

almost empty by the time the network pressure increases. As a consequence, the float valve is completely open and allows water to pass at a very high flow rate.

The dependency of the apparent losses on the presence of private tanks and the behaviour of their users has been analysed previously (Rizzo and Cilia, 2005; Cobacho et al., 2008; Criminisi et al., 2009). The effect of a tank on consumption flow rates raises the global error of a meter, ranging from approximately -10% for a new meter to -40% for a worn-out meter. In terms of apparent losses, experimental evidence shows that the average under-registration of worn-out meters ranges from approximately 10 to 50 % of total household consumption (Criminisi et al., 2009).

Several other factors cause water meters to lose their accuracy. Thornton and Rizzo (2002) identified meter wear and tear, incorrect installation, lack of maintenance or calibration, incorrect meter type and class for the current application, incorrect meter sizing and demand patterns as possible causes of meter inaccuracy. Arregui et al. (2005) presented real field and laboratory data on the impact of several parameters on the accuracy of both domestic and industrial water meters and on different meter technologies (single-jet, multi-jet, oscillating-piston, Woltman and Tangential meters). Incorrect mounting position, wear of moving parts, suspended solids and deposits, leaks and user's storage tanks, and partial blockage of the inlet strainer are all said to influence the error curve of domestic water meters (although this varies with the meter technology). In addition to these factors, velocity profile distortions and proper meter sizing also have an effect on water meter error for industrial applications.

The pressure level of the network is not generally included amongst the causes affecting metrological performance, but such a factor is especially important for systems that include private tanks because the rate that water flows into these tanks is driven in part by network pressure and not only by user demand.

The aim of this paper is to provide an initial basis for analysis of the effect of network pressure on ageing water meters' under-registration and subsequent apparent losses. For this investigation, two coupled experimental studies are required: one based on laboratory analyses aimed at obtaining meter error curves that account for meter age

and network pressure, and another based on field studies aimed at analysing user water demand and how it may be modified by network pressure, water scarcity and the idiosyncrasies of a user's private distribution system. As discussed above, the characteristic shape of the error curve shows that for flow rates higher than the transitional value, metering errors are negligible, but they become extremely important and negative for flow rates ranging from the starting flow to the minimum flow rate. For this reason, in the present study, the error curves of several worn-out water meters were defined for different operational pressure values at lower flow rates only. The laboratory tests were performed using a standard test bench. For each pressure value, measured upstream of the meter, meter error at the selected tested flow rates was interpolated to provide an empirical equation of the error curve. The starting flow of each meter was evaluated using this equation, and the effect of pressure on the starting flow was studied. Finally, three demand profiles were determined using an experimental field campaign to assess apparent losses due to meter under-registration: the first profile is of a user with a private tank subject to intermittent supply; the second is of a user with a private tank subject to continuous supply; and the third is of a user without a tank. Apparent losses were calculated for each pressure value as a percentage of consumption at flow rates lower than the meter's starting flow. The results of the experimental study can be used in the planning process to maximise the effectiveness of selective water meter replacement in reducing apparent losses.

2 Laboratory studies

Laboratory experiments were performed to obtain the error curves at low flow rates of 143 worn-out water meters, replaced with new instruments in 2006 by the local Palermo water utility (AMAP S.p.A.). The meters were installed to monitor residential users, who comprise 85 % of the total Palermo water distribution network. The meters are multi-jet, with a diameter of 15 mm and a permanent flow rate, Q_3 , of $1.5 \text{ m}^3 \text{ h}^{-1}$. The meters manufactured before 1997 belong to class B, with the rest belonging to

class C, according to the ISO 4064:1993 standard. Their service lives ranged from 0 (new meter) to 45 yr, which was taken into account when considering the wear and tear on each meter. As shown in Table 1, the meters were divided in nine age classes as defined by service life. The service life of the meters in age classes 7, 8, and 9 is very long. Most utilities, including the Palermo utility, replace meters on a run-to-fail basis, and there is not yet a mandatory requirement to replace water meters. As a consequence, residential water meters can be very old. This policy indicates a lack of planning and can result in important economic losses (Fontanazza et al., 2012).

