

# Ideas for Citizen Science in Astronomy

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## Abstract

We review the relatively new, internet-enabled, and rapidly evolving field of citizen science, focusing on projects in astronomy and solar system science but sampling beyond those fields. In doing so, we look for ideas from which astronomy could benefit in the future. We consider contributions to science in the form of observations, data processing, data modeling and the design of new scientific inquiries, and from this experience derive some commonalities between the most successful projects, and use them to suggest some guidelines for future projects. The limits of citizen science are not yet well understood, but we make some speculations for citizen astronomy in particular.

**TO-DO: Phil: #65 Work conclusions into abstract**

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## 1 Introduction (2 pages)

The term “citizen science” refers to the activities of people who are not paid to carry out scientific research, but who make intellectual contributions to scientific research nonetheless.<sup>1</sup> They come from all walks of life, and their contributions

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<sup>1</sup>In this review we differentiate between the data analysis that citizens carry out themselves, and distributed “grid” computing farmed out to computers owned by citizens, and omit the latter since it does not fit our definition of citizen science as involving intellectual contributions

are diverse, both in type and research area. This review is about the science projects they have participated in to date, the tasks they have performed, and how astronomy has benefited – and could benefit further – from their efforts.

Citizen involvement in science pre-dates the profession itself; the earliest example of collaboration between professional and amateur astronomers seems to have been Edmund Halley’s call for observations of the 1715 total eclipse of the Sun which crossed central England (Halley 1716).<sup>2</sup> Since then there has since been a long and honourable tradition of amateur observers making important discoveries and significant sustained contributions. However, the advent of the world wide web has changed the face of professional and amateur collaboration, providing new opportunities and accelerating the sharing of information. People are now connected to each other on a scale that has never happened before. Professional scientists can interact with citizens via a range of web-based media, including on purpose-built citizen science websites which increase the potential for shared data analysis and exploration, as well as for data collection. Meanwhile, communities of citizens have sprung into existence as like-minded people have been able to find and talk to each other in a way that is almost independent of their geographical location. The result has been an exponential increase in citizen involvement in science. The field is evolving very quickly, with more and more professional scientists becoming aware of the possibilities offered by collaborating with, for

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from citizens.

<sup>2</sup>The aim was to refine estimates of the size of the shadow cast by the Moon, and citizen observations were much needed. Although Halley was successful in observing, his colleagues at Oxford were clouded out, and those in Cambridge were “oppressed by too much Company, so that, though the heavens were very favourable, [they] missed both the time of the beginning of the Eclipse and that of total darkness.”

example, specialists operating outside the usual parameters of professional astronomical observation, or tens of thousands of people eager to perform microtasks in their spare time.

Our aim in this work is to review the scientific literature as it stands for ideas implemented in citizen science projects, primarily in astronomy but occasionally in other fields, and then produce a summary of successful project characteristics for future research groups to learn from. As our title states, this is a review of ideas for astronomy. We will look forward as well as back, and try to answer the questions: How can the full potential of citizen science be realised in astronomy? What are the particular niches that citizen science can fill, in our field? How might it contribute to the solutions of the Big Data problem in astronomy?

This review is organised as follows. Astronomy research typically starts with observations: so do we, in Section 2. We then proceed to consider visual classification, data modeling and finally citizen-led enquiry in Sections 3–5. With this overview in place, we review in Section 6 the literature on, and the collected experience of, the population of citizens who have taken part, or are currently taking part, in scientific research. We then turn to the future, and speculate on how citizens might contribute to astronomy there (Section 7); in Section ?? we consider possible limits to citizen science, including challenges associated with data rates and volumes, data complexity, the difficulties of large-scale collaboration, and the barriers to accessibility. Finally, we make some concluding remarks in Section 8, and suggest some guidelines for the implementation of future citizen science projects in astronomy.

## 2 Amateur Observing (5 pages)

There is currently an active community of well-equipped amateur observers making astronomical observations of great utility. There are also many other citizens observing the night sky with less sophisticated equipment – and as we shall see, there are even citizens making astronomical observations almost inadvertently. What astronomical data are the citizenry taking, and what is it being used for?

### 2.1 Active Observing (3 pages)

**TO-DO: Leigh: #59 Shorten solar system parts of active observations.**

The steady improvements and increasing affordability of digital technology, in addition to the ease of data sharing and communications, have considerably expanded the realm of amateur astronomy in the past two decades. In this section, we review some of the citizen contributions to active observations of the night sky. Passive contributions, via data mining and web-based studies, will be described in Section 2.2 below.

How is active citizen observing beneficial to astronomy research? **TO-DO: Phil: We give three benefits to the pro community of amateur observing: time availability, flexibility and context. We should discuss this to make sure we have a complete picture.**

The first key advantage is time availability: professional observatories are always over-subscribed, with resources ending up being concentrated on one area of sky, or on a few astronomical questions. Such observations are rarely tuned to the optimum timescales for other scientific enquiries: for example, determinations of meteor frequencies on short timescales (minutes), or slow evolution of giant planets on longer timescales (years and decades). In contrast, amateur

observations can be frequent, repetitive, and long standing, and are naturally well sampled across the globe during an exciting event of interest.

The second, related, advantage is that of flexibility: whenever a new phenomenon is discovered (e.g., a new comet, or anything changing the appearance of the familiar planetary discs), observers will be keen to catch a glimpse irrespective of the scientific value of their observations. This reaction can be near instantaneous, compared to the need to allocate telescope resources among the professional community.

The third benefit is contextual. Professional observations often are taken in a very different wavelength range, focus on a narrower spatial region, or employ spectroscopic techniques that don't yield images. In some situations, near-simultaneous wide field optical imaging by citizen scientists can provide useful additional constraints on the process of interest.

The example case studies below serve to illustrate this synergy between amateur and professional observations, and also to highlight instances of professional-amateur ("Pro-Am") collaboration. While the solar system provides some of the most amenable targets for amateur observation, "deep sky" observations by the non-professional community provide important further insight into the capabilities and potential of citizen astronomers.

- *Discovering and Characterizing Asteroids and Comets.* Although survey telescopes provide the vast majority of modern solar system discoveries, citizen astronomers occupy some interesting observational niches, making discoveries in the places professional observatories can't reach. Targets include near-earth asteroids (NEAs, with orbits intersecting those of the terrestrial planets), main belt asteroids between Mars and Jupiter, and comets making their journey towards

the Sun from the outer solar system. A protocol for citizen discovery has been established: the position of any new object is compared to existing catalogues, and if no existing details are found then the new discovery and its ephemerides can be reported to the IAU Minor Planet Centre<sup>3</sup>. If observations are repeated for at least two nights by one or several observers, then a new denomination is provisionally assigned to the discovery. An electronic circular then reports the discovery to the wider world.

Small solar system objects moving against the fixed-star background can be detected in a set of CCD frames either by eye or by automated software. The extreme familiarity of citizen astronomers with a particular region, planet or nebula, allowing them to immediately identify peculiarities or new features (a good example of this is the study of meteorological activity on giant planets (?)). [For example?] the recent close flyby of asteroid 2012 DA14 on February 15th 2013 was initially reported by a team of amateur observers affiliated with the La Sagra Sky Survey at the Astronomical Observatory of Mallorca. **TO-DO: Leigh: add reference for 2012 DA14. How did they spot it? By eye or with automated detection software? If all the amateur detection is being done by eye we need to note that – and in any case this paragraph will need finishing.**

Some amateur observers have been involved in new cometary discoveries, including David Levy, part of the team that discovered Shoemaker-Levy 9 before its spectacular impact with Jupiter. **TO-DO: Leigh: add reference for Shoemaker-Levy?** As with asteroids, the majority of new comet discoveries are made by automated surveys, but a small and stable number of discoveries

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<sup>3</sup><http://www.minorplanetcenter.net>

come from amateurs with small telescopes, typically in regions poorly covered by survey telescopes (e.g., regions close to the Sun). C/2011 W3 Lovejoy, a Kreutz sungrazer comet, is one such example. C/2012 S1 ISON was spotted by A. Novichonok and V. Gerke in images from the International Scientific Optical Network.

**TO-DO: Leigh: add reference for C/2012 S1 ISON?**

Amateurs are also contributing to the search for a sub-category of objects with a detectable cometary coma within the asteroid belt. Recent discoveries of these Main Belt Comets, which appear to be asteroids that are actively venting their volatiles at perihelion, are beginning to blur the distinction between asteroids and comets. The T3 project, a collaboration between the University of Rome and several amateur observers, began in 2005 with the detection of a coma around asteroid 2005 SB216 (?), and has gone on to detect at least eight main belt comets (?). Comae may be detected as an extended FWHM of the asteroid compared to the background stars, or visually distinctive cometary comae.<sup>4</sup>

Citizen observations allow detailed characterisation of physical and orbital characteristics of these newly discovered solar system bodies, and amateur-led contributions are published online.<sup>5</sup> There is a need to continuously track Near Earth Asteroids as gravitational effects can significantly alter orbital trajectories over time, and this astrometry for both asteroids and comets can be provided to the Minor Planet Centre for collation and use in computing orbital trajectories.

