks. In short, a laboratory notebooks is:

- · a daily record of every experiment you do, think of doing, or plan to do
- a daily record of your thoughts about each experiment and the results thereof
- · the basis of every paper and thesis you write
- the record used by patent offices and, in the case of disputes, courts of law (in the event you file patents on your findings)
- a record that would enable successive scientists, working on the same project, to pick up where you left off or reproduce your results

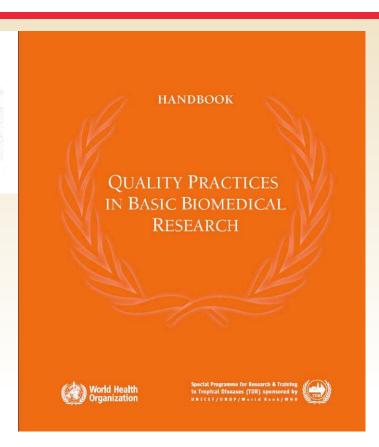
http://www.iphandbook.org/handbook/ch08/p02/

### Reconstructability

#### 4.3 Documentation

Making a full record of all information is essential not only to permit appropriate scientific interpretation of the results but also to enable complete reconstruction of the study, should this be necessary. Documentation is the only way of demonstrating what actually went on at the time of the experiment. Without documentation the process is meaningless; essentially there has been no study.

> "One group is attempting to reproduce the results of another group. The group is unable to reproduce the same results even after following the methods. After contacting the original author withheld crucial information out of the methods section."





# Closing Comment—Data Integrity

"When data are published, the authors must be fully responsible for its integrity, and thus must be prepared to discuss their findings and help investigators troubleshoot experiments when others are unable to reproduce their work..."





A Survey on Data Reproducibility in Cancer Research Provides Insights into Our Limited Ability to Translate Findings from the Laboratory to the Clinic

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#### Abstract

Background: The pharmaceutical and biotechnology industries depend on findings from academic investigators prior to initiating programs to develop new diagnostic and therapeutic agents to benefit cancer patients. The success of these programs depends on the validity of published findings. This validity, represented by the reproducibility of published findings, has come into question recently as investigators from companies have raised the issue of poor reproducibility of published results from academic laboratories. Furthermore, retraction rates in high impact journals are climbing.

Methods and Findings: To examine a microcosm of the academic experience with data reproducibility, we surveyed the faculty and trainees at MD Anderson Cancer Center using an anonymous computerized questionnaire; we sought to ascertain the frequency and potential causes of non-reproducible data. We found that ~50% of respondents had experienced at least one episode of the inability to reproduce published data; many who pursued this issue with the original authors were never able to identify the reason for the lack of reproducibility; some were even met with a less than "collegial" interaction.

May 2015, Vol. 8, Issue 5, e63221 www.Plosone.org



#### Case Study 1—Data Sharing

Your research study will include data from approximately 500 subjects being screened for three bacterial sexually transmitted diseases (STDs) at an inner city STD clinic. The final dataset will include self-reported demographic and behavioral data from interviews with the subjects and laboratory data from urine specimens provided. Because the STDs being studied are reportable diseases, you will be collecting identifying information.

Even though the final dataset will be stripped of identifiers prior to release for sharing, there remains the possibility of deductive disclosure of subjects with unusual characteristics. Identify options (i.e., conditions) for sharing the data.

#### Case Study 1—Data Sharing

The proposed research will include data from approximately 500 subjects being screened for three bacterial sexually transmitted diseases (STDs) at an inner city STD clinic. The final dataset will include self-reported demographic and behavioral data from interviews with the subjects and laboratory data from urine specimens provided. Because the STDs being studied are reportable diseases, we will be collecting identifying information.

Even though the final dataset will be stripped of identifiers prior to release for sharing, we believe that there remains the possibility of deductive disclosure of subjects with unusual characteristics. Thus, we will make the data and associated documentation available to users only under a data-sharing agreement that provides for: (1) a commitment to using the data only for research purposes and not to identify any individual participant; (2) a commitment to securing the data using appropriate computer technology; and (3) a commitment to destroying or returning the data after analyses are completed.

http://guides.lib.vt.edu/NIH/examples

#### Case Study 2—Data Retention

One of the important elements about rigor and reproducibility is knowing exactly how to "reproduce" conducting an experiment. This in turn depends upon access to primary records of research studies. Sometimes, it is difficult to obtain the primary lab/research records after investigators leave an institution.

What are best practices to insure that primary data and precise methodologies are maintained in an EASILY accessible format for required time periods and maybe even longer?



# (Homework): Watch the Data Management and Sharing Video

https://www.youtube.com/watch?v=N2zK3sAtr-4



# References – Data Management

https://dmptool.org/dm\_guidance (General guidance)

http://www.icpsr.umich.edu/files/datamanagement/DataManagementPlans-All.pdf (Guidelines)

http://www.icpsr.umich.edu/icpsrweb/content/datamanagement/dmp/plan.html (Sample plan)

http://lgdata.s3-website-us-east-

1.amazonaws.com/docs/2784/1033892/DM101Checklist101314.pdf (Checklist)

http://libguides.northwestern.edu/datamanagement/federalfundingagency (Other Federal funding agencies)

http://www.hhmi.org/sites/default/files/Educational%20Materials/Lab%20Management/Making%2 Othe%20Right%20Moves/moves2 ch8.pdf (Data Management and Lab Notebooks)

https://grants.nih.gov/grants/NIH-Public-Access-Plan.pdf (NIH Public Access Plan)

http://www.data-archive.ac.uk/media/2894/managingsharing.pdf (Managing and Sharing Data, UK Data Archive)

http://data.library.virginia.edu/data-management/lifecycle/ (Data lifecycle)



# References—Resource Sharing

https://grants.nih.gov/grants/policy/data sharing/data sharing guidance.htm (Policy and Implementation)

https://grants.nih.gov/grants/policy/data\_sharing\_faqs.htm (FAQs)

https://grants.nih.gov/grants/peer/guidelines general/Resource sharing plans.pdf

https://www.niaid.nih.gov/grants-contracts/resource-sharing-plan (Resource sharing)

#### Other References:

http://www.iphandbook.org/handbook/ch08/p02/ (Laboratory Notebook)

http://www.who.int/tdr/publications/documents/quality\_practices.pdf?ua=1 (WHO Handbook)