Figure 2 shows the total water volume measured by the meters during their service lives, another relevant factor influencing meter wear and tear. The water volume measured by the meters is understandably proportional to meter age because of the homogeneity of user demand. This pattern effectively represents the relationship between meter wear and tear and the age of the instrument and justifies the meter classification used in this study (Table 1).

The water meter tests were performed using a test bench in the hydraulics laboratory of the Department of Civil, Environmental, Aerospace and Materials Engineering at the University of Palermo. The test bench is a weight calibration device consistent with the ISO 4064:2005 standard. It consists of a water supply system (mains, one unpresurised tank, two pumps); a test section in which the meter is placed; four fluxmeters to establish the approximate flow rates at which the meter is tested; two pneumatic and automatic gate valves; two pressure gauges to measure the pressure upstream and downstream of the tested meter; a vacuum gauge; two calibrated tanks, each placed on a precision electronic balance; a temperature sensor; and a control panel (Fig. 3). The test bench is connected to a computer for test automation, measurement acquisition and result calculation.

The test method applied to determine the measurement error was the “collection” method (ISO 4064:2005). The water passing through the meter is collected in one of the two calibrated tanks, with the quantity determined by weight. Meter error is defined by Eq. (1), where the actual volume is that collected in the tank and measured by the

precision balances. The volume indicated by the meter, corresponding to the actual volume, is determined by reading the meter when its sensor stops.

The error curve of the 143 water meters for low flow rates were defined for four different pressure values, representing the network pressure measured upstream of the instruments: 0.5, 1.0, 1.5 and 2.0 bar. For each test pressure, the meters were tested at four different experimentally determined flow rates: the first being the highest flow rate at which the meter sensor remains at rest, with the other three at increasing intervals of one litre per hour (i.e. the second flow rate is one litre per hour greater than the first, and so on). Therefore, the testing flow rates were different for each meter and changed with test pressure. A total of 31 water meters were not tested because they were blocked or unreadable (Table 1). Nearly all the meters in the last two groups (age classes 8 and 9) worked despite their very long service lives. As a result, at the end of laboratory analysis, the low flow rate error curves for 112 water meters were defined for each of the four tested pressure values.

3 Field studies

To apply the results of the laboratory study to the evaluation of the relationship between apparent losses, meter age and network pressure level, three consumption profiles were developed to represent different types of users:

- Consumption profile A corresponds to users who have a private tank that is filled during the night and emptied during the day. During the monitoring period, the network was managed via intermittent distribution, which is often applied by utilities to cope with water scarcity (Criminisi et al., 2009). When the network is empty, the user’s consumption is guaranteed by the water collected in the tank. During the service period, water entered the tank at flow rates much higher than common consumption pulses and, for this reason, apparent losses were small.

around the average without any visible dependency on pressure, while k assumes higher values when the test pressure increases, but with variations in a range of 3% around the average value. The effect of these parameters can thus be considered negligible (Table 2).

5 For the different pressure values, the starting flow of each tested meter was calculated using Eq. (2). Estimating Q_{start} from laboratory experiments is difficult because the incipient movement of the mechanism must be noted by the test operator (and so the value is dependent on the operator's experience) and because Q_{start} is affected by the hysteretic behaviour of the initial part of the error curve (Arregui et al., 2006).
10 For this reason, interpolation is more reliable because it is more objective and less dependent on the above-mentioned hysteresis.