Amateur monitoring of a comet's coma, dust and plasma tails and their photometric parameters can reveal dynamic structures and determine the locations of active venting regions. Photometric monitoring of comets over time of the

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<sup>4</sup>Recent discoveries of cometary activity can be found here:

<http://schiaparelli204.wordpress.com/>

<sup>5</sup><http://www.minorplanet.info/mpbdownloads.html>



magnitude and tail characteristics provides insights into the levels of activity as the comet moves along its trajectory, and can also reveal outbursts and other events associated with the outgassing. Photometric monitoring of an asteroid as it rotates provides information on the physical parameters, such as the shape, rotation rate and orientation. Observations of mutual encounters can also yield information on the density, and hence the composition of the asteroids. **TO-DO:**

**Leigh: Add refs for monitoring comets, asteroids etc.**

- *Monitoring the Planets.*

**TO-DO: Leigh: Please read this case study carefully, and please rearrange so that it makes sense: I worry that I have not merged the software and observing parts well. Could it be split into two or three, on order to better focus on the various techniques? Or, could it be compressed into a shorter version?**

Jupiter's size, brightness and colourful cloud contrasts make it an ideal target for citizen monitoring. Winds, waves, storms and plumes shape the day-to-day appearance, with cloud structures being blown east and west by powerful zonal jets. The jet velocities, and the dispersion of cloud material, reveals processes at work beneath the visible ammonia ice clouds, and serve as a probe of the fluid dynamics at work. Tracking the motion of features requires correlation of images obtained night after night; with filters sounding in and out of strong jovian methane absorption features providing a three dimensional view of the planet's cloud decks. Monitoring studies are also underway for Saturn, Uranus and Neptune, with increasing levels of difficulty.

The global distribution of giant planet observers permits global monitoring of Jupiter and Saturn as they rotate over 10 hours. Citizens upload raw filtered

images and colour composites, organised by date and time, to online servers, such as the Planetary Virtual Observatory and Laboratory (PVOL<sup>6</sup>) maintained for the International Outer Planets Watch (IOPW<sup>?</sup>). Those images can be used by amateurs and professionals alike to quantitatively study the zoology of activity; from measuring wind speeds (?); investigating the strength and changes to the large vortices (e.g., the 2006 reddening of Oval BA, ?); to determining the life cycle of the belt/zone structure (??).

Long-term monitoring exercises prove particularly fruitful in response to new and unexpected solar system events, including eruptions of gigantic plumes (??), changes to giant ovals (?) and impact debris produced by asteroidal and cometary collisions (?). An impact scar near Jupiter’s south polar region was first discovered in imaging by Australian amateur Anthony Wesley on July 19th, 2009, and led to an international campaign of professional observations to understand the asteroidal collision that had created the scar (e.g., ??).

Video monitoring has been used to enable high resolution “lucky” imaging of Jupiter: the best images at moments of clear seeing from the high-resolution video frames are selected, extracted and stacked together, using custom software to correct for the distortions associated with the telescope optics and residual atmospheric seeing. Software written by citizen scientists for free distribution to active observers, such as Registax<sup>7</sup> and Autostakkert<sup>8</sup>, allows them to process their own video files to search for impacts in an automated way (e.g., Jupiter impact detections<sup>9</sup> and LunarScan from the ALPO Lunar Meteoritic Impact

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<sup>6</sup><http://www.pvol.ehu.es/pvol>

<sup>7</sup>[www.astronomie.be/registax](http://www.astronomie.be/registax)

<sup>8</sup>[www.autostakkert.com](http://www.autostakkert.com)

<sup>9</sup><http://www.pvol.ehu.es/software/>

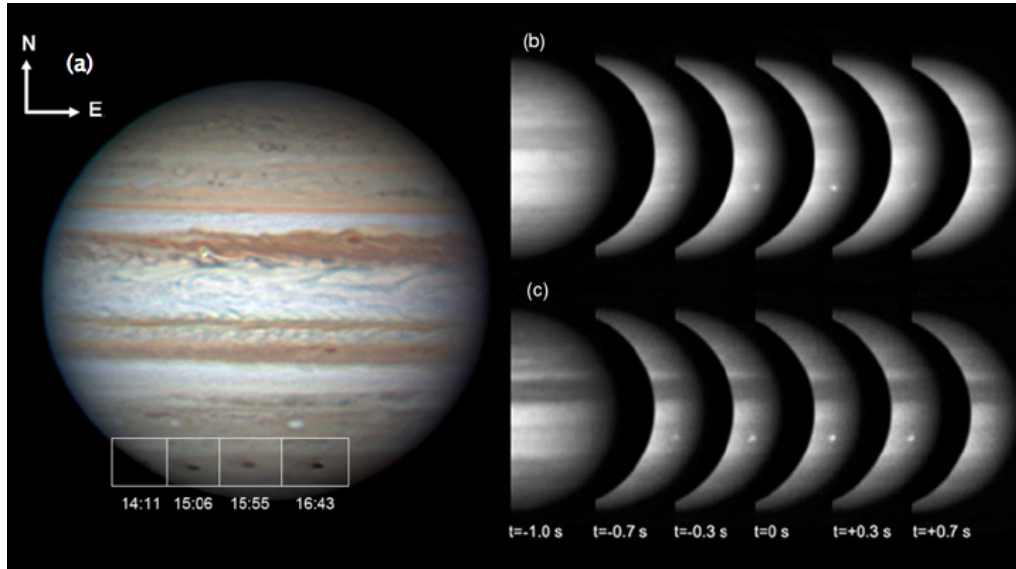


Figure 1: Citizen science contributions to monitoring of impacts in the Jupiter system. (a) Dark impact scar in Jupiter’s atmosphere imaged by Anthony Wesley on July 19th 2009 (?). (b) The evolution of a smaller bolide impact on June 3rd 2010 at red wavelengths, also imaged by Wesley. (c) The evolution at blue wavelengths by Christopher Go, figure from ?).

Search<sup>10</sup>), thus avoiding the need for transfer and storage of large datasets on some centralised server. The lucky imaging technique has the added benefit of providing long baseline time-sampled observations for impact flash detection. At least three flashes were confirmed between 2010 and 2012, and the light curves used to determine the sizes and frequency of objects colliding with Jupiter (e.g., ?) (Figure 1).

Descriptive records of morphological changes and events are maintained and continuously updated by organisations of citizen scientists such as the British Astronomical Association (BAA) and the Association of Lunar and Planetary

<sup>10</sup><http://alpo-astronomy.org/lunaryupload/lunimpacts.htm>

Observers (ALPO). The BAA’s Jupiter section<sup>11</sup> is a team of amateurs with substantial expertise in Jupiter’s appearance (?); Their regular bulletins describe the changing appearance of the banded structure, the emergence of new turbulent structures and weather phenomena, and keep a record of the long-term atmospheric changes.

Recent software developments have provided a much more quantitative angle on these observations. The WinJUPOS software<sup>12</sup> was developed by a team of amateurs led by G. Hahn and H.-J. Mettig. It allows multiple images to be stacked with a correction for the rapid (once per ten hour) rotation of Jupiter or Saturn, then reprojected onto a latitude-longitude coordinate system, so that the precise positional details of atmospheric features can be determined via “point-and-click.” By doing this over many nights surrounding Jupiter’s opposition, the team builds up enormous drift charts (tens of thousands of positional measurements) for features, ranging from the tiniest convective feature being moved by the jet streams, to the largest vortices. The positions can be extrapolated forward in time, enabling targeted observations by professional observatories or even visiting spacecraft. This long-term record of Jupiter’s visible appearance by citizen scientists has proven invaluable for jovian atmospheric scientists.

Saturn’s appearance is typically more subdued than that of Jupiter, but 20-30 cm diameter telescopes are capable of resolving small convective cloud activity. A close collaboration between amateurs and Cassini spacecraft scientists allows correlation of lightning-related radio emissions detected by the spacecraft with visible cloud structures on the disc (known as Saturn Storm Watch) (e.g., ?), which would not be possible with the targeted regional views provided by

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<sup>11</sup><http://www.britastro.org/jupiter/>

<sup>12</sup><http://jupos.privat.t-online.de/>

Cassini's cameras alone. This provides insights into the moist convective processes thought to power the dynamics of the giant planet, and is a good example of how citizen science can support an international planetary mission. Amateur observations of Uranus and Neptune are in their infancy and require telescopes with diameters exceeding 25 cm, but there have been confirmed reports of atmospheric banding and discrete cloud features when near-infrared filters are used to maximise contrast between white clouds and the background and long exposure times of tens of minutes.

Active citizen observing also provides long-term monitoring in the inner solar system. Venus' photochemical smog shields the planet's surface from view, but discrete cloud features can be used to study the super-rotation of the Venusian atmosphere and the occurrence of a mysterious ultraviolet absorber high in the planet's atmosphere (i.e., using near-UV filters). The Venus Ground-Based Image Active Archive was created by ESA to provide contextual observations supporting the Venus Express mission (?). Near-infrared imaging can be used to sample thermal emission from the Venusian surface on the nightside (?).