Figure 6 illustrates the non-parametric distribution of the starting flow of the meters at each specified test pressure. For each age class plotted in Fig. 6, the points represent the minimum and maximum starting flow registered during the laboratory experiments;
15 the ends of the whiskers represent the 5th and 95th percentiles and the ends of the boxes the 25th and 75th percentiles of meter starting flow, with the thin and thick lines within the boxes representing the median and mean of the empiric distribution, respectively. The short distance between both the maximum and minimum values and between the extremes is typical of age classes in which starting flows are similar for all meters analysed. This behaviour is characteristic of homogeneous classes in which
20 all meters have the same technical features and are often made by the same manufacturer. In contrast, a wide distance between maximum and minimum values and between the extremes, with an overall larger box, is typical of age classes in which meters have different features and manufacturers. The classes in which the median is lower than the mean (e.g. class 3 and class 5) include a few meters with high starting flows and a majority that demonstrate better performance. The oldest classes display the opposite behaviour, with a small number of meters that have retained very good performance with low starting flows, comparable to those of newer meters, and a large majority of meters with high starting flows. In fact, not all meter models deteriorate in
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the same manner and therefore respond to past experiences in different ways (Zhen and Tao, 2008). Finally, classes such as class 4 that have small boxes with a wide distance between the extremes reflect the presence of outliers, i.e. single meters with very low or very high starting flows. Analysis of Fig. 6 shows that starting flow, as well as the dispersion of data around the median, tends to increase with age, except for classes 4
5 and 6. Increasing pressure reduces the starting flow, in terms of both the percentiles and the extremes. This effect appears to be more evident for the older classes, most likely because of the higher absolute values of the starting flows.

To better separate the effects of age and pressure, Fig. 7 shows the average starting flow of each class after interpolation by an exponential law. The exponential regression can be substituted by a linear law (neglecting classes 7, 8 and 9) without reducing the regression R^2 . This shows that the increase in starting flow over time can be explained by a linear function of age. With meter ageing, the starting flow begins to increase proportionally more than age, demonstrating the increasing impact of wear and tear.
10 The effect of pressure on the average starting flow gradually becomes less evident as the meter ages.

Figure 8 shows the dependency of the average starting flow on pressure, with the starting flow obtained under a test pressure of 0.5 bar taken as 1.0. The effect of pressure is significant, with Fig. 8 confirming that increasing pressure reduces the starting flow and consequently the potential apparent losses due to meter under-registration.
20 The effect of pressure on the starting flow is essentially linear, most likely because the effect is related to water flowing around the moving parts of the meter at very low velocities, thus maintaining laminar flow. The newer the meter is, the greater the influence of pressure: for class 1 (0–5 yr), the starting flow at 2.0 bar is only 60% of its value at 0.5 bar, while for class 3, the starting flow is reduced by approximately 15% for the same increase in pressure. The older age classes have relative reductions in starting flow similar to class 3, thus showing that the effect of pressure weakens with ageing. These results imply that controlling background losses in distribution networks by reducing pressure may cause an increase in meter under-registration. Real losses
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Table 1. Age class divisions of the tested meters with indication of the water meters not tested because they were blocked or unreadable.

Age class	Service life [years]	No. meters	%	Metrological class	No. meters not tested
1	0–5	18	12.6	C	0
2	5–10	16	11.2	C	6
3	10–15	19	13.3	B	5
4	15–20	17	11.9	B	4
5	20–25	15	10.5	B	2
6	25–30	19	13.3	B	8
7	30–35	19	13.3	B	5
8	35–40	10	7.0	B	0
9	40–45	10	7.0	B	1
TOT		143	100		31

Table 2. Variation of the parameters *Per* and *k* with meter age class and pressure.

Age class	<i>Per</i> [lh ⁻¹]				<i>k</i> [-]			
	<i>p</i> = 0.5 bar	<i>p</i> = 1.0 bar	<i>p</i> = 1.5 bar	<i>p</i> = 2.0 bar	<i>p</i> = 0.5 bar	<i>p</i> = 1.0 bar	<i>p</i> = 1.5 bar	<i>p</i> = 2.0 bar
1	25.11	25.27	24.96	24.96	1.92	1.94	1.97	2.00
2	34.26	34.26	34.41	34.41	1.90	1.92	1.95	1.99
3	39.22	39.53	39.37	39.53	1.78	1.81	1.84	1.85
4	49.76	50.53	50.84	49.91	1.71	1.74	1.76	1.79
5	53.01	52.86	52.86	53.17	1.65	1.69	1.71	1.76
6	29.45	29.61	29.61	29.76	1.62	1.63	1.66	1.69
7	39.37	39.99	38.91	38.91	1.60	1.61	1.64	1.66
8	36.43	36.58	36.27	36.12	1.56	1.58	1.60	1.63
9	31.31	31.62	31.16	31.78	1.54	1.56	1.57	1.60

Table 5. Apparent losses for each meter age class and test pressure: profile C.

Age class	Apparent losses [%]				Average [%]
	$P = 0.5 \text{ bar}$	$P = 1.0 \text{ bar}$	$P = 1.5 \text{ bar}$	$P = 2.0 \text{ bar}$	
1	4.29	2.69	1.51	1.22	2.43
2	8.24	7.86	7.11	6.37	7.40
3	11.19	10.80	10.53	10.34	10.72
4	9.41	9.00	8.76	8.58	8.94
5	15.82	14.53	13.72	13.37	14.36
6	11.32	10.74	10.59	10.36	10.75
7	15.69	14.77	14.24	13.92	14.65
8	17.57	16.07	15.28	14.61	15.88
9	27.29	25.45	24.05	23.06	24.97

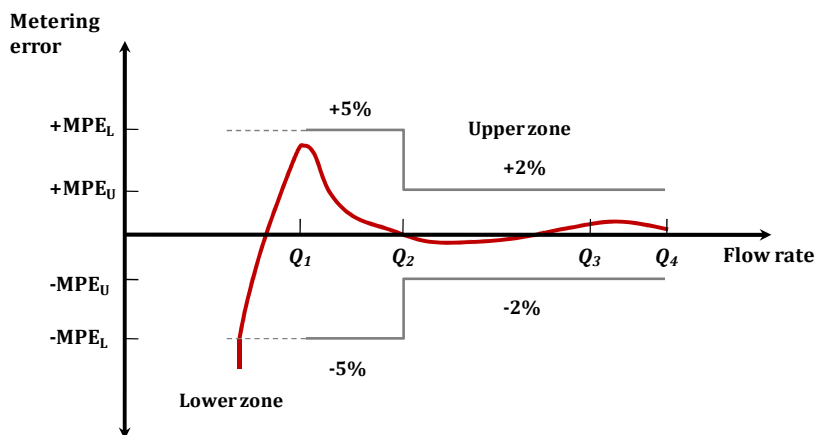


Fig. 1. Sample error curve for a new water meter.

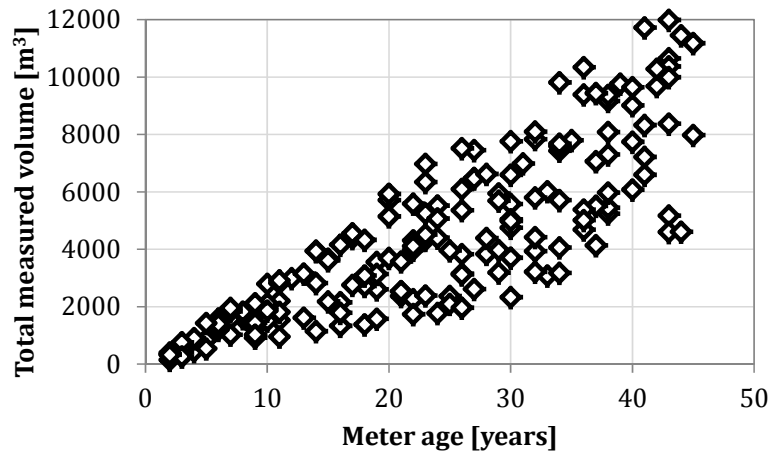


Fig. 2. Total volume measured by the tested meters over their service lives.