The Martian atmosphere, with its ephemeral clouds, seasonal CO<sub>2</sub> polar ice cycles and dust storms, continues to prove popular among citizen observers, although these typically supplement the wealth of high-resolution information being returned by orbital and surface missions to the red planet. As with other planetary targets, amateur observations provide the long temporal records for the evolution of atmospheric features. Groups such as the International Society of Mars Observers (ISMO<sup>13</sup>), the British Astronomical Association (BAA) and the International Mars Watch program quantitatively and qualitatively assess these

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<sup>13</sup><http://www.mars.dti.ne.jp/cmo/ISMO.html>

amateur images.

In Figure 2 we show some examples of planetary images obtained by the amateur community.

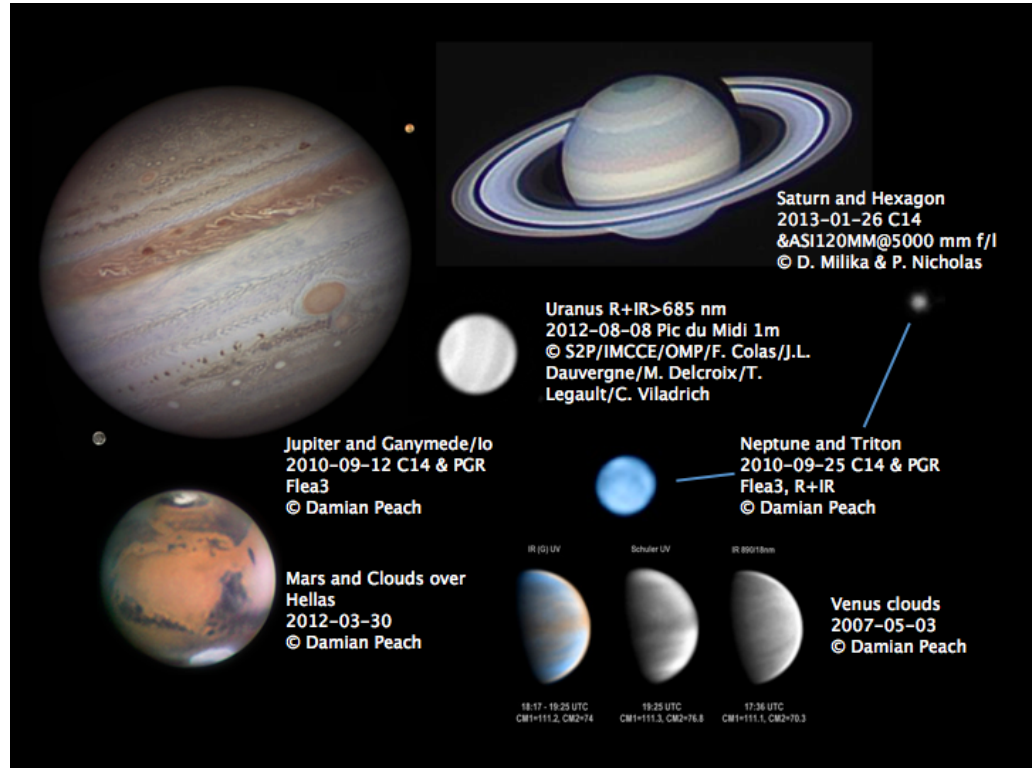


Figure 2: Examples of high fidelity images obtained by amateur planet observers.

- *The International Meteor Organisation: Distributed Meteor Detection and Tracking.* Closer to home, citizen scientists play a crucial role in the recording of rare and unpredictable events such as the fireballs from meteoroid impacts, such as the February 2013 Chelyabinsk meteor. Video footage of the fireball and shockwave were essential to scientifically characterise the impactor and its likely origins. **TO-DO: Leigh: Add refs for Chelyabinsk meteor strike.** These reconstructed trajectories even permit the recovery of meteorites from a strewn field (i.e., when the meteor survives the intense heat of atmospheric entry and reaches the ground). These objects are the remnant debris left over from the

epoch of planetary formation, and fragments left over from comets and asteroids, so their numbers, sizes and composition provide a window onto the earliest evolutionary stages of our solar system. The statistics of these impacts can only be obtained via a global network of enthusiastic citizen scientists, sharing and publicising their observations of meteors via the International Meteor Organisation (IMO<sup>14</sup>). **TO-DO: Phil: Provide high level comment on operation of IMO.**

- *Exoplanet Transits.* Beyond our solar system, amateurs have contributed to exoplanetary transit discoveries, attempting to measure the 1% diminution in starlight as a giant planet transits in front of its parent star. (?) points out three methods where amateurs can contribute to characterising exoplanetary systems ? (i) by frequent observations of known transits to refine ephemeris; (ii) searching for transit time variations that can reveal additional planets in a system; and (iii) searching for previously unidentified transits in known planetary systems (e.g., the discovery of the transit of HD 80606b from a 30 cm telescope near London, ?).

- *Nearby Supernova Detection.* **TO-DO: Chris: #19 Review nearby supernovae work.**

- *Variable Star Monitoring: the AAVS.* **TO-DO: Phil: #66 Review variable star monitoring, AAVS.**

- *John Johnson's Stellar Target Selection* . An example of an amateur expert collaborating with a professional team. **TO-DO: Phil: #67 Review Johnson et al's program.**

**TO-DO: Chris, Leigh, Phil: We could review how these various projects**

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<sup>14</sup><http://www.imo.net>

**accumulate monitoring data, and ask if this is scalable/repeatable.**  
**How does it compare to efforts in other fields, notably ecology?**

## 2.2 Passive Observing (1 page)

While amateur astronomers have acquired a great deal of very useful data, the general population is better equipped than ever to image the sky and make that data available for scientific analysis. This has been demonstrated by two recent professionally-led studies, that made use of a largely passive observing community connected via online social networks not usually associated with astronomy.

- *The Orbit of Comet Holmes from the Photographs Uploaded to Flickr.*   ?)

used more than 2000 images scraped from the photo sharing website Flickr as inputs to a reconstruction of the orbit of Comet Holmes. This comet was bright enough to be visible with the naked eye during its 2007 apparition, and a large number of photographs were taken of it and uploaded to the Flickr site. ? were able to astrometrically calibrate many of the images using their automatic image registration software, **astrometry.net**, which relies on background stars. This had been enabled as a Flickr “bot,” crawling over all images submitted to the **astrometry.net** group and sending the photos’ owners messages showing them where on the sky their images were taken. The time of observation could in many cases be derived from metadata included in the image headers. The calibrated images trace out the trajectory of the comet, producing a result which is close to that obtained from the JPL Horizons system (Giorgini et al. 1997). Estimates of orbital parameters from Flickr images alone are accurate, when compared to the Horizons values, to within a few standard deviations. As the authors point out, while in this case the photographers did not realize they were participating in



a scientific study, the potential of combining powerful calibration software with large amounts of citizen-supplied imaging data is made clear. This method of ‘unconscious’ citizen science may prove to have significant value in fields beyond astronomy too - if models of statistical sampling can be developed - with ecological surveys of images submitted to sites like Flickr likely in the next few years.

- *Detecting Meteor Showers with Twitter.* By saving a nightly (?) log of all tweets submitted to the web service Twitter, (?) were able to detect several new meteor showers simply by searching for the text string “meteor.” Unwitting naked-eye observers had spotted shooting stars and tweeted about them, giving rise to a detectable signal in the stream of tweets that night. The detected sample is incomplete/unlocalised/ etc... However, this work illustrates the potential both of Twitter as a communication system for connecting large numbers of observers with a science team, and of networks of unequipped observers for doing very bright object transient astronomy.

**TO-DO: Chris: #23 Condense Geert’s Twitter study, add meteor tracking app, move Chelyabinsk meteor mention to here**

### 3 Visual Classification (6 pages)

Observing the night sky with a telescope is perhaps the most familiar of the activities of amateur astronomers, but as the previous section showed, citizens are also actively involved in the processing and interpretation of the data they have taken. In this and the next section we look at projects where much larger archival datasets are made available to crowds of citizens, who are asked to inspect astronomical images, and help describe and characterize the features in them. Despite significant advances in machine learning and computer vision, the visual

inspection of data remains an important part of astronomy, as it continues to take advantage of the amazing human capacity for visual pattern recognition. While many in the 1990s predicted that the increasing size of astronomical datasets would make such time-intensive inspection impossible, the extensive reach of the world wide web has enabled the involvement of hundreds of thousands of citizen scientists in this form of “crowd-sourced” data analysis.

### 3.1 Visual Classification in Astronomy (3 pages)

- *Stardust@home* While significant preliminary work had been carried out by NASA’s clickworkers project (see below), the project that first illustrated the potential of crowd-sourcing for astronomical purposes was Stardust@home<sup>15</sup>. This effort, which asked volunteers to scan through images of samples returned from Comet Wild-2 by the Stardust mission, attracted a large audience to the apparently unprepossessing task of looking for dust grains in an effort to identify samples of material from outside our Solar System. The site was built on BOSSA, an early attempt to build a generalized platform for such crowd-sourcing projects (see next section), and featured a stringent test which had to be passed before classifications were counted. Despite this hurdle, more than 20,000 people took part and a variety of dust grains were removed from the aerogel for further study, at least one later proving an excellent candidate for an interstellar grain. Perhaps the most significant long-term impact of Stardust@home, though, was the demonstration that large amounts of volunteer effort were available even for relatively ‘unsexy’ tasks such as hunting dust grains which do not involve intrinsically beautiful images, and that, with a suitable website design and stringent

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<sup>15</sup><http://stardustathome.ssl.berkeley.edu/>

testing, scientifically valuable results could be obtained.

**TO-DO: Chris: #30 Add Stardust@home references, Stardust@home url as a footnote.**

- *Galaxy morphology with Galaxy Zoo* These results directly inspired the development of Galaxy Zoo, perhaps the most prominent scientific crowd-sourcing project to date. Galaxy Zoo was built on the continued importance of morphological classification of galaxies. First introduced in a systematic fashion by Hubble, and later developed by amongst others de Vaucouleurs, it remains the case that the morphology of a system is closely related to – but not entirely defined by – parameters such as colour, star formation history, dynamics, concentration and so on. In an effort to prepare for large surveys such as the Sloan Digital Sky Survey (SDSS), Lahav et al. (1995) and Lahav et al. (1996) followed especially by the work of Ball et al. (2004) developed neural networks trained on small samples of expert classifications in order to automate the process of classification,<sup>16</sup> arguing that the size of the then up-coming surveys left no place for manual classification.

**TO-DO: Chris: #32 Add real refs for automatic morphology classification**

The performance of the automatic classifiers depended on the input parameters, including colour, magnitude and size. These variables correlate well with morphology, but are not themselves morphological, and when included they dominate the classification. In particular for galaxies which do not fit the general trends, such as spirals with dominant bulges or star-forming ellipticals, automated classifiers whether using these simple measures or more complex proxies

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<sup>16</sup>The Lahav papers are perhaps as interesting for their psychology as for their astrophysics, as the classifications reveal the relations between the senior classifiers employed to be experts

for morphology such as texture fail to match the performance of expert classifiers. Schawinski (Schawinski et al. 2007) and Nair (Nair & Abraham 2010), amongst others, spent substantial time during their time as graduate students classifying tens of thousands of galaxies.

Inspired by Stardust@home, a group led by one of the authors (Lintott) created Galaxy Zoo in 2007 to provide basic classifications of SDSS galaxies<sup>17</sup>. Classifiers were presented with a coloured image centered on and scaled to one of more than 800,000 galaxies, and could select from one of six options: clockwise, anti clockwise and edge-on spirals, ellipticals, mergers and “star/don’t know.” Aside from an easily-passed initial test, little knowledge was required or indeed presented to classifiers, enabling them to proceed quickly doing something real soon after arriving at the site; this approach, in contrast to Stardust@home where a difficult test needed to be completed before authentic and useful contributions could be made, was successful in encouraging large numbers of visitors to participate. This tactic – in which both passing and sustained engagement provide substantial contribution – is illustrated in Figure ?? which shows results from Galaxy Zoo 2. **TO-DO: Chris: #34 Insert box diagram illustrating Zooniverse tactics / GZ2 results.** This later version of the project asked for more detailed classifications via a decision tree containing questions such as ‘How prominent is the bulge?’, and later iterations of the project have applied a similar approach to galaxies drawn from Hubble Space Telescope surveys including GEMS, GOODS, COSMOS and CANDELS.

**TO-DO: Chris: #35 Add galaxy zoo participation statistics.**

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<sup>17</sup>The original Galaxy Zoo is preserved at <http://zoo1.galaxyzoo.org> with the current incarnation at <http://www.galaxyzoo.org>.

These figures are undoubtedly impressive, but they would be meaningless if the classifications provided were not suitable for science. With sufficient effort to ensure each galaxy is classified multiple times (as many as 80 for many Galaxy Zoo images), these independent classifications need to be combined into a consensus. As discussed in later sections, this can become complex but for Galaxy Zoo a simple weighting which rewards consistency, first described in ?), was sufficient.

**Phil: By what criteria was it sufficient?**

**TO-DO: Chris: #68 Check consistency between sections: there are lots of “see below” type statements in this section.**

Importantly, combining classifications provides not only the assignment of a label but, in the vote fraction in a particular category, an indication of the reliability of the classification. This allows more subtle biases, such as the propensity for small, faint or distant galaxies to appear as elliptical regardless of their true morphology, to be measured and accounted for (see Bamford et al.). The net result is that the Galaxy Zoo classifications are an excellent match for results from expert classification, and have produced science ranging from studies of red spirals (Masters et al.) to investigations of spiral spin (Slosar et al.).

**TO-DO: Chris: #36 Let’s back up the accuracy claims here with some citations and numbers.**

A full review of Galaxy Zoo science is beyond the scope of this review; a recent summary is given in the Galaxy Zoo 2 data release paper by Willett et al. However, it is worth noting that many of the project’s most important results have been the result not of interaction with the main interface but represent rather serendipitous discoveries made by participants. We return to these in Section 5 below.

- *Surfaces of solar system bodies: Moon Zoo, Moonwatch.* If studying galaxies remains, at least in part, a visual pursuit, then the same is certainly true of planetary science. NASA’s clickworkers, which asked volunteers to identify craters on the Martian surface, lays claim to be the oldest astronomical crowd-sourcing project. The consensus results matched those available from experts at the time, but failed to go beyond this promising start to produce results of real scientific value. More recently, interfaces inviting classifiers to look at the Moon, Mercury, Mars and Vesta have been launched and attracted significant numbers of classifications, but although preliminary results have been promising these projects have yet to produce datasets that have been used by the planetary science community in the same way that Galaxy Zoo has by the astronomical community.

**TO-DO: Chris: #38 Fact-check the consensus with experts claim about clickworkers, provide clickworkers reference.**

**TO-DO: Chris: #69 Tidy up the multiple mentions of clickworkers. Whats the optimal ordering of these sections given their inter-references?**

- *Time domain astronomy: Supernova Zoo and Planet Hunters* (?) describe the three ways – volume, velocity and variety – in which the analysis of “Big Data” can be challenging. Time-domain astronomy projects, that require the immediate inspection of challenging volumes of live, high velocity, complex data, can benefit from citizen science, as shown in these two recent projects. While transients such as supernovae or asteroids can often be found through the use of automatic routines, visual inspection is still used by many teams as part of the process of selecting candidates for follow-up.

**TO-DO: Chris: #70 Provide citation for “3 V’s” of big data**

The most successful attempt to use crowd-sourcing to attack these problems to date has been the offshoot of Galaxy Zoo described in Smith et al. (2011) “Supernova Zoo.” Data from the Palomar Transient Factory (Law et al. 2009) was automatically processed and images of candidate supernovae uploaded on a nightly basis; this triggered an email to volunteers who, upon responding were shown the new image, a reference image and the subtraction between the two. By analyzing the answers to a series of questions, candidates were sorted into three categories, roughly corresponding to “probable supernova,” “likely astrophysical but non-supernova transient” and “artifact.” The results were displayed on a webpage and used to select targets for follow-up. Despite the site attracting many fewer classifiers than the main Galaxy Zoo, it was highly effective in sorting through data, with consensus typically reached on all images within 15 minutes of the initial email being sent. The large dataset generated by this project was used by Brink et al. (2013) to develop a supervised learning approach to automatic classification for PTF transients. The performance of this routine, which for a false-positive rate of 1% is more than 90% complete, depends on the kind of large training set that can be provided by citizen science and suggests a future path for large surveys in which citizen science provides initial, training data followed by machines taking on much of the work. Encouragingly, the routine, which makes use of a set of 42 features extracted from survey images, has performance which is insensitive to a small fraction of mislabeled training data, suggesting that the requirements for accuracy of citizen science projects which aim to calibrate later machine learning may be less stringent than otherwise thought.

A different approach to time-domain astronomy through citizen science is exemplified by the Planet Hunters project, which asks volunteers to examine light

curves drawn from the dataset provided by the Kepler mission in order to identify interesting events in retrospect. While the task of identifying transits from extra-solar planets is, at first glance, one which seems more suited for automated than for human analysis, the success of Planet Hunters in identifying more than fifty planet candidates missed by the automatic routines suggests that there remains a role for inspection by eye in cases where the relevant science requires completeness of classifications. While several of the planets found by Planet Hunters, including the project's first confirmed planet (PH1b, Schwamb et al.), are unusual – PH1b, a circumbinary, is the first planet known in a four-star system – others, including the more than forty candidates identified in the habitable zone of their parent star by Ji et al., might have been expected to be recovered by more conventional searches. Planet Hunters, therefore, is acting as an independent test of the Kepler pipeline's efficiency (Schwamb et al.) and has inspired improvements in subsequent analysis (Bathalia et al.). More recently, advanced users have participated in the identification of unusual stellar properties - see XXXXXXXX.