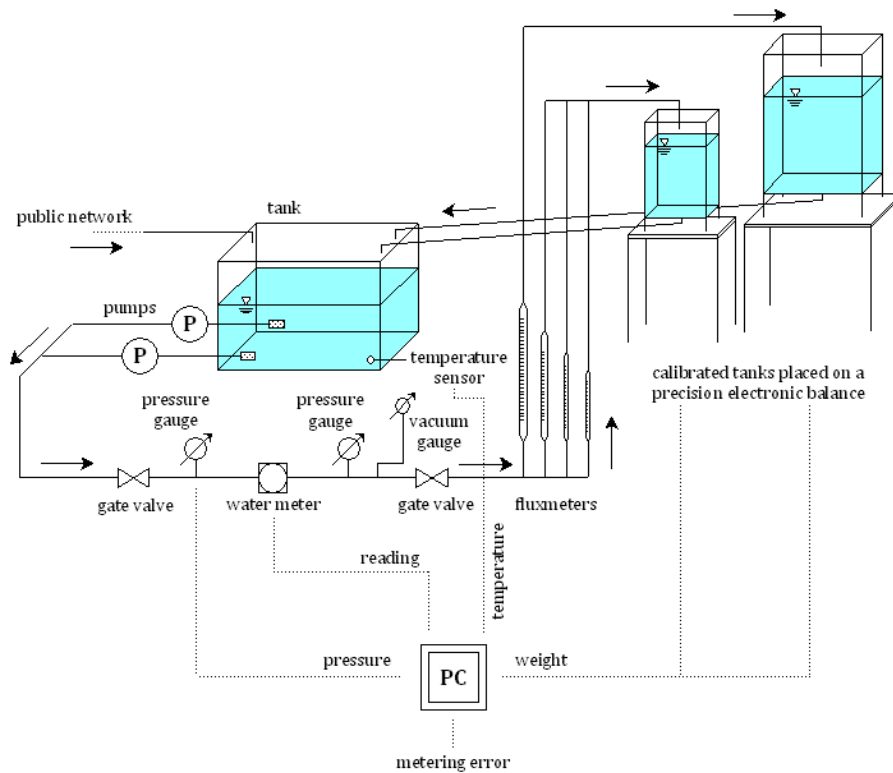


Fig. 3. A schematic of the test bench.

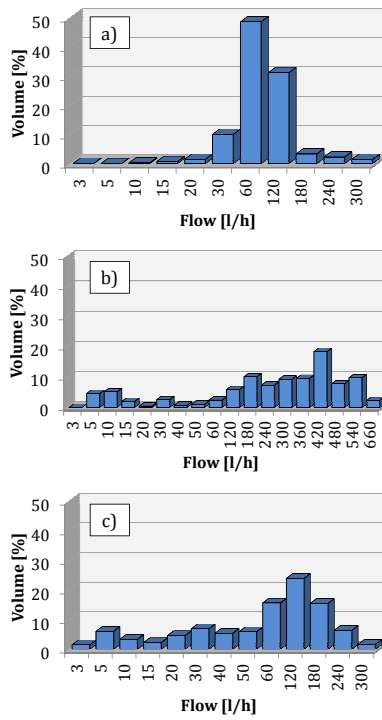


Fig. 4. Consumption profiles used in the study: **(a)** user A; **(b)** user B; **(c)** user C.

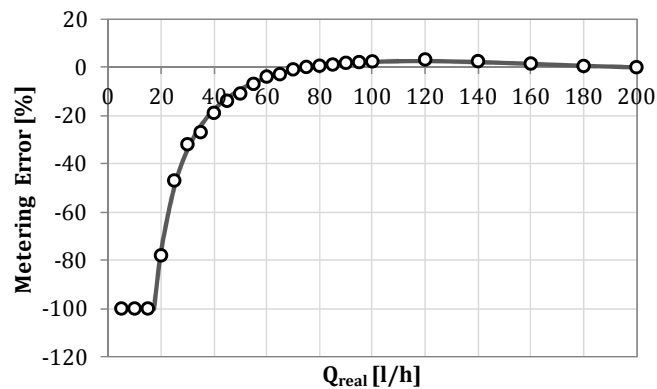


Fig. 5. Example of an error curve for a sample water meter in age class 3 (test pressure $p = 1.0 \text{ bar}$): $Q_{\text{start}} = 17.20 \text{ l h}^{-1}$, $k = 1.84$ and $Per = 38.22 \text{ l h}^{-1}$.