**Question for Leigh:** Are there any visual classification activities going on in solar system science that belong in this section rather than anywhere else? They needn't be based around purpose-built websites like Galaxy Zoo, but if the product of them is visual classification data we might include them here for contrast. I guess there is visual classification going on in the rapid-reaction events (jovian/lunar impacts, storm/plume eruptions) described in the observing section – but how are those classifications collected and analyzed? Is there any systematic analysis of the crowd's outputs?

**Question for Leigh:** I also have a note here that says “Data mining for



asteroids and TNOs” - know anything about that?

### 3.2 Classification Analysis

In most visual classification projects, working on archived image data with little time pressure, the random assignment of task to classifier, followed by simple, democratic treatment of the classifications has been judged sufficient. However, the need for rapid processing of images in time domain astronomy projects has prompted the investigation of more efficient analysis of the classification data. Using the supernova project’s archive as a test, (?) developed a Bayesian method, IBCC, for assessing classifier performance; in this view, each classification provides information both about the subject of the classification and about the classifier themselves. Classifier performance given subject properties can thus be predicted and an optimum set of task assignments calculated. For systems involving tens of thousands of classifiers and perhaps a similar volume of subjects to be classified, an exact solution is computationally extremely expensive, but a MCMC approach can be used to find efficient solutions. Work by Simpson et al., as well as Horvitz et al and Waterhouse on Galaxy Zoo data, suggests that accuracy can be maintained with as few as 30% of classifications.

This sort of optimization will be increasingly important for online citizen science, and especially in projects that use a live stream of data, rather than an archive, since the classification analysis will need to be done in real time. Analysis strategies which require re-optimisation of the entire subject and agent model using all past classifications will likely be prohibitively expensive, and so the focus will need to be on “progressive” strategies that approximate the full inference.

- *Rare event detection: Space Warps* Steps towards real-time classification

analysis have been taken in the Space Warps project. Space Warps is a rare object search: volunteers are shown deep sky survey images and asked to mark features that look as though they are gravitationally lensed galaxies (Marshall et al, in prep). Extensive training is provided via an ongoing tutorial that includes simulated lenses and known non-lenses, and immediate pop-up feedback as to whether these training images were correctly classified. Because real lenses are rare (appearing once every  $10^{2-4}$  images, depending on the dataset), the primary goal is to reject the multitude of uninteresting images so that new ones can be inspected – and this drives the need for efficiency. Marshall et al (in preparation) derived a simplified version of the IBCC classification analysis that updates a probabilistic model of both the subjects and the agents that represent the classifiers in a statistically online manner (Marshall et al, in prep). This analysis was run daily during each of the Space Warps projects, and subjects retired from the stream as they crossed a low probability threshold. This algorithm is being implemented into the web application itself for future datasets.

The increased efficiency of visual classification projects that will come with real-time analysis will enable feedback on the projects' progress to be given much more promptly – an important part of the collaboration between professionals and amateurs in crowd-sourcing projects.

### **3.3 Visual Classification in Other Fields (Chris, Phil, 1 page)**

Although, as described in the previous section, astronomical analysis led the development of citizen science as a data analysis tool, it has quickly been adopted by other fields. In some cases, this adoption has been explicit. The tools developed for Stardust@home were developed into a general purpose library for citizen

science, BOSSA, which has recently been ported to python as PyBOSSA, and the Zooniverse platform which supports many of the examples described above includes projects from fields as diverse as ecology and papyrology. This diversity of projects allows general lessons about project design to be drawn; indeed, this is an active area of research for academic fields as diverse as computer science, economics and social science.

**TO-DO: Chris: #73 Add citation to Crowston paper on PH and Seafloor Explorer**

Projects from other fields can also suggest strategies which could be adopted by future astronomical projects. In particular, future projects involving analysis of survey data which has been collected for a multitude of purposes may require a more sophisticated model for data analysis than the simple decision tree presented by projects such as Galaxy Zoo.

- *Snapshot Serengeti* This Zooniverse project enables the analysis of images from more than two hundred motion-sensitive “camera traps,” offers a particularly sophisticated response. Driven in part by the need for an interface which allows volunteers to state the obvious (for example, identifying elephants, lion or zebra) and also to provide more obscure classification (for example, distinguishing between types of gazelle), a variety of classification paths are presented. In addition to just clicking buttons identifying species, volunteers can opt for a decision tree-like approach, or choose from a variety of similar species (“Looks like an antelope...”) or search the descriptions provided (“Show me all animals whose descriptions involve ‘ears’ ”). This hybrid model has proved successful not only in encouraging classification, but also in encouraging learning; over a Snapshot Serengeti classifier’s “career” they are increasingly likely to chose more

direct routes.

**TO-DO: Chris: #43 Move SS to GZ section endnote? Include Snapshot Serengeti citations**

**Phil: Is the SS experience relevant to open ended exploration of survey images? I've wondered for a while whether diagnosis of astronomical image reduction problems could be done by crowd-sourcing: DES do a lot of visual inspection just within their team, LSST could serve a larger volume to more people, potentially. The viewer ubrets with their shareable URLs seem vital for this; might SS-style interfaces also be important?**

Another aspect of project strategy, and design, relates to the engagement of the volunteers. The online citizen astronomy projects developed so far have tended to emphasise co-operation between volunteers, and the results being due to a team effort. Elsewhere, experiments with a more competitive approach to citizen science, “gamifying” the activity, have been performed.

- *Visual inspection of 3-D biological scans: Eyewire.* This project, based at MIT, seeks to supplement machine learning identification of neurons in three-dimensional scans. Notably, this project incorporated some “gamified” elements into its design. Participants in the project, who are asked to identify connected regions throughout a three-dimensional scan, earn points based on participation and also have a separate, publicly visible, accuracy score. In addition to overall leader boards, the project also runs short challenges including a regular Friday “happy hour” in which participants compete on specific problems. Eyewire is also notable for its other strong community elements, with a chat room open and available to all participants in the project (supplemented, incidentally, by a “bot”

built by a participant which answers frequently asked questions from new users).

#### 4 Data Modelling (4 pages)

New understanding of the world comes from the interpretation (fitting) of data with a physical model. Such “data modelling” often involves technical difficulties that computers may find hard to overcome, associated with complex and/or computationally expensive likelihood functions. Humans, by applying their developed intuition, can often contribute a great deal to the exploration of a model’s parameter space by closing in quickly on the model configurations that fit the data well. This process can be particularly satisfying, rather like solving a puzzle. How have citizen scientists been involved in model making and data fitting in astronomy, and other fields, to date?

- *The Milky Way Project* Simpson et al. (2012) provided volunteers with a fairly flexible set of annulus-drawing tools, for annotating circularly-symmetric “bubble” features in colour-composite (24.0, 8.0 and  $4.5\mu\text{m}$ ) infrared images from surveys carried out by the Spitzer space telescope, which are hypothesised to have been caused by a recently-formed high mass star at the centre of the bubble. The (bubble) model in this case is simple and recognisable, making both the interface constructions and its operation relatively straightforward. The large sample of bubble models have been used to investigate the possibility of further star formation being triggered at the bubble surfaces (Kendrew et al. 2012). A subsequent effort (Beaumont et al. 2014) used data provided by the project to train a machine learning algorithm, Brut, in bubble finding. Brut is able to identify a small number of sources which are excluded from the Simpson et al. catalogue, and which are difficult for humans, specifically bubbles near bright

sources which have low contrast relative to their surroundings. In addition, Brut proved effective at identifying suspect bubbles included in the previous surveys. Given the relatively small size of the MWP sample, the main use of machine learning is to contribute an independent check on the data; for larger samples, as discussed below, an approach in which machine learning trains on citizen science data and gradually takes over the classification task could be considered.

- *Modelling Lens Candidates* The Space Warps project<sup>18</sup> has an informal data modeling element. The classification interface is restricted to enabling identification of candidate gravitationally-lensed features, but all the images are available via the project’s discussion forum. A small team of volunteers (including several citizens who helped design the project) has engaged in modeling the identified lens candidates using web-based software developed and supported by the project science team.<sup>19</sup> Results from a small test programme show that the model parameters derived by the ensemble of citizens are as accurate as those derived by experts (?). A pilot collaborative modeling analysis was carried out and written up by a small group of Space Warps volunteers<sup>20</sup> (Capella\_05 2014).

**TO-DO: Phil: #64 Provide URLs for spaghettilens and modeling thread on Talk.**

- *Galaxy Zoo: Mergers* This has been perhaps the most advanced attempt at data modeling in astronomical web-based citizen science (Holincheck et al. 2010, Wallin et al. 2010). **TO-DO: Chris: #49 Check through the merger zoo case study, ensure refs are up to date.** Here, simple N-body simulations

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<sup>18</sup><http://spacewarps.org>

<sup>19</sup><http://mite.physik.uzh.ch>

<sup>20</sup>See <http://talk.spacewarps.org/#/boards/BSW0000006/discussions/DSW00008fr> for the forum thread that was used.

of galaxy mergers were performed in a java applet, and the results selected according to visual similarity to images of galaxy mergers (previously identified in the Galaxy Zoo project). A key proposal in this project is that the inspectors of the simulation outputs would be able to find matches to the data more readily than a computer could, for two reasons. First is that humans are good at *vague* pattern matching: they do not get distracted by detailed pixel value comparisons but instead have an intuitive understanding of when one object is “like” another. The second is that initialising a galaxy merger simulation requires a large number of parameters to be set – and its this high dimensionality that makes the space of possible models hard to explore for a machine. Humans should be able to navigate the space using their intuition, which is partly physical and partly learned from experience gained from playing with the system. Initial tests on Arp 86 showed the crowd converging on a single location in parameter space, and that the simulated mergers at this location do indeed strongly resemble the Arp 86 system. The authors have since collected thousands of citizen-generated models for a sample of a large number SDSS merging systems (Holincheck et al, in preparation).

- *Online Data Challenges* A very interesting case of citizens being involved in astronomical data modelling is that of the analysis challenges organised by members of the professional astrophysics community. The measurement of weak gravitational lensing by large scale structure (“cosmic shear”) relies on the measurement of the shapes of distant, faint galaxies with extreme accuracy. The STEP (Heymans et al. 2006, Massey et al. 2007) and GREAT (Bridle et al. 2010, Kitching et al. 2012a, 2013) blind galaxy shape estimation challenges have had an enormous impact on the field, revealing biases present in existing techniques,



Figure 3: Examples of image modeling in web-based citizen science projects. Top row: star formation “bubble” identification and interpretation in Spitzer images in the Milky Way Project, with the annotation interface shown on the left, and some example (selected, averaged) bubbles on the right. Images from Simpson et al. (2012). Bottom row: matching N-body simulated merging galaxies to SDSS images in the Galaxy Zoo Mergers project (left), and exploring parameter space two parameters at a time to refine the models (right). Screenshots from Holincheck et al. (2010).

and providing a way for researchers outside the world of professional cosmology to participate. In particular, the GREAT08 challenge saw very successful entries (including the winner) from two (out of a total of 11) teams of researchers from outside of astronomy (albeit still professional researchers). A companion, somewhat streamlined galaxy shape measurement challenge, “Mapping Dark Matter,”



which was hosted at the Kaggle website<sup>21</sup> (Kitching et al. 2012b). The wider reach of this platform led to over 70 teams making over 700 entries to the competition; many of the teams did not contain professional astronomers, although most were still from academia.

In a comparison with the GREAT challenges, the Kitching et al. found a factor of several improvement in shear accuracy over comparable previous challenges, and suggested two interesting explanations for this success. First, the challenge was designed to be as accessible as possible, with an extensive training set of data that needed very little explanation; in this way the challenge was geared towards *idea generation*. Second, they noted that the competitive nature of the challenge (a webpage leaderboard was updated in real time as entries were submitted) seemed to stimulate the analysts into improving their submissions. Kaggle offers cash prizes, which will have had some effect as well (the pot was \$3000 for this challenge, even if indirectly).

Two more astronomical Kaggle challenges have since been set. The first involved inferring the positions of dark matter halos based on their weak lensing effects (“Dark Worlds,”<sup>22</sup> Harvey et al, in prep.) This challenge attracted the attention of 357 teams, perhaps due to its larger prizes. It also sparked some debate in its forums as to the design of the challenge: the models used to generate the data, the size of the test datasets (and consequent stability of the leaderboard), the choice of leaderboard metric and so on. These issues are also of generic importance for scientists looking to crowd-source algorithm development. It is interesting to note that the Kaggle forums are a useful resource for the Kaggle development team: the citizens who are active there do influence the

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<sup>21</sup><http://www.kaggle.com/c/mdm>

<sup>22</sup><http://www.kaggle.com/c/DarkWorlds>

design of the site infrastructure and challenge rules (D. Harvey, priv. comm.).

The most recent Kaggle astronomy challenge was to reproduce the Galaxy Zoo crowd-sourced galaxy morphologies, given the same color images. XXX teams entered this challenge ... (K. Willett, priv. comm.).

**TO-DO: Phil: #52 Update Kaggle challenge references – Harvey paper. Update Kaggle challenge with Galaxy Zoo. Ask Kyle for spiel.**

- *Protein Modelling with Fold.It*    Gamification therein.

**TO-DO: Phil: #62 Review Fold.It, discuss gamification.**

## 5 Citizen-led Enquiry (3 pages)

The previous sections have focused on specific, and isolated, activities in which citizens have participated. In most cases, the community's involvement has been a *contribution* to a scientific investigation, while not being involved in the design of that investigation. The most important part of any scientific investigation is the question at its heart: what is it we are trying to find out about the universe? In this section we look at some cases where the process of enquiry, the science, has been instigated or led by citizens.

In principle, this is an area of great potential. The constraints of funding proposals and management of research groups can often mean that professional scientists focus very narrowly on particular topics of research, using a particular technique for which they become known. Steering away from this course implies taking risks with time management, and allocation of resources to an ultimately fruitless research area can be detrimental to careers. Citizen scientists are largely free of these managerial and budgetary constraints, and are able to devote their attentions to whatever topics interest them: we might expect out-

siders to ask some unusual questions, and make connections and suggestions that highly focused professionals may not have thought of. What are some enquiries that citizens have led in astronomy to date, and how have they been enabled and supported?

- *Saturn Storm Watch.* In this project, Cassini’s observations of lightning emissions are connected with active amateur observations of convective cloud structures within the giant planet atmosphere; and the tracking of the vertices of Saturn’s bizarre north polar hexagon (?), a 6-sided planet encircling wave that has persisted for at least 30 years but that has only recently been observed, by amateur astronomers. In the first case, citizen scientists wished to identify the source of Saturn’s radio emissions. In the latter case, the long-term evolution of the hexagon vertices is being used to understand what sort of wave this is, and to identify its origins.

- *The Galaxy Zoo Forum.* The best known serendipitous discovery emerging from the Galaxy Zoo project is “Hanny’s Voorwerp” (Lintott et al.), a galaxy-scale light echo which reveals a recent ( $\sim 100000$  years) shutdown of AGN activity in IC 2497, the neighboring spiral galaxy. The discovery of the Voorwerp was first recorded in the Galaxy Zoo forum a few weeks after the project started, and inspired a more systematic search for similar phenomena in other galaxies. This project, made possible by the deep engagement of Galaxy Zoo science team member Bill Keel in the forum community, succeeding in finding more than forty instances of clouds which appear to have been ionized by AGN activity, in systems a third of which show signs of significant drops in activity on a timescale of tens of thousands of years. **TO-DO: Chris: #63 Add more citations for voorwerp science?**

The ability of the Zoo volunteers to carry out their own research, moving far beyond the mere “clockwork” required by the main interface, is best illustrated by the discovery of the Galaxy Zoo Peas. These small, round and, in SDSS imaging, green systems are dwarf systems with specific star formation rates (SFR per unit mass) which are unprecedented in the local Universe, matched only by high-redshift Lyman-break galaxies. Volunteers not only identified these systems, but organized a systematic search and further review of them, including using tools designed by SDSS for professional astronomers to acquire and study spectral data.

The discovery of the Peas marked the first time the Galaxy Zoo team realized the potential of the community of citizen scientists the project had acquired, but it is important to note that the simpler, initial interaction provided by the main interface was necessary in order to develop that community in the first place. The participants in the citizen scientists’ investigation of the Peas did not arrive on the site wanting to dig into spectra or confident of their ability to do so; these were the results of their participation. The project acted as an “engine of motivation” in inspiring its participants to become more involved.

- *Galaxy Zoo: Quench.* Participants in citizen science projects like Galaxy Zoo and Saturn Storm Watch are highly motivated but geographically widely distributed; the impact of the internet in enabling them to find each other and collaborate is remarkable. The most recent Zooniverse astronomy project, Galaxy Zoo: Quench, is actively encouraging citizen-led enquiry, by providing flexible tools for investigating not individual objects but samples of galaxies classified in the main Galaxy Zoo project. It is an experiment in citizen-driven scientific investigation, and we await its results with interest.

**TO-DO: Chris: #55 Write up Galaxy Zoo Quench...**

- *Lightcurve analysis on Planet Hunters Talk.* The data modelling examples of the Section 4 all involved modeling infrastructure provided by either the project's developers or their science teams. Planet Hunters is a case where citizens have carried out their own modeling analysis, using their own tools. A small group [?] of volunteers downloaded full Kepler lightcurve datasets for the best [?] community-selected candidates, and fitted transit lightcurves to them using [?].

**TO-DO: Chris: #51 Add text to PH talk case study: What did the super-users actually do in the way of advanced lightcurve modeling? Did it make it to the paper? References. This seems to me like something of a triumph, we must give it a good airing!**

## 6 Understanding the Citizens (2 pages)

Having surveyed some of the activities involving citizen scientists, we can now consider some questions about this community itself. Who participates in citizen science, and what motivates them?