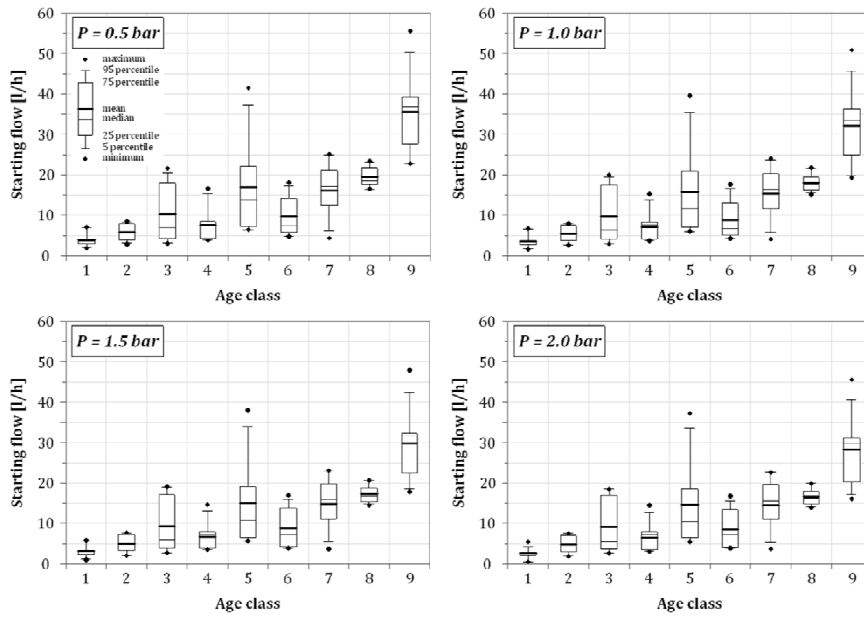


Fig. 6. Box plots summarising the laboratory results.

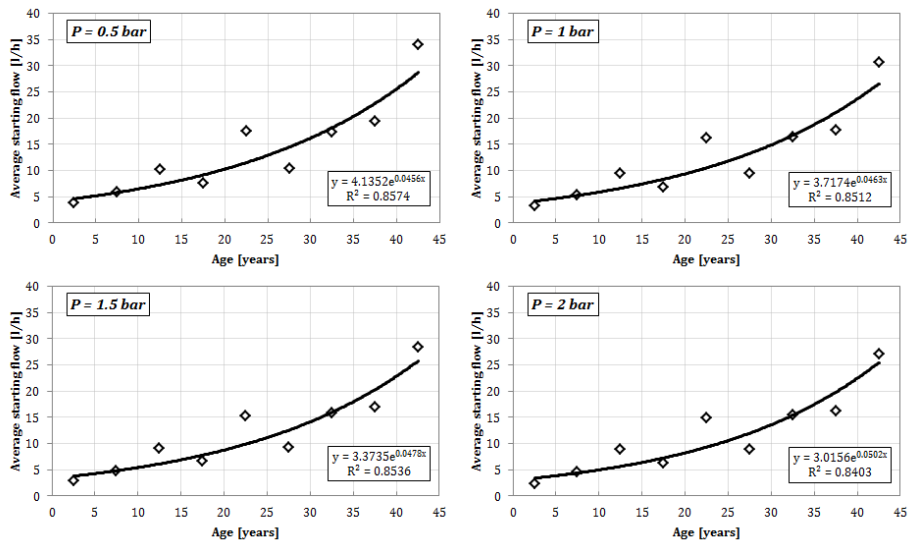


Fig. 7. Relationship between average starting flow and meter age for the four different test pressures.

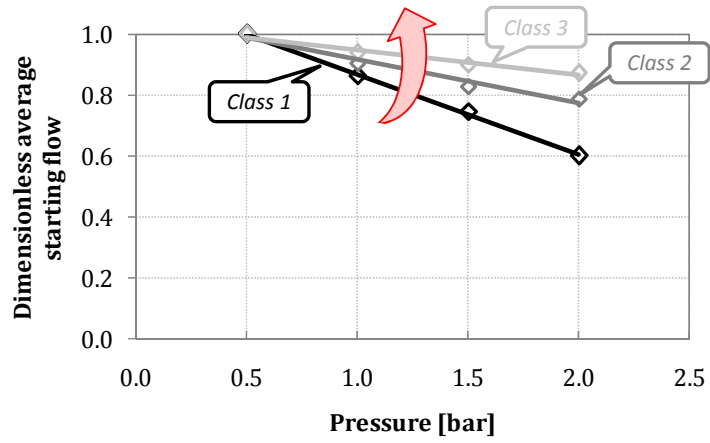


Fig. 8. Relationship between average starting flow and test pressure for the different meter age classes.