### 6.1 Demographics

Who is participating in citizen astronomy? We might expect the demographics to vary with activity, and with the level of commitment required. We have some understanding of at least the former division from two studies that were carried out approximately simultaneously, one of the community participating in Galaxy Zoo, and another of the American Association of Variable Star Observers (AAVSO). (?) surveyed the Galaxy Zoo volunteer community to investigate their motivations (Section 6.2 below), via a voluntary online questionnaire. The 11,000

self-selected Galaxy Zoo users identified as 80% male, with both genders having an approximately uniform distribution in age between their mid-twenties and late fifties. The authors point out that this is close to the US internet user age distribution, except for slight but significant excesses in numbers of post-50s males, post-retirement people of both genders, and a deficit in males under 30. The survey respondents also tended to be more highly educated than average US internet users, with most holding at least an undergraduate degree, and around a quarter having a masters or doctorate.

These findings can be compared with a survey of the members of AAVSO: Price & Paxson (2012) received over 600 responses (corresponding to about a quarter of the community of observers and society members). The education levels of the AAVSO respondents matches the Galaxy Zoo community very closely; the AAVSO age distribution is more peaked (in the mid fifties), with a similar post-60 decline but also a marked absence of younger people. The online nature of the Galaxy Zoo project seems to have increased the participation of younger (pre middle-age) people. Likewise, the Galaxy Zoo gender bias, while itself extreme, is less so than at AAVSO, where some 92% of survey respondents were male. One additional piece of information provided by the AAVSO survey is the profession of the variable star observers: most (nearly 60%) of the survey respondents were found to be working in science, computer science, engineering and education.

The Galaxy Zoo and AAVSO communities differ by more than just the nature of their activity. The smaller AAVSO community is arguably more engaged in its research, in the sense that a larger fraction of its membership is active in taking observations and contributing to analyses. It would be very interesting to know how citizen scientist motivation varied with the level of participation:

dividing the Galaxy Zoo community into volunteers that contribute to the forum and those who don't could be interesting; perhaps more so would be to repeat the analysis of Raddick et al. over a wide range of projects, and look for trends there. The emergent picture thus far, however, is of a well-educated (and often scientifically trained) but male-dominated citizen science community, whose female and younger membership is likely to have been, at least in part, enabled via projects being hosted online. Continuing to lower the barriers to entry for currently under-represented demographic groups would seem both important, and within reach.

## 6.2 Motivation

**TO-DO: Chris: #45 Read the motivation section to see if it works well for discussing motivation and gamification the way you want to.**

What motivates citizen scientists? The two demographic studies referred to above also covered this question; it was the primary motivation for Raddick et al.. Having previously identified 12 categories of motivation in an earlier pilot study (Raddick et al. 2010), Raddick et al. asked the 170,000 volunteers at the time to comment on how motivated they were by each of these categories, and which was their primary motivation. The 6% who responded gave consistent answers to around 900 forum users who responded in a separate appeal, allowing us to draw conclusions about this presumably more engaged sub-population. A desire to *contribute* to science was found to be the dominant primary motivation, being selected by 40% of respondents. *Astronomy*, *science*, *vastness*, *beauty* and *discovery* were all motivation categories that were found to very important to the volunteers, while *fun*, *learning* and *community* were less important.

The AAVSO demographic survey (Price & Paxson 2012) found similar results: over a third of variable star observers cited involvement in science and research as their primary source of motivation. However, a similar number gave an interest in variable stars as theirs, perhaps reflecting a stronger focus on the science questions involved than is present in the Galaxy Zoo community. Both groups of citizen scientists are clearly quite serious in their reasons for taking part: their motivations are actually very close to those of professional scientists, as many readers of this review will recognize. Perhaps surprisingly, the participants in online data analysis citizen science projects seem to a large extent to be a distinct community from those who participate in more traditional amateur astronomical activities; Galaxy Zoo classifiers, for example, are not for the most part regular observers.

**TO-DO: Phil: #51 Update the motivation section with any latest results. Ask the Zooniverse team.**

**TO-DO: Phil: #56 Update the demographics section with any latest results. Ask the Zooniverse team.**

### 6.3 Competition or Collaboration?

As seen in Section 3.3 above, non-astronomical projects may have much to teach us about “gamification” as a motivator – the inclusion, either explicitly or implicitly, of game-like mechanics such as scores, “badges” or other rewards, leader boards, and so on. An early experiment with Galaxy Zoo showed that the addition of a score de-incentivised poor classifiers, but also resulted in the best classifiers leaving, presumably having been satisfied once a top score was achieved. A recent study by Eveleigh et al. of the Old Weather project, which included basic



rankings for classifiers, highlighted these dangers, identifying volunteers who were alienated by the addition of this game-like score, feeling discouraged when top scores could not be matched, or worrying about data quality if the scoring scheme rewarded quantity of classifications rather than accuracy. Taking seriously the above finding that citizen scientists are motivated by a perception of authentic participation in research, it seems right to be cautious about introducing elements which are or which are perceived to be in tension with this primary motivation.

Furthermore, the introduction of a significant incentivizing scheme relies on an accurate model of what ‘correct’ behaviour would look like; this may prove a significant barrier both to accuracy if such a model is not available (for example, in Planet Hunters such a model would not have included unusual systems such as PH1b) and, where a strong incentive scheme results in near-uniform classifier behaviour, a loss of flexibility in later data analysis. A strong comparison of the type of reward structure utilized by Eyewire and the approach used by projects such as Galaxy Zoo is needed, in order to inform future project design.

The surveys described in the previous section reveal a community of people many of whom may have left academic science behind as soon as they finished their education, but who still maintained a passion for astronomy and the boundaries of knowledge. Their thirst for new information, and the desire to be part of the scientific process drives them to actively observe the night sky or to participate in analysis of large datasets.

For the more motivated people involved in citizen science, being part of a community, albeit a distributed one, that brings great enjoyment and satisfaction. With the connectivity of the internet, there is a social aspect of citizen science that unites people with shared interests. These pastimes and hobbies are often

far removed from someone’s “normal” life. However, *community* was not found to be a strong motivator for the Galaxy Zoo volunteers – but it is nevertheless very important for the Galaxy Zoo forum users. More recent Zooniverse projects have sought to widen participation in community discussion, hypothesizing not that it will motivate people better, but because it will help them make better contributions. Citizen scientists, like professional scientists, are primarily motivated by getting science done.

## 7 The Future of Citizen Astronomy (4 pages)

The future of significant citizen science in astrophysics depends to some extent on progress in machine learning; if the “technological singularity” first described by John von Neumann (?), is imminent then there will be little need for either citizen or professional astronomers! A less ambitious astronomical equivalent – the point at which human review of data is no longer necessary – has been forecast for decades, but as we have described above the number of problems for which manual review of images or data is still carried out is enormous. Even if the proportion of data for which human inspection is necessary decreases dramatically over the next decade, the continued growth in the size of astronomical datasets should ensure that there remains plenty for citizen scientists to do. This section discusses what seems to be the likely path that citizen astronomy will take, and in it we also try and identify some key challenges, and possible limitations, that those planning projects are likely to have to face.

Consider the example of optical transients. The LSST system paper (Ivezic et al. 2008) gives a conservative estimate of  $10^5 - 10^6$  alerts per night. Even if, after automated brokerage, only 1% of these require human classification,

then that still might produce 10,000 images requiring inspection every night – roughly one every 100 seconds. Given the increased reliability and likelihood of serendipitous discovery provided by review by citizen scientists, we should take seriously the incorporation of open inspection into plans for LSST transients. Similar arguments (with large error bars) can be made for other surveys, including the vast amounts of data expected to be generated by the Square Kilometer Array; review of transients for LOFAR already requires some human intervention.

Implicit in this way of thinking is a division of task between human and machine classifiers, something which the majority of online classification projects to date have been able to avoid. Galaxy Zoo and similar projects err instead on the side of including the widest possible range of data, in order to make the task of combining classifications simpler; once the data volume or rate grows further, this must be sacrificed for efficiency’s sake. A simple example of human/machine task allocation was mentioned in Section 3.2, where machine analysis of PTF images identified those that contained candidate supernovae needing inspection by volunteers. It is worth noting that the inclusion of human inspection changes the nature of the machine learning task; instead of optimising for purity (producing a small but accurately classified set of candidates), the task for machine learning became one of identifying a subset of the images which contained many false positives but also a complete set of all supernovae.

In this example, human and machine classification proceed in series rather than in parallel, but more complex interactions are possible. The accuracy of machine learning typically depends on the quality of the ‘gold standard’ data which can be provided for the problem in question, and, although citizen science projects can assist by providing training sets which are orders of magnitude larger than

might otherwise have been available, work by ?) established that the confidence intervals provided by classifications from multiple volunteers can also improve accuracy. Predicting human responses (in the form of probabilities of classification) is an easier task than straightforward sorting. We expect, therefore, intermediate-size surveys to benefit in the future from a ‘citizen science’ phase, in which data is classified by volunteers prior to the automation of the task. This pattern has already been followed by the PTF supernova project discussed above, but perhaps it is more useful to think of the citizen scientists as providing training sets on demand, so that as conditions change from night to night, or the performance of the instrument evolves over time, a small percentage of the total data is always processed by humans in order to provide a constantly updated training set.

This approach is complementary to the work discussed in 3.2, which mentioned efforts to improve the efficiency of classification projects by assigning tasks more efficiently. If we are using classifications of gold standard data to assess the performance of human classifiers, it is trivial to include machine classification in the same system. In this way, the task of classification could be split in real time between machine and human classifiers. Significant work has already been carried out for the nearly analogous problem of assigning tasks to an ensemble of imperfect machine classifiers whose characteristics are known (XXXX REF XXX) and for Mechanical Turk-like systems where a fixed payment is provided for a task but the problem of adding in volunteers is significantly harder (XXXX REF XXXX). For the machine-only case, each classification task can be treated as having a known cost (perhaps the processing time necessary for a given routine) but for volunteers who are able to leave whenever they like other costs must be taken into account when assigning tasks. In order to create a viable system it is,

in fact, necessary to measure how interesting a task or set of tasks is, and this requirement may conflict with the need for efficiency. As an example, consider a Galaxy Zoo-like system which assigned the hardest galaxies to the best classifiers. This would result in a steady diet of faint fuzzy objects for the best classifiers; if they are motivated in part by the variety of images seen (which seems a reasonable assumption) this system will systematically drive away its best classifiers.

The development of novel systems for task assignment is therefore necessary to scale citizen science to the challenges for the next generation of surveys. Existing data from both Zooniverse and other projects can help investigate possible schemes, but first experiments suggest that assigning a 'cost' to a single classification will not be easy. For example, one might propose that the number of classifications contributed after seeing each image might be a reasonable measure of the 'cost' of a classification, with the expectation that seeing an especially interesting image might prolong a classification session. For most Zooniverse projects, though, there is little statistical significance in this measure. A study of Snapshot Serengeti even revealed that seeing a more impressive image early in a classifier's career (as measured by the number of volunteers who added it to their list of favourites) decreased the number of classifications received in the long run. Considering individual classifications in isolation is clearly not sufficient; the entirety of a volunteer's career must be considered when assigning tasks. Put more simply, we should be wary of over-specialisation even when efficiency is paramount.

Volunteers can, as we have seen in previous sections, also move beyond simple classification problems and such behaviour will become increasingly important as the volume of data continues to increase. Transient surveys, in particular, will re-

quire rapid decisions about follow-up which could even be carried out by citizens themselves using remote observing facilities with suitable capabilities (XXXXX PHIL : could be a way to put in the text below as part of a continuous section - what do you think?). This requires the appearance of a significantly increased cohort of advanced citizen scientists, many of whom will need to be fluent in the tools of professional astronomy. An example of best practice exists in the way that the Sloan Digital Sky Survey's sky server provided both tools for professional (or advanced) researchers alongside simplified versioned aimed primarily at educational use. This structure has the twin benefits of providing near-seamless transitions from simple to more advanced interfaces, and of providing extra pressure to make the resulting interfaces easily usable, something which benefits all users, not just citizen scientists. Designers of science user interfaces for projects such as LSST would do well to bear these twin audiences in mind.

Strands:

Human-computer partnerships

Scaling up

Advanced, enabling, open-ended/flexible interfaces: gamification and gamified behaviour, bottom-up citizen science

Leading to:

LIVE and AT SCALE

both as strategy to enable rapid follow-up, and coping with data volume (and complexity), and to keep motivation of crowd high.

## 7.1 Possibilities

**TO-DO: Chris: #58 Write up possibilities of future citizen astronomy.**

**Access to networked telescopes**

Sadly, future advances in technology may begin to widen the gap between citizen scientists and professionals once again. These may include distributed arrays of robotic 1m+ class telescopes, operating in remote regions with excellent atmospheric conditions, and trained to observe a target in a regular fashion over multiple nights (e.g., developing a consistent high-quality dataset for cloud tracking on Venus, Mars or the giant planets; or a monitoring network for meteor showers to permit 3D trajectory reconstruction). These networks would allow quasi-continuous observations over 24 hours. Certain robotic telescopes could be dedicated educational platforms, run and maintained by schools and colleges. As the images would be regularised, we could envisage automated software to track features, detect impacts, identify morphological peculiarities over time, replacing the crowd-sourced citizen analysis currently underway. But such an investment will require both international funding and considerable time and effort, and we might question whether it will ever be necessary to replace citizen scientists with automated facilities.

**Pushing observational boundaries**

More exciting will be the advances in hardware available to the individual citizen observers - larger optics, more sensitive cameras, and spectral coverage extending to longer wavelengths in the infrared. Such advances in sensitivity and optics could permit citizen investigations of Uranus and Neptune; the cold and icy bodies in the distant solar system (e.g., Trans-Neptunian objects and the Kuiper Belt), and asteroids and solar system debris closer to home. Transits of extrasolar planets in front of their parent stars would be permitted from modest observatories provided they had stable conditions. New platforms might also

become available to the citizen scientist, including balloon-borne observatories rising up and out of the majority of the atmospheric turbulence to provide crisper and more detailed observations of astronomical targets.

**Survey mining** Huge public databases from wide field surveys: LSST, Euclid, SKA. User interfaces designed for anyone, with social networking enabled.

**Live Visual Classification** Time domain astronomy, live data: dynamic task assignment?

**Human-computer partnerships.** Machine learning in tandem with citizens, in various ways.

**Crowd-sourcing professional data modeling** Easily installed apps or browser-based tools enable outsourcing of data modelling. Operation of code, development of code. Crowd-sourcing of current detailed analyses. Importance of social aspect.

**Bottom-up citizen science.** Web apps so flexible they enable citizen-led enquiry. What might these look like?

**Survey curation** Citizens – amateur survey astronomers – become experts in niche areas within LSST, SKA etc surveys. To be encouraged.

**Large survey image inspection** For diagnosis of processing errors, identification of strange new features.

**Citizen publications** Provide publishing support, see Letters.

## 7.2 Challenges (3 pages)

**TO-DO: Phil: #60 Write up challenges of future citizen astronomy.**

We have argued that a critical part of “citizen science” lies in the ability of



the amateur to make an authentic contribution to science. We now consider the some potential limitations of citizen science – which are perhaps best read as challenges to be overcome.

**Accessibility.** Citizen-friendly interfaces: SDSS sky server a good example for super users. Making online projects easier to set up?

Connections between citizen science and open data, and open publishing.

Citizens reading papers: accessibility, potential barriers.

International CS. Language barriers, cultural issues.

Fast and reliable access to the internet is a pre-requisite for participation, and active observation typically requires financial investment in equipment, biasing the demographics to the developed world. An alternative is participation in educational projects such as the Faulkes telescope, providing access to world-leading equipment for educational purposes.

**Data volume.** LSST numbers. How many images? How many classifiers? How to keep motivation high? Case studies: Large samples of lenses? Transients with SKA?

**Data complexity.** difficult/complex analysis. Encouragement from Kaggle, online modeling, image lucky processing. Problem is tool provision and support. Generic cloud-based tools already there (see PH analysis), specialist tools can be provided and supported. Requires greater Pro-Am collaboration. Microtasking only suitable for certain parts of the process?

**Collaboration.** hard with very large numbers. Need systems to crowd-source the interactions: hierarchies of moderators etc. Broadcast capabilities: Galaxy Zoo hangouts as an example. LSSTV?  
Online discussion forums also arguably provide the most direct connection

between citizens and professional scientists, and have already featured in this review.

Collaboration between professional and citizen astronomers. Does it scale?

Communication issues: forum, letters. Contrast supervisor to student, with scientist to crowd. Prospects for large collaborations? Collaborations between citizens, eventually linked to professionals?

Relationships between citizens and professionals. Mostly one-way? Examples of two-way interactions: zoo forum, solar system monitoring, spacecraft support.

## 8 Concluding Remarks: Characteristics of Successful Citizen Science (1 page)

**TO-DO: Phil: #61 Compose 2-3 paragraphs, or better, an itemized list, of the characteristics of successful citizen science, drawing from the notes already in place.**

To emerge.

Uncompromising stance on value of citizen contributions: focus on tasks that cannot be done by machines or professionals.

Low barrier to entry, easy to make a contribution.

Emphasis on science that volunteer is to be part of.

Good communication between professionals and citizens.

High level of respect for volunteers: think of citizens as collaborators, research assistants.

Leigh writes: “Citizen scientists currently do this for fun and education, and professionals should be wary of demanding more standardised, calibrated and

uniform techniques for temporal and spatial monitoring, as enthusiasm may wane if this begins to sound like ‘work’ rather than a hobby.”

Provide tools that professionals would want to use – alternatively, design professional tools so that any citizen can use them.

The need to understand black box systems – especially if the box is full of people. But: people must be treated as ends in themselves.

